## LabWindows/CVI

## LabWindows/CVI Advanced Analysis Library Reference Manual

## Internet Support

E-mail: support@natinst.com
FTP Site: ftp.natinst.com
Web Address: http://www.natinst.com

## Bulletin Board Support

BBS United States: 5127945422
BBS United Kingdom: 01635551422
BBS France: 0148651559

## Fax-on-Demand Support

5124181111

## Telephone Support (USA)

Tel: 5127958248
Fax: 5127945678

## International Offices

Australia 039879 5166, Austria 0662457990 0, Belgium 0275700 20, Brazil 011288 3336,
Canada (Ontario) 905785 0085, Canada (Québec) 514694 8521, Denmark 457626 00, Finland 09725725 11, France 01481424 24, Germany 08974131 30, Hong Kong 2645 3186, Israel 03 6120092, Italy 02 413091, Japan 035472 2970, Korea 02596 7456, Mexico 5520 2635, Netherlands 0348 433466, Norway 328484 00, Singapore 2265886, Spain 91640 0085, Sweden 0873049 70, Switzerland 056200 51 51, Taiwan 02377 1200, United Kingdom 01635523545

## National Instruments Corporate Headquarters

6504 Bridge Point Parkway Austin, Texas 78730-5039 USA Tel: 5127940100

## Important Information

The media on which you receive National Instruments software are warranted not to fail to execute programming instructions, due to defects in materials and workmanship, for a period of 90 days from date of shipment, as evidenced by receipts or other documentation. National Instruments will, at its option, repair or replace software media that do not execute programming instructions if National Instruments receives notice of such defects during the warranty period. National Instruments does not warrant that the operation of the software shall be uninterrupted or error free.
A Return Material Authorization (RMA) number must be obtained from the factory and clearly marked on the outside of the package before any equipment will be accepted for warranty work. National Instruments will pay the shipping costs of returning to the owner parts which are covered by warranty.
National Instruments believes that the information in this manual is accurate. The document has been carefully reviewed for technical accuracy. In the event that technical or typographical errors exist, National Instruments reserves the right to make changes to subsequent editions of this document without prior notice to holders of this edition. The reader should consult National Instruments if errors are suspected. In no event shall National Instruments be liable for any damages arising out of or related to this document or the information contained in it.
Except as specified herein, National Instruments makes no warranties, express or implied, and specifically disclaims any warranty of merchantability or fitness for a particular purpose. Customer's right to recover damages caused by fault or negligence on the part of National Instruments shall be limited to the amount theretofore paid by the customer. National Instruments will not be liable for damages resulting from loss of data, profits, use of products, or incidental or consequential damages, even if advised of the possibility thereof. This limitation of the liability of National Instruments will apply regardless of the form of action, whether in contract or tort, including negligence. Any action against National Instruments must be brought within one year after the cause of action accrues. National Instruments shall not be liable for any delay in performance due to causes beyond its reasonable control. The warranty provided herein does not cover damages, defects, malfunctions, or service failures caused by owner's failure to follow the National Instruments installation, operation, or maintenance instructions; owner's modification of the product; owner's abuse, misuse, or negligent acts; and power failure or surges, fire, flood, accident, actions of third parties, or other events outside reasonable control.

Under the copyright laws, this publication may not be reproduced or transmitted in any form, electronic or mechanical, including photocopying, recording, storing in an information retrieval system, or translating, in whole or in part, without the prior written consent of National Instruments Corporation.

## Trademarks

CVI $^{\mathrm{TM}}$, natinst. com ${ }^{\mathrm{TM}}$, National Instruments ${ }^{\mathrm{TM}}$, the National Instruments logo, and The Software is the Instrument ${ }^{\mathrm{TM}}$ are trademarks of National Instruments Corporation.

Product and company names listed are trademarks or trade names of their respective companies.

## WARNING REGARDING MEDICAL AND CLINICAL USE OF NATIONAL INSTRUMENTS PRODUCTS

National Instruments products are not designed with components and testing intended to ensure a level of reliability suitable for use in treatment and diagnosis of humans. Applications of National Instruments products involving medical or clinical treatment can create a potential for accidental injury caused by product failure, or by errors on the part of the user or application designer. Any use or application of National Instruments products for or involving medical or clinical treatment must be performed by properly trained and qualified medical personnel, and all traditional medical safeguards, equipment, and procedures that are appropriate in the particular situation to prevent serious injury or death should always continue to be used when National Instruments products are being used. National Instruments products are NOT intended to be a substitute for any form of established process, procedure, or equipment used to monitor or safeguard human health and safety in medical or clinical treatment.

## Contents

About This Manual
Organization of This Manual ..... xiii
Conventions Used in This Manual ..... xiii
Related Documentation ..... xiv
Customer Communication ..... xvi
Chapter 1
Advanced Analysis Library Overview
Product Overview ..... 1-1
Advanced Analysis Library Function Panels ..... 1-1
Class and Subclass Descriptions ..... 1-8
Hints for Using Advanced Analysis Function Panels ..... 1-10
Reporting Analysis Errors ..... 1-11
About the Fast Fourier Transform (FFT) ..... 1-11
About Windowing ..... 1-13
About Digital Filters ..... 1-15
FIR Filters ..... 1-16
IIR Filters ..... 1-17
About Measurement Functions ..... 1-19
About Curve Fitting Functions ..... 1-21
About Vector \& Matrix Algebra Functions ..... 1-21
Chapter 2
Advanced Analysis Library Function Reference
Abs1D ..... 2-2
ACDCEstimator. ..... 2-3
Add1D ..... 2-4
Add2D ..... 2-5
AllocIIRFilterPtr ..... 2-6
AmpPhaseSpectrum ..... 2-8
ANOVA1Way ..... 2-10
ANOVA2Way ..... 2-16
ANOVA3Way ..... 2-27
ArbitraryWave ..... 2-42
AutoPowerSpectrum ..... 2-44
BackSub ..... 2-46
Bessel_CascadeCoef ..... 2-48
Bessel_Coef ..... 2-50
BkmanWin ..... 2-52
BlkHarrisWin ..... 2-53
Bw_BPF ..... 2-54
Bw_BSF ..... 2-56
Bw_CascadeCoef ..... 2-58
Bw_Coef ..... 2-60
Bw_HPF ..... 2-62
Bw_LPF ..... 2-64
CascadeToDirectCoef ..... 2-66
Ch BPF ..... 2-68
Ch_BSF ..... 2-70
Ch_CascadeCoef ..... 2-72
Ch Coef ..... 2-74
Ch_HPF ..... 2-76
Ch_LPF ..... 2-78
CheckPosDef ..... 2-80
Chirp ..... 2-81
Cholesky ..... 2-82
Clear1D ..... 2-84
Clip ..... 2-85
ConditionNumber ..... 2-86
Contingency_Table ..... 2-88
Convolve ..... 2-92
Copy1D ..... 2-94
Correlate ..... 2-95
CosTaperedWin ..... 2-97
CrossPowerSpectrum ..... 2-98
CrossSpectrum ..... 2-100
CxAdd ..... 2-102
CxAdd1D ..... 2-103
CxCheckPosDef ..... 2-104
CxCholesky ..... 2-105
CxConditionNumber ..... 2-107
CxDeterminant ..... 2-109
CxDiv ..... 2-111
CxDiv1D ..... 2-112
CxDotProduct ..... 2-113
CxEigenValueVector ..... 2-114
CxExp ..... 2-116
CxGenInvMatrix ..... 2-117
CxGenLinEqs ..... 2-119
CxLinEv1D ..... 2-121
CxLn ..... 2-123
CxLog ..... 2-124
CxLU ..... 2-125
CxMatrixMul ..... 2-127
CxMatrixNorm ..... 2-129
CxMatrixRank ..... 2-131
CxMul ..... 2-133
CxMul1D ..... 2-134
CxOuterProduct ..... 2-135
CxPolyRoots ..... 2-137
CxPow ..... 2-139
CxPseudoInverse ..... 2-140
CxQR ..... 2-142
CxRecip ..... 2-144
CxSpecialMatrix ..... 2-145
CxSqrt ..... 2-148
CxSub. ..... 2-149
CxSub1D ..... 2-150
CxSVD ..... 2-151
CxSVDS ..... 2-153
CxTrace ..... 2-154
CxTranspose ..... 2-155
Decimate ..... 2-156
Deconvolve ..... 2-157
Determinant ..... 2-158
Difference ..... 2-159
Div1D. ..... 2-161
Div2D ..... 2-162
DotProduct ..... 2-163
Elp_BPF ..... 2-164
Elp_BSF ..... 2-166
Elp_CascadeCoef ..... 2-168
Elp_Coef ..... 2-170
Elp_HPF ..... 2-172
Elp_LPF ..... 2-174
Equi_Ripple ..... 2-176
EquiRpl_BPF ..... 2-180
EquiRpl_BSF ..... 2-182
EquiRpl_HPF ..... 2-184
EquiRpl_LPF ..... 2-186
ExBkmanWin. ..... 2-188
ExpFit. ..... 2-189
ExpWin ..... 2-191
F_Dist. ..... 2-192
FFT ..... 2-193
FHT. ..... 2-195
FIR_Coef ..... 2-197
FlatTopWin ..... 2-199
ForceWin ..... 2-200
ForwSub ..... 2-201
FreeAnalysisMem ..... 2-203
FreeIIRFilterPtr ..... 2-204
GaussNoise ..... 2-205
GenCosWin ..... 2-206
GenDeterminant ..... 2-207
GenEigenValueVector ..... 2-209
GenInvMatrix ..... 2-211
GenLinEqs ..... 2-213
GenLSFit ..... 2-215
GenLSFitCoef ..... 2-224
GetAnalysisErrorString ..... 2-227
HamWin ..... 2-228
HanWin. ..... 2-229
HarmonicAnalyzer ..... 2-230
Histogram ..... 2-232
IIRCascadeFiltering ..... 2-234
IIRFiltering ..... 2-236
Impulse ..... 2-238
ImpulseResponse ..... 2-239
Integrate ..... 2-241
InvCh_BPF ..... 2-243
InvCh_BSF ..... 2-245
InvCh_CascadeCoef ..... 2-247
InvCh_Coef ..... 2-249
InvCh_HPF ..... 2-251
InvCh_LPF ..... 2-253
InvF_Dist ..... 2-255
InvFFT ..... 2-257
InvFHT ..... 2-259
InvMatrix ..... 2-261
InvN Dist ..... 2-262
InvT_Dist ..... 2-263
InvXX_Dist ..... 2-264
Ksr_BPF ..... 2-265
Ksr_BSF ..... 2-267
Ksr_HPF ..... 2-269
Ksr_LPF ..... 2-271
KsrWin ..... 2-273
LinEqs ..... 2-275
LinEv1D ..... 2-276
LinEv2D ..... 2-277
LinFit ..... 2-278
LU ..... 2-280
MatrixMul ..... 2-282
MatrixNorm ..... 2-284
MatrixRank ..... 2-286
MaxMin1D ..... 2-288
MaxMin2D ..... 2-289
Mean ..... 2-291
Median ..... 2-292
Mode ..... 2-293
Moment ..... 2-294
Mul1D ..... 2-296
Mul2D ..... 2-297
N_Dist ..... 2-298
Neg1D ..... 2-299
NetworkFunctions ..... 2-300
NonLinearFit ..... 2-303
NonLinearFitWithMaxIters ..... 2-305
Normal1D ..... 2-307
Normal2D ..... 2-309
NumericIntegration ..... 2-311
OuterProduct ..... 2-314
PeakDetector ..... 2-315
PolyEv1D ..... 2-318
PolyEv2D ..... 2-320
PolyFit ..... 2-322
PolyInterp ..... 2-324
PowerFrequencyEstimate ..... 2-326
Prod1D ..... 2-329
PseudoInverse ..... 2-330
Pulse ..... 2-332
PulseParam ..... 2-334
QR ..... 2-337
QScale1D ..... 2-339
QScale2D ..... 2-340
Ramp ..... 2-341
RatInterp ..... 2-343
ReFFT ..... 2-345
ReInvFFT ..... 2-346
ResetIIRFilter ..... 2-347
Reverse ..... 2-349
RMS ..... 2-350
SawtoothWave ..... 2-351
Scale1D ..... 2-353
Scale2D ..... 2-355
ScaledWindow ..... 2-357
Set1D ..... 2-359
Shift ..... 2-360
Sinc ..... 2-362
SinePattern ..... 2-363
SineWave ..... 2-365
Sort ..... 2-367
SpecialMatrix ..... 2-368
Spectrum ..... 2-371
SpectrumUnitConversion ..... 2-372
SpInterp ..... 2-376
Spline ..... 2-378
SquareWave ..... 2-380
StdDev ..... 2-382
Sub1D ..... 2-383
Sub2D ..... 2-384
Subset1D ..... 2-385
Sum1D ..... 2-386
Sum2D ..... 2-387
SVD ..... 2-388
SVDS ..... 2-390
SymEigenValueVector ..... 2-391
T_Dist ..... 2-393
ToPolar ..... 2-394
ToPolar1D ..... 2-395
ToRect ..... 2-396
ToRect1D ..... 2-397
Trace ..... 2-398
TransferFunction ..... 2-399
Transpose ..... 2-401
Triangle ..... 2-402
TriangleWave ..... 2-403
TriWin ..... 2-405
Uniform ..... 2-406
UnWrap1D ..... 2-407
Variance ..... 2-408
WhiteNoise ..... 2-409
Wind_BPF ..... 2-410
Wind_BSF ..... 2-412
Wind_HPF ..... 2-414
Wind_LPF ..... 2-416
XX_Dist ..... 2-418
Appendix A
Error Codes
Appendix B
Customer Communication
Glossary
Index
Figures
Figure 1-1. Windowed Spectrum in the Continuous Case ..... 1-14
Figure 1-2. Cascaded Filter Stages. ..... 1-18
Tables
Table A-1. Advanced Analysis Library Error Codes, Sorted Alphabetically ..... A-1
Table A-2. Advanced Analysis Library Error Codes, Sorted Numerically ..... A-4

## About This Manual

The LabWindows/CVI Advanced Analysis Library Reference Manual describes the functions in the LabWindows/CVI Advanced Analysis Library. To use this manual effectively, you should be familiar with the material presented in the LabWindows/CVI User Manual and with the LabWindows/CVI software. Please refer to the LabWindows/CVI User Manual for specific instructions on operating LabWindows/CVI.

## Organization of This Manual

The LabWindows/CVI Advanced Analysis Library Reference Manual is organized as follows:

- Chapter 1, Advanced Analysis Library Overview, contains a brief product overview and general information about the Advanced Analysis Library functions and panels.
- Chapter 2, Advanced Analysis Library Function Reference, contains a brief explanation of each of the functions in the LabWindows/CVI Advanced Analysis Library in alphabetical order.
- Appendix A, Error Codes, contains error codes the Advanced Analysis Library functions return.
- Appendix B, Customer Communication, contains forms you can use to request help from National Instruments or to comment on our products and manuals.
- The Glossary contains an alphabetical list and description of terms used in this manual, including abbreviations, acronyms, metric prefixes, mnemonics, and symbols.
- The Index contains an alphabetical list of key terms and topics in this manual, including the page where you can find each one.


## Conventions Used in This Manual

The following conventions are used in this manual:
This icon to the left of bold italicized text denotes a note, which alerts you to important information.


This icon to the left of bold italicized text denotes a caution, which advises you of precautions to take to avoid injury, data loss, or a system crash.

| bold | Bold text denotes the names of menus, menu items, parameters, dialog box <br> buttons, and 1D and 2D arrays. 1D arrays appear in lowercase, and <br> 2D arrays appear in uppercase. |
| :--- | :--- |
| bold italic | Bold italic text denotes a note or caution. |
| italic | Italic text denotes variables, emphasis, a cross reference, an introduction to <br> a key concept, or a single number or one element of an array or a matrix. <br> Parameter names in formulas appear in italic text. This font also denotes <br> text from which you supply the appropriate word or value. |
| monospace | Text in this font denotes text or characters that you should literally enter <br> from the keyboard, sections of code, programming examples, and syntax <br> examples. This font is also used for the proper names of disk drives, paths, <br> directories, programs, functions, filenames and extensions, and for <br> statements and comments taken from programs. |
| monospace italic | Italic text in this font denotes that you must enter the appropriate words or <br> values in the place of these items. |

## Related Documentation

The following documents contain information you might find helpful as you use advanced analysis functions.

- Baher, H. Analog \& Digital Signal Processing. New York: John Wiley \& Sons, 1990.
- Bates, D.M., and Watts, D.G. Nonlinear Regression Analysis and its Applications. New York: John Wiley \& Sons, 1988.
- Bracewell, R.N. "Numerical Transforms." Science. Science-248. 11 May, 1990.
- Burden, R.L., and Faires, J.D. Numerical Analysis, 3rd ed. Boston: Prindle, Weber \& Schmidt, 1985.
- Chen, C.H., et al. Signal Processing Handbook. New York: Marcel Dekker, Inc., 1988.
- DeGroot, M. Probability and Statistics, 2nd ed. Reading, MA: Addison-Wesley Publishing Co., 1986.
- Dowdy, S., and Wearden, S. Statistics for Research, 2nd ed. New York: John Wiley \& Sons, 1991.
- Dudewicz, E.J., and Mishra, S.N. Modern Mathematical Statistics. New York: John Wiley \& Sons, 1988.
- Duhamel, P., et al. "On Computing the Inverse DFT." IEEE Transactions on ASSP. ASSP-34 (1986): 1 (February).
- Dunn, O., and Clark, V. Applied Statistics: Analysis of Variance and Regression, 2nd ed. New York: John Wiley \& Sons, 1987.
- Elliot, D.F. Handbook of Digital Signal Processing Engineering Applications. San Diego: Academic Press, 1987.
- Golub, G.H., and VanLoan, C.F. Matrix Computations, 2nd ed. Baltimore: The Johns Hopkins University Press, 1989.
- Harris, Fredric J. "On the Use of Windows for Harmonic Analysis with the Discrete Fourier Transform," Proceedings of the IEEE-66 (1978)-1.
- Maisel, J.E. "Hilbert Transform Works With Fourier Transforms to Dramatically Lower Sampling Rates." Personal Engineering and Instrumentation News. PEIN-7 (1990): 2 (February).
- McClellan, J.H. "A Computer Program for Designing Optimum FIR Linear Phase Digital Filters," IEEE Transactions on Audio and Electroacoustics. AU-21 (1973): (December).
- Miller, I., and Freund, J.E. Probability and Statistics for Engineers. Englewood Cliffs, NJ: Prentice-Hall, Inc., 1987.
- Neter, J., et al. Applied Linear Regression Models. Richard D. Irwin, Inc., 1983.
- Neuvo, Y., Dong, C.-Y., and Mitra, S.K. "Interpolated Finite Impulse Response Filters," IEEE Transactions on ASSP. ASSP-32 (1984): 6 (June).
- O’Neill, M.A. "Faster Than Fast Fourier." BYTE. (1988) (April).
- Oppenheim, A.V., and Schafer, R.W. Discrete-Time Signal Processing. Englewood Cliffs, NJ: Prentice-Hall, Inc., 1989.
- Parks, T.W., and Burrus, C.S. Digital Filter Design. New York: John Wiley \& Sons, 1987.
- Pearson, C.E. Numerical Methods in Engineering and Science. New York: Van Nostrand Reinhold Co., 1986.
- Press, W.H., et al. Numerical Recipes in C: The Art of Scientific Computing. Cambridge: Cambridge University Press, 1988.
- Rabiner, L.R., and Gold, B. Theory and Application of Digital Signal Processing. Englewood Cliffs, NJ: Prentice-Hall, Inc., 1975.
- Sorensen, H.V., et al. "On Computing the Split-Radix FFT." IEEE Transactions on ASSP. ASSP-34 (1986):1 (February).
- Sorensen, H.V., et al. "Real-Valued Fast Fourier Transform Algorithms." IEEE Transactions on ASSP. ASSP-35 (1987): 6 (June).
- Stoer, J., and Bulirsch, R. Introduction to Numerical Analysis. New York: Springer-Verlag, 1987.
- Vaidyanathan, P.P. Multirate Systems and Filter Banks. Englewood Cliffs, NJ: Prentice-Hall, Inc., 1993.
- Wichman, B., and Hill, D. "Building a Random-Number Generator: A Pascal routine for very-long-cycle random-number sequences." BYTE, March 1987, pp. 127-128.


## Customer Communication

National Instruments wants to receive your comments on our products and manuals. We are interested in the applications you develop with our products, and we want to help if you have problems with them. To make it easy for you to contact us, this manual contains comment and configuration forms for you to complete. These forms are in Appendix B, Customer Communication, at the end of this manual.

## Advanced Analysis Library Overview

This chapter contains a brief product overview and general information about the Advanced Analysis Library functions and panels.

## Product Overview

The LabWindows Advanced Analysis Library adds additional analysis functions to the standard LabWindows/CVI Analysis Library. The Advanced Analysis Library includes functions for signal generation, one-dimensional (1D) and two-dimensional (2D) array manipulation, complex operations, signal processing, statistics, curve-fitting, and matrix operations.

## Advanced Analysis Library Function Panels

The Advanced Analysis Library function panels are grouped in the following tree structure according to the types of operations they perform.

The first- and second-level headings in the tree are the names of function classes and subclasses. Function classes and subclasses are groups of related function panels. The third-level headings are the names of individual function panels. Each analysis function panel generates one analysis function call.

The following shows the structure of the Advanced Analysis Library function tree.

Signal Generation<br>Array Operations<br>1D Operations<br>2D Operations<br>Complex Operations<br>Complex Numbers<br>1D Complex Operations<br>Signal Processing<br>Frequency Domain<br>Time Domain

Signal Processing (continued)
IIR Digital Filters
Cascade Filter Functions
Filter Information Utilities
One-Step Filter Functions
Old-Style Filter Functions
FIR Digital Filters
Windows
Measurement
Statistics
Basics
Probability Distributions
Analysis of Variance
Nonparametric Statistics
Curve Fitting
OldStyle Function
Interpolation
Vector \& Matrix Algebra
Real Matrices
Complex Matrices
Additional Numerical Methods
Table 1-1. Functions in the Advanced Analysis Library Overview Function Tree
Class/Panel Name
Signal Generation
ImpulsePulseRampTriangleSine Pattern
Uniform Noise
White Noise
Gaussian Noise
Arbitrary Wave
Chirp
Sawtooth Wave
Sinc Waveform
Sine Wave
Square Wave
Triangle Wave
Array Operations
1D Operations
1D Clear Array
1D Set Array
1D Copy Array
1D Array Addition
1D Array Subtraction
Function Name

Function Name

Impulse
Pulse
Ramp
Triangle
SinePattern
Uniform
WhiteNoise
GaussNoise
ArbitraryWave
Chirp
SawtoothWave
Sinc
SineWave
SquareWave
TriangleWave

Clear1D
Set1D
Copy1D
Add1D
Sub1D

Table 1-1. Functions in the Advanced Analysis Library Overview Function Tree (Continued)

## Class/Panel Name

Array Operations (continued) 1D Operations (continued)

1D Array Multiplication
1D Array Division
1D Absolute Value
1D Negative Value
1D Linear Evaluation
1D Polynomial Evaluation
1D Scaling
1D Quick Scaling
1D Maximum \& Minimum
1D Sum of Elements
1D Product of Elements
1D Array Subset
1D Reverse Array Order
1D Shift Array
1D Clip Array
1D Sort Array
1D Vector Normalization
2D Operations
2D Array Addition
2D Array Subtraction
2D Array Multiplication
2D Array Division
2D Linear Evaluation
2D Polynomial Evaluation
2D Scaling
2D Quick Scaling
2D Maximum \& Minimum
2D Sum of Elements
2D Matrix Normalization
Complex Operations
Complex Numbers
Complex Addition
Complex Subtraction
Complex Multiplication
Complex Division
Complex Reciprocal
Complex Square Root
Complex Logarithm
Complex Natural Log
Complex Power
Complex Exponential
Rectangular to Polar
Polar to Rectangular

## Function Name

Mul1D
Div1D
Abs1D
Neg1D
LinEv1D
PolyEv1D
Scale1D
QScale1D
MaxMin1D
Sum1D
Prod1D
Subset1D
Reverse
Shift
Clip
Sort
Normal1D
Add2D
Sub2D
Mul2D
Div2D
LinEv2D
PolyEv2D
Scale2D
QScale2D
MaxMin2D
Sum2D
Normal2D

CxAdd
CxSub
CxMul
CxDiv
CxRecip
CxSqrt
CxLog
CxLn
CxPow
CxExp
ToPolar
ToRect

Table 1-1. Functions in the Advanced Analysis Library Overview Function Tree (Continued)
Class/Panel Name
Complex Operations (continued)
1D Complex Operations
1D Complex Addition
1D Complex Subtraction
1D Complex Multiplication
1D Complex Division
1D Complex Linear Evaluation
1D Rectangular to Polar
1D Polar to Rectangular
Signal Processing
Frequency Domain
FFT
Inverse FFT
Real Valued FFT
Real Valued Inverse FFT
Power Spectrum
FHT
Inverse FHT
Cross Spectrum
Time Domain
Convolution
Correlation
Integration
Differentiate
Pulse Parameters
Decimate
Deconvolve
Unwrap Phase
IIR Digital Filters
Cascade Filter Functions
Bessel Cascade Coeff
Butterworth Cascade CoeffChebyshev Cascade CoeffInv Chebyshev Cascade CoeffElliptic Cascade CoeffsIIR Cascade FilteringFilter Information UtilitiesAllocate Filter InformationReset Filter InformationFree Filter InformationCascade to Direct Coefficients
One-Step Filter Functions
Lowpass Butterworth
Highpass ButterworthBandpass Butterworth

## Function Name

CxAdd1D
CxSub1D
CxMul1D
CxDiv1D
CxLinEv1D
ToPolar1D
ToRect1D

## FFT

InvFFT
Refft
ReInvFFT
Spectrum
FHT
InvFHT
CrossSpectrum
Convolve
Correlate
Integrate
Difference
PulseParam
Decimate
Deconvolve
UnWrap1D

```
Bessel_CascadeCoef
Bw_CascadeCoef
Ch_CascadeCoef
InvCh_CascadeCoef
Elp_CascadeCoef
IIRCascadeFiltering
```

AllocIIRFilterPtr
ResetIIRFilter
FreeIIRFilterPtr
CascadeToDirectCoef
Bw_LPF
Bw_HPF
Bw_BPF

Table 1-1. Functions in the Advanced Analysis Library Overview Function Tree (Continued)

## Class/Panel Name

Signal Processing (continued)
IIR Digital Filters (continued)
One-Step Filter Functions (continued) Bandstop Butterworth Lowpass Chebyshev Highpass Chebyshev Bandpass Chebyshev Bandstop Chebyshev Lowpass Inverse Chebyshev Highpass Inverse Chebyshev Bandpass Inverse Chebyshev Bandstop Inverse Chebyshev Lowpass Elliptic Highpass Elliptic Bandpass Elliptic Bandstop Elliptic
Old-Style Filter Functions
Bessell Coefficients
Butterworth Coefficients
Chebyshev Coefficients
Inverse Chebyshev Coefficients
Elliptic Coefficients
IIR Filtering
FIR Digital Filters
Lowpass Window Filters
Highpass Window Filters
Bandpass Window Filters
Bandstop Window Filters
Lowpass Kaiser Window
Highpass Kaiser Window
Bandpass Kaiser Window
Bandstop Kaiser Window
General Equi-Ripple FIR
Lowpass Equi-Ripple FIR
Highpass Equi-Ripple FIR
Bandpass Equi-Ripple FIR
Bandstop Equi-Ripple FIR
FIR Coefficients
Windows
Triangular Window
Hanning Window
Hamming Window
Blackman Window
Kaiser Window
Blackman-Harris Window

## Function Name

Bw_BSF
Ch_LPF
Ch_HPF
Ch_BPF
Ch_BSF
InvCh_LPF
InvCh_HPF
InvCh_BPF
InvCh_BSF
Elp_LPF
Elp_HPF
Elp_BPF
Elp_BSF

```
Bessell_Coef
Bw_Coef
Ch_Coef
InvCh_Coef
Elp_Coef
IIRFiltering
```

Wind_LPF
Wind_HPF
Wind_BPF
Wind_BSF
Ksr_LPF
Ksr_HPF
Ksr_BPF
Ksr_BSF
Equi_Ripple
EquiRpl_LPF
EquiRpl_HPF
EquiRpl_BPF
EquiRpl_BSF
FIR_Coef
TriWin
HanWin
HamWin
BkmanWin
KsrWin
BlkHarrisWin

Table 1-1. Functions in the Advanced Analysis Library Overview Function Tree (Continued)
Class/Panel Name
Signal Processing (continued)Windows (continued)Tapered Cosine WindowExact Blackman WindowExponential Window
Flat Top Window
Force Window
General Cosine Window
Measurement
AC/DC EstimatorAmplitude/Phase Spectrum
Auto Power Spectrum
Cross Power Spectrum
Impulse Response
Network Functions
Power Frequency Estimate
Scaled Window
Spectrum Unit Conversion
Transfer Function
Total Harmonic Distortion
Statistics
Basics
Mean
Standard Deviation
Variance
Root Mean Squared Value
Moments about the Mean
Median
Mode
Histogram
Probability Distributions
Normal Distribution
T-Distribution
F-Distribution
Chi-Square Distribution
Inv. Normal Distribution
Inv. T-Distribution
Inv. F-Distribution
Inv. Chi-Square Dist.
Analysis of Variance
One-way ANOVA
Two-way ANOVA
Three-way ANOVA
Nonparametric Statistics

## Function Name

```
CosTaperedWin
ExBkmanWin
ExpWin
FlatTopWin
ForceWin
GenCosWin
ACDCEstimator
AmpPhaseSpectrum
AutoPowerSpectrum
CrossPowerSpectrum
ImpulseResponse
NetworkFunctions
PowerFrequencyEstimate
ScaledWindow
SpectrumUnitConversion
TransferFunction
HarmonicAnalyzer
```

Mean
StdDev
Variance
RMS
Moment
Median
Mode
Histogram
N_Dist
T_Dist
F_Dist
XX_Dist
InvN_Dist
InvT_Dist
InvF_Dist
InvXX_Dist
ANOVA1Way
ANOVA2Way
ANOVA3Way
Contingency_Table

Table 1-1. Functions in the Advanced Analysis Library Overview Function Tree (Continued)

## Class/Panel Name

Curve Fitting
Linear Fit
Exponential Fit
Polynomial Fit
General Least Squares Fit
Non-Linear Fit
Non-Linear Fit with Maximum Iterations
OldStyle Function
Gen Least Squares Fit Coeff
Interpolation
Polynomial Interpolation
Rational Interpolation
Spline Interpolation
Spline Interpolant
Vector \& Matrix Algebra
Real Matrices
Create Special Matrix
Dot Product
Transpose
Determinant
Determinant (General)
Trace
Invert Matrix
Invert Matrix (General)
Solution of Linear Equations
Solution of Linear Eqs (General)
Multiply Matrices
Outer Product
Rank
Norm
Condition Number
Eigenvalues \& Eigenvectors (Symmetric)
Eigenvalues \& Eigenvectors (General)
Singular Values of a Matrix
SVD Factorization
QR Factorization
Cholesky Factorization
PseudoInverse Matrix
Test Positive Definiteness
LU Decomposition
Forward Substitution
Backward Substitution
Complex Matrices
Create Special Complex Matrix
Complex Dot Product

## Function Name

LinFit
ExpFit
PolyFit
GenLSFit
NonLinearFit
NonLinearFitWithMaxIters

GenLSFitCoef

PolyInterp
RatInterp
SpInterp
Spline

SpecialMatrix
DotProduct
Transpose
Determinant
GenDeterminant
Trace
InvMatrix
GenInvMatrix
LinEqs
GenLinEqs
MatrixMul
OuterProduct
MatrixRank
MatrixNorm
ConditionNumber
SymEigenValueVector
GenEigenValueVector
SVDS
SVD
QR
Cholesky
PseudoInverse
CheckPosDef
LU
ForwSub
BackSub

CxSpecialMatrix
CxDotProduct

Table 1-1. Functions in the Advanced Analysis Library Overview Function Tree (Continued)

## Class/Panel Name

Vector \& Matrix Algebra (continued)
Complex Matrices (continued)
Complex Transpose
Complex Determinant
Complex PseudoInverse Matrix
Complex Trace
Complex Invert Matrix
Solution of Complex Linear Eqs
Complex Multiply Matrices
Complex Outer Product
Complex Rank
Complex Norm
Complex Condition Number
Complex Eigenvalues \& Eigenvectors
Complex Singular Values
Complex SVD Factorization
Complex QR Factorization
Complex Cholesky Factorization
Complex Test Positive Definite
Complex LU Factorization
Additional Numerical Methods
Complex Polynomial Roots
Numeric Integration
Peak Detector
Free Analysis Memory
Get Error String

## Function Name

CxTranspose
CxDeterminant
CxPseudoInverse
CxTrace
CxGenInvMatrix
CxGenLinEqs
CxMatrixMul
CxOuterProduct
CxMatrixRank
CxMatrixNorm
CxConditionNumber
CxEigenValueVector
CxSVDS
CxSVD
CxQR
CxCholesky
CxCheckPosDef
CxLU
CxPolyRoots
NumericIntegration
PeakDetector

FreeAnalysisMem
GetAnalysisErrorString

## Class and Subclass Descriptions

- The Signal Generation function panels initialize arrays with predefined patterns.
- The Array Operations function panels perform arithmetic operations on 1D and 2D arrays.
- 1D Operations, a subclass of Array Operations, contains function panels that perform 1D array arithmetic.
- 2D Operations, a subclass of Array Operations, contains function panels that perform 2D array arithmetic.
- The Complex Operations function panels perform complex arithmetic operations. These function panels can operate on complex scalars or 1D arrays. The functions process the real and imaginary parts of complex numbers separately.
- Complex Numbers, a subclass of Complex Operations, contains function panels that perform scalar complex arithmetic.
- 1D Complex Operations, a subclass of Complex Operations, contains function panels that perform complex arithmetic on 1D complex arrays.
- The Signal Processing function panels perform data analysis in the frequency domain, time domain, or by using digital filters.
- Frequency Domain, a subclass of Signal Processing, contains function panels that perform transformations between the time domain and the frequency domain and that perform analysis in the frequency domain.
- Time Domain, a subclass of Signal Processing, contains function panels that perform direct time series analysis of signals.
- IIR Digital Filters, a subclass of Signal Processing, contains function panels that perform infinite impulse response (IIR) digital filtering on signals by mapping analog specifications into digital specifications. This subclass contains Butterworth, Chebyshev, inverse Chebyshev, and elliptic filters.
- FIR Digital Filters, a subclass of Signal Processing, contains function panels that perform the designs of finite impulse response (FIR) filters. These functions do not actually perform the digital filtering. This subclass contains window and equi-ripple FIR filters.
- Windows, a subclass of Signal Processing, contains function panels that create windows that are frequently used to smooth data and reduce truncation effects in data acquisition applications.
- The Measurement function panels perform spectrum analysis, using real units such as hertz and seconds, and total harmonic distortion analysis.
- The Statistics function panels perform basic statistics functions.
- Basics, a subclass of Statistics, contains function panels that use various common methods to describe a set of data.
- Probability Distributions, a subclass of Statistics, contains function panels that operate as cumulative distribution functions from various probability distributions and contains other function panels that operate as corresponding inverse functions.
- Analysis of Variance, a subclass of Statistics, contains function panels that perform various analysis of variance in various statistical models.
- Nonparametric Statistics, a subclass of Statistics, contains a function panel that analyzes data without assuming that the data is normally distributed.
- The Curve Fitting function panels perform curve fitting using least squares techniques. Linear, exponential, polynomial, and nonlinear fits are available.
- The Interpolation function panels take a set of points at which a function is known and guess the value the function takes at a specific intermediate point.
- The Vector \& Matrix Algebra function panels perform vector and matrix operations. Vectors and matrices are represented by 1D and 2D arrays, respectively.
- Real Matrices, a subclass of Vector \& Matrix Algebra, contains functions that operate on real-valued matrices and vectors.
- Complex Matrices, a subclass of Vector \& Matrix Algebra, contains functions that operate on matrices and vectors that contain complex numbers. The complex numbers are represented in the form of a structure that the ComplexNum typedef defines.
- The Additional Numerical Methods function panels perform operations such as numeric integration and peak detection that are widely used in signal processing applications.

The online help with each panel contains specific information about operating each function panel.

## Hints for Using Advanced Analysis Function Panels

With the analysis function panels, you can manipulate scalars and arrays of data interactively. You might find it helpful to use the Advanced Analysis Library function panels in conjunction with the User Interface Library functions panels to view the results of analysis routines. When using the Advanced Analysis Library function panels, remember the following:

- The computer on which you run LabWindows/CVI affects the processing speed of the analysis functions. A numeric coprocessor, especially, increases the speed of floating-point computations. If you are using an Analysis Library function panel and nothing seems to happen for an unusually long time, remember the constraints of your hardware.
- Many analysis routines for arrays run in place. That is, the functions can store the input and output data in the same array. This point is important to keep in mind when you process large amounts of data. Large double-precision arrays consume a lot of memory. If the results you want do not require that you keep the original array or intermediate arrays of data, perform analysis operations in place where possible.
- The Interactive window maintains a record of generated code. If you forget to keep the code from a function panel, you can cut and paste code between the Interactive and Program windows.


## Reporting Analysis Errors

Each analysis function returns an integer error code. If the function executes properly, the function returns a zero; otherwise, the function returns an appropriate error value.

The return value corresponds to one of the enumeration values of the type that AnalysisLibErrType declares in the header file analysis.h. The analysis functions are declared in the header file with this return type so that the function panel controls for return values display the symbolic name instead of the integer value of the error code. Declaring a variable to be the type AnalysisLibErrType allows the Variables window to display its value as a symbolic name instead of as an integer.

You can find a list of error codes in Appendix A, Error Codes.

## About the Fast Fourier Transform (FFT)

The functions in the Frequency Domain subclass are based on the discrete implementation and optimization of the Fourier Transform integral. The functions obtain the Discrete Fourier Transform (DFT) of a complex sequence $X$ that contains $n$ elements using the following formula:

$$
Y_{i}=\sum_{k=0}^{n-1} X_{k} e^{\frac{-j 2 \pi i k}{n}} \text { for } i=0,1, \ldots, n-1
$$

where $Y_{i}$ is the $i^{\text {th }}$ element of the DFT of $X$ and $j=\sqrt{-1}$
The DFT of $X$ also results in a complex sequence $Y$ of $n$ elements. Similarly, the functions obtain the Inverse Discrete Fourier Transform (IDFT) of a complex sequence $Y$ that contains $n$ elements using the following formula:

$$
X_{i}=\frac{1}{n} \sum_{k=0}^{n-1} Y_{k} e^{\frac{j 2 \pi i k}{n}} \quad \text { for } i=0,1, \ldots, n-1
$$

where $X_{i}$ is the $i^{\text {th }}$ element of the IDFT of $Y$ and $j=\sqrt{-1}$
The discrete implementation of the DFT is a numerically intense process. However, it is possible to implement a fast algorithm when the size of the sequence is a power of two. These algorithms are known as FFTs and can be found in many introductory texts about digital signal processing (DSP).

The current algorithm implemented in the LabWindows/CVI Advanced Analysis Library is known as the Split-Radix algorithm. This algorithm is highly efficient because it minimizes the number of multiplications and has the form of the Radix-4 algorithm and the efficiency of the Radix-8 algorithm. The resulting complex FFT sequence has the conventional DSP format as described in the following paragraphs.

If there are $n$ number of elements in the complex sequence and $k=\frac{n}{2}$, the output of the FFT is organized as follows:

| $Y_{0}$ | DC component |
| :---: | :--- |
| $Y_{1}$ | Positive first harmonic |
| $Y_{2}$ | Positive second harmonic |
| $\cdot$ | $\cdot$ |
| $\cdot$ | $\cdot$ |
| $\cdot$ | $\cdot$ |
| $Y_{k-1}$ | Positive $k-1$ harmonic |
| $Y_{k}$ | Nyquist frequency |
| $Y_{k+1}$ | Negative $k-1$ harmonic |
| $\cdot$ | $\cdot$ |
| $\cdot$ | $\cdot$ |
| $\cdot$ | $\cdot$ |
| $Y_{n-2}$ | Negative second harmonic |
| $Y_{n-1}$ | Negative first harmonic |

The following conventions and restrictions apply to the functions in the Frequency
Domain subclass:

- All arrays must be a power of two: $n=2^{m}$, for $m=1,2,3, \ldots, 12$.
- The functions manipulate complex sequences using two arrays. One array contains the real elements. The other array contains the imaginary elements.

This manual uses the following notation to describe the FFT operations the functions in the Frequency Domain subclass perform:

- $\quad Y=\mathrm{FFT}(X)$, the sequence $Y$ is the FFT of the sequence $X$.
- $\quad Y=\mathrm{FFT}^{-1}(X)$, the sequence $Y$ is the inverse FFT of the sequence $X$.
$X$ is usually a complex array but can be treated as a real array.


## About Windowing

Almost every application requires you to use finite length signals. This requires that continuous signals be truncated, using a process called windowing.

The simplest window is a rectangular window. Because this window requires no special effort, it is commonly referred to as the no window option. Remember, however, that a window always affects a discrete signal and its spectrum. Let $x_{n}$ be a digitized time-domain waveform that has a finite length of $n . w_{n}$ is a window sequence of $n$ points. The windowed output is calculated as follows:

$$
\begin{equation*}
y_{i}=x_{i} \times w_{i} \tag{1-1}
\end{equation*}
$$

If $X, Y$, and $W$ are the spectra of $x, y$, and $w$, respectively, the time-domain multiplication in Equation (1-1) is equivalent to the frequency domain convolution shown as follows:

$$
Y_{k}=X_{k} \Theta W_{k}
$$

Convolving with the window spectrum always distorts the original signal spectrum in some way. A window spectrum consists of a mainlobe and several sidelobes.

The mainlobe is the primary cause of lost frequency resolution. When two signal spectrum lines are too close to each other, they might fall in the width of the mainlobe, causing the output of the windowed signal spectrum to have only one spectrum line. Use a window with a narrower mainlobe to reduce the loss of frequency resolution. A rectangular window has the narrowest mainlobe, so it provides the best frequency resolution.

The sidelobes of a window function affect frequency leakage. A signal spectrum line leaks into the adjacent spectrum if the sidelobes are large. Once again, the leakage results from the convolution process. Select a window with relatively smaller sidelobes to reduce spectral leakage. Unfortunately, a narrower mainlobe and smaller sidelobes are mutually exclusive.

For this reason, selecting a window function is application dependent. Figure 1-1 shows an example of a windowed spectrum in the continuous case.


Figure 1-1. Windowed Spectrum in the Continuous Case
The original signal spectrum in Figure 1-1 is convolved with the window spectrum, and the output is a smeared version of the original signal spectrum. In Figure 1-1, you can still see four distinctive peaks from the original signal; but each peak is smeared, and the frequency leakage effect is clear.

Window definitions used in National Instruments analysis libraries are designed in such a way that the window operations in the time domain are equivalent to the operations of the same window in the frequency domain. To meet this requirement, the windows are not symmetrical in the time domain, that is:

$$
w_{0} \neq w_{N-1} \quad \text { where } N \text { is the window length }
$$

However, the windows are usually symmetrical in the frequency domain. For example, the Hamming window definition uses the formula:

$$
\begin{equation*}
w_{i}=0.54-0.46 \cos \left(\frac{2 \pi i}{N}\right) \tag{1-2}
\end{equation*}
$$

Other manufacturers might use a slightly different definition, such as:

$$
\begin{equation*}
w_{i}=0.54-0.46 \cos \left(\frac{2 \pi i}{N-1}\right) \tag{1-3}
\end{equation*}
$$

The difference is small if $N$ is large.
Equation (1-2) is not symmetrical in the time domain, but it ensures that the time domain windowing is equivalent to the frequency domain windowing. If you want to have a perfectly symmetrical sequence in the time domain, you must write your own windowing function using Equation (1-3).

The choice of a window depends on the application. For most applications, the Hamming or Hanning windows deliver good performance.

## About Digital Filters

There are two types of digital filters in the LabWindows/CVI Advanced Analysis Library: Finite Impulse Response (FIR) filters and Infinite Impulse Response (IIR) filters. FIR filters have a linear phase response. IIR filters generally have a nonlinear phase response but offer much better amplitude response.

The choice of a particular type of filter depends on the application. If you want a linear phase response, choose one of the FIR filters. If performance and better amplitude response are more important, chose an IIR filter. No matter what type of filter you choose, enter a sampling frequency and other cutoff frequencies when you design your filter. You can design a digital filter using a normalized sampling frequency. The LabWindows/CVI Advanced Analysis Library provides a sampling frequency parameter so that you do not need to normalize other frequencies.

## FIR Filters

The FIR filter is a set of filter coefficients that alters the signal spectrum when convolving with the signal. Let $c_{k}$, for $k=0,1,2, \ldots$, be the filter coefficients, $x_{N}$ the input signal, and $y_{N}$ the output in the following formula:

$$
y_{i}=\sum_{k=0}^{K-1} x_{i-k} c_{k} \quad \text { when } i=0,1, \ldots, N-1
$$

LabWindows/CVI implements the formula using the convolution function Convolve. The purpose of an FIR filter is to design the coefficients $c_{k}$. Remember that an FIR filter function does not actually perform filtering. You must subsequently call Convolve to perform the filtering. The advantage of this process is that after you obtain the filter coefficients, you can use them repeatedly without redesigning the filter.

If you have never used an FIR filter before, start with a window FIR filter. These filters are easy to design, though other techniques might design a better filter with the same number of coefficients.

Use the windType parameter to choose the window type to use in a window FIR filter. windType determines the amount of attenuation the window filter can achieve. It also determines the transitional bandwidth of the window filter, which is the frequency range from the specified cutoff frequency to the point where the desired attenuation is obtained. A bigger transitional bandwidth usually gives better attenuation. Use a Kaiser window FIR filter to choose windows that are not available from windType.

If you are experienced in using filters and you want to design an optimal FIR filter, use the LabWindows/CVI Advanced Analysis Library Equi_Ripple function. These filters are based on the general Parks-McClellan algorithm, that, in turn, is based on an alternation theorem in the polynomial approximation. As the name suggests, the frequency response of an Equi_Ripple filter has equal ripples within each specified frequency band. The ripples can be different in different bands depending on the weighting factors.

You have to specify more parameters when you use Equi_Ripple filters. For each frequency band, specify the starting and ending points, the amplitude response, and a weighting factor associated with the amplitude response of that band. A weighting factor of one is usually sufficient for all bands, but you can select different weighting factors. A bigger weighting factor results in a smaller ripple in the corresponding frequency band; a smaller weighting factor results in a larger ripple.

If you want to design an optimal FIR multiband filter, such as lowpass, highpass, bandpass, and bandstop, but do not want to specify the weighting factor, use EquiRpl_LPF, EquiRpl_HPF, EquiRpl_BPF, and EquiRpl_BSF. These filters call Equi_Ripple internally but have simplified input parameters. fail and give erroneous results. It is extremely important that you verify the filter design after you obtain the filter coefficients.

## IIR Filters

Mathematically, an IIR digital filter assumes the following form:

$$
\begin{equation*}
y_{i}=\frac{1}{a_{0}}\left(\sum_{j=0}^{N_{b}-1} b_{j} x_{i-j}-\sum_{k=1}^{N_{a}-1} a_{k} y_{i-k}\right) \quad \text { where } a_{k} \text { and } b_{k} \text { are the filter coefficients } \tag{1-4}
\end{equation*}
$$

The current filter output $y_{i}$ depends on the current and previous values $x_{i-k}$ and previous output $y_{i-k}$. If $y_{i} \neq 0$, its effect on the subsequent points persists indefinitely. For these reasons, these filters are called infinite impulse response filters.

Filters implemented directly using the structure Equation (1-4) defines are known as direct-form IIR filters. Direct-form implementations are often sensitive to errors introduced by coefficient quantization and by computational precision limits. Also, a filter designed to be stable can become unstable with increasing coefficient length, which is proportional to filter order.

A less-sensitive structure can be obtained by breaking up the direct-form transfer function into lower-order sections, or filter stages. The direct-form transfer function of the filter given by Equation (1-4) (with $a_{0}=1$ ) can be written as a ratio of $z$ transforms, as follows:

$$
\begin{equation*}
H(z)=\frac{b_{0}+b_{1} z^{-1}+\ldots+b_{N_{b}-1} z^{-\left(N_{b}-1\right)}}{1+a_{1} z^{-1}+\ldots+a_{N_{a}-1} z^{-\left(N_{a}-1\right)}} \tag{1-5}
\end{equation*}
$$

By factoring Equation (1-5) into second-order sections, the transfer function of the filter becomes a product of second-order filter functions:

$$
H(z)=\prod_{k=1}^{N_{s}} \frac{b_{0_{k}}+b_{1_{k}} z^{-1}+b_{2_{k}} z^{-2}}{1+a_{1_{k}} z^{-1}+a_{2_{k}} z^{-2}}
$$

where $N_{s}=\left\lfloor\frac{N_{a}}{2}\right\rfloor$ is the largest integer $\leq \frac{N_{a}}{2}$, and $N_{a} \geq N_{b}$

This new filter structure can be described as a cascade of second-order filters, as shown in Figure 1-2.


Figure 1-2. Cascaded Filter Stages
Each individual stage is implemented using the direct-form II filter structure because it requires a minimum number of arithmetic operations and a minimum number of delay elements, or internal filter states. Each stage has one input, one output, and two past internal states $\left(s_{k}[i-1]\right.$ and $\left.s_{k}[i-2]\right)$.

If $n$ is the number of samples in the input sequence, the filtering operation proceeds as in the following equations:

$$
\begin{array}{cc}
y_{0}[i]=x[i] \\
s_{k}[i]=y_{k-1}[i-1]-a_{1_{k}} s_{k}[i-1]-a_{2_{k}} s_{k}[i-2] & \text { for } k=1,2, \ldots, N_{s} \\
y_{k}[i]=b_{0_{k}} s_{k}[i]+b_{1_{k}} s_{k}[i-1]+b_{2_{k}} s_{k}[i-2] & \text { for } k=1,2, \ldots, N_{s} \\
y[i]=y_{N_{s}}[i] &
\end{array}
$$

for each sample $i=0,1,2, \ldots, n-1$
For lowpass and highpass filters with a single cutoff frequency, second-order filter stages can be designed directly. The overall IIR lowpass or highpass filter contains cascaded second-order filters.

For bandpass and bandstop filters with two cutoff frequencies, fourth-order filter stages are a more natural form. The overall IIR bandpass or bandstop filter contains cascaded fourth-order filters. The filtering operation for fourth-order stages proceeds as in the following equations:

$$
\begin{gathered}
y_{0}[i]=x[i] \\
s_{k}[i]=y_{k-1}[i-1]-a_{1_{k}} s_{k}[i-1]-a_{2_{k}} s_{k}[i-2]-a_{3_{k}} s_{k}[i-3]-a_{4_{k}} s_{k}[i-4]
\end{gathered}
$$

for $k=1,2, \ldots, N_{s}$

$$
y_{k}[i]=b_{0_{k}} s_{k}[i]+b_{1_{k}} s_{k}[i-1]+b_{2_{k}} s_{k}[i-2]+b_{3_{k}} s_{k}[i-3]+b_{4_{k}} s_{k}[i-4]
$$

for $k=1,2, \ldots, N_{s}$

$$
y[i]=y_{N_{s}}[i]
$$

Notice that in the case of fourth-order filter stages, $N_{s}=\left\lfloor\frac{N_{a}+1}{4}\right\rfloor$.
The IIR filters provided in the LabWindows/CVI Advanced Analysis Library are derived from analog filters. There are four major types of IIR filters:

- Butterworth filters
- Chebyshev filters
- Inverse Chebyshev filters
- Elliptic filters

Lowpass, highpass, bandpass, and bandstop filters exist for each type of filter. The frequency response of a Butterworth filter is characterized by a smooth response at all frequencies and a monotonic decrease from the specified cut-off frequencies. Butterworth filters are maximally flat in the passband and zero in the stopband. The rolloff between the passband and stopband is slow, so a lower-order Butterworth filter does not provide a good approximation of an ideal filter.

Chebyshev filters have equal ripples in the passband and a monotonically decreasing magnitude response in the stopband. These filters have much sharper rolloffs than Butterworth filters. The inverse Chebyshev filters are similar to Chebyshev filters except that the ripple occurs in the stopband and the frequency response is flat in the passband.

If ripples are allowable in both the passband and the stopband, use elliptic filters. Elliptic filters have the sharpest rolloffs for the same order compared with Butterworth or Chebyshev filters.

## About Measurement Functions

Measurement functions perform DFT-based and FFT-based analysis with signal acquisition for frequency measurement applications as seen in typical frequency measurement instruments such as dynamic signal analyzers.

Several measurement functions perform commonly used time domain-to-frequency domain transformations such as amplitude and phase spectrum, signal power spectrum, network transfer function, and so on. Other supportive measurement functions perform scaled time-domain windowing and power and frequency estimation, and total harmonic distortion analysis.

You can use the measurement functions for the following applications:

- Spectrum analysis applications
- Amplitude and phase spectrum
- Power spectrum
- Scaled time-domain window
- Power and frequency estimate
- Network, or frequency response, and dual-channel analysis applications
- Transfer function
- Impulse response function
- Network functions, including coherence
- Cross power spectrum

The DFT, FFT, and power spectrum functions are useful for measuring the frequency content of stationary or transient signals. The FFT provides the average frequency content of the signal over the entire time that the signal was acquired. For this reason, you use the FFT mostly for stationary signal analysis, when the signal is not significantly changing in frequency content over the time that the signal is acquired, or when you want only the average energy at each frequency line. A large class of measurement problems falls in this category. For measuring frequency information that changes during the acquisition, use joint time-frequency analysis.

The measurement functions are built on top of the signal processing functions and have the following characteristics that model the behavior of traditional benchtop frequency analysis instruments:

- Assumed real-world, time-domain signal input.
- Outputs in magnitude and phase, scaled in units where appropriate, ready for immediate graphing.
- Single-sided spectrums from DC to $\frac{f s}{2}$, where $f s$ is the sampling frequency.
- Sampling period-to-frequency interval conversion for graphing with appropriate x -axis units, in hertz.
- Corrections for the windows being used applied where appropriate.
- Scaled windows; each window gives same peak spectrum amplitude result within its amplitude accuracy constraints.
- Viewing of power or amplitude spectrum in various unit formats, including decibels and spectral density units, ( $\left.\mathrm{V}^{2} / \mathrm{Hz}, \mathrm{V} / \sqrt{\mathrm{Hz}}\right)$, and so on.


## About Curve Fitting Functions

The algorithm used to find the best curve fit in the Curve Fitting class is the Least Squares method. The purpose of the algorithm is to find the curve coefficients $a$, which minimize the squared error $e(a)$ in the following formula:

$$
e(a)=\sum_{i}\left|Y_{i}-\mathrm{f}\left(X_{i}, a\right)\right|^{2} \quad \text { where } \mathrm{f}\left(X_{i}, a\right) \text { is the function that represents the desired curve }
$$

You can find the coefficient $a$ by solving the linear system of equations the following formula generates:

$$
\frac{\partial}{\partial a} e(a)=0
$$

Given a set of $n$ sample points $(x, y)$ represented by the sequences $X$ and $Y$, the curve-fitting functions determine the coefficients that best represent the data. The best fit $Z$ is an array of expected values given the coefficients and the $X$ set of values. Thus, you can express $Z$ as a function of $X$ and the following coefficients:

$$
Z=\mathrm{f}(X, a)
$$

When you establish the best fit values, you can obtain the mean squared error (mse) by applying the following formula:

$$
m s e=\sum_{i=0}^{n-1} \frac{\left(Z_{i}-Y_{i}\right)^{2}}{n}
$$

## About Vector \& Matrix Algebra Functions

The functions in the Vector \& Matrix Algebra class perform operations such as multiplication, transposition, and outer product calculation on 2D arrays or matrices. You can use these functions to calculate matrix properties, such as determinant, rank, norm, and condition number.

Many applications require you to solve a linear system of equations and/or to determine the eigenvalues and eigenvectors of a matrix. This class contains functions you can use for this purpose and functions that you can use to calculate different types of factorizations, such as Cholesky factorization, QR factorization, and Singular Value Decomposition. You can use the functions in this class to calculate special types of matrices, such as Toeplitz matrix, Vandermonde matrix, and Companion matrix.

## Advanced Analysis Library Function Reference

This chapter contains a brief explanation of each of the functions in the LabWindows/CVI Advanced Analysis Library in alphabetical order.

## Abs1D

```
int status = Abs1D (double x[], int n, double y[]);
```


## Purpose

Finds the absolute value of the $\mathbf{x}$ input array. Abs1D can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Absolute value of input array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## ACDCEstimator

```
int status = ACDCEstimator (double x[], int n, double *acEstimate,
    double *dcEstimate);
```


## Purpose

Calculates an estimation of the AC and DC contents of the input signal. $\mathbf{x}$ is the input signal, usually in volts.
acEstimate is the estimate of the input signal AC content in volts, root-mean-square, if the input signal is in volts.
dcEstimate is the estimate of the input signal DC content in volts, if the input signal is in volts.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Contains the time-domain signal, usually in <br> volts. This array must contain at least three <br> cycles of the signal for a valid estimate. |
| $\mathbf{n}$ | integer | Number of elements in the input array. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| acEstimate | double-precision | Contains the estimate of the AC level of the <br> input signal in volts, root-mean-square, if <br> the input signal is volts. |
| dcEstimate | double-precision | Contains the estimate of the DC level of the <br> input signal in the same units as the <br> input signal. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Add1D

```
int status = Add1D (double x[], double y[], int n, double z[]);
```


## Purpose

Adds 1D arrays. Add1D obtains the $i^{\text {th }}$ element of the output array using the following formula:

$$
z_{i}=x_{i}+y_{i}
$$

Add1D can perform the operation in place; that is, $\mathbf{z}$ can be the same array as either $\mathbf{x}$ or $\mathbf{y}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{y}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements to add. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z}$ | double-precision array | Result array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Add2D

```
int status = Add2D (void *x, void *y, int n, int m, void *z);
```


## Purpose

Adds 2D arrays. Add2D obtains the $(i, j)^{t h}$ element of the output array using the following formula:

$$
z_{i, j}=x_{i, j}+y_{i, j}
$$

Add2D can perform the operation in place; that is, $\mathbf{z}$ can be the same array as either $\mathbf{x}$ or $\mathbf{y}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision <br> 2D array | Input array. |
| $\mathbf{y}$ | double-precision <br> 2D array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in first dimension. |
| $\mathbf{m}$ | integer | Number of elements in second dimension. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z}$ | double-precision <br> 2D array | Result array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## AllocIIRFilterPtr

```
IIRFilterPtr filterInformation = AllocIIRFilterPtr (int type, int order);
```


## Purpose

Allocates and initializes the filterInformation structure. Returns a pointer to the filter structure for use with the IIR cascade filter coefficient design calls.

You input the type of the filter, such as lowpass, highpass, bandpass, or bandstop, and the order. AllocIIRFilterPtr allocates the filter structure as well as the internal coefficient arrays and internal filter state array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| type | integer | Controls the filter type of IIR filter <br> coefficients. <br> lowpass $=0$ (default) <br> highpass $=1$ <br> bandpass = 2 <br> bandstop $=3$ |
| order | integer | Specifies the order of the IIR filter. <br> The default value is 3. |

## Return Value

| Name | Type | Description |
| :---: | :--- | :--- |
| filterInformation | IIRFilterPtr | Pointer to the filter structure. When an error <br> occurs, filterInformation is zero. |

## Parameter Discussion

filterInformation is the pointer to the filter structure that contains the filter coefficients and the internal filter information. Call this function to allocate filterInformation before you call one of the cascade IIR filter design functions.

The definition of the filter structure is as follows:

```
typedef struct {
    intnum type; /* type of filter (lp, hp, bp, bs) */
    intnum order /* order of filter */
    intnum reset; /* 0 - don't reset, 1 - reset */
    intnum na;
    floatnum *a;
    intnum nb;
    floatnum *b;
    intnum ns;
    floatnum *s;
    } *IIRFilterPtr;
```


## AmpPhaseSpectrum

```
int status = AmpPhaseSpectrum (double x[], int n, int unwrap, double dt,
    double ampSpectrum[], double phaseSpectrum[],
    double *df);
```


## Purpose

Calculates the single-sided, scaled amplitude and phase spectra of a time-domain signal, $X$. AmpPhaseSpect rum calculates the amplitude spectrum as

$$
\left|\frac{\mathrm{FFT}(X)}{n}\right|
$$

and converts it to single-sided form. AmpPhaseSpect rum calculates the phase spectrum as

$$
\operatorname{phase}(\operatorname{FFT}(X))
$$

and converts it to single-sided form.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Contains the time-domain signal. |
| $\mathbf{n}$ | integer | Number of elements in the input array. <br> $\mathbf{n}$ must be a power of 2. |
| unwrap | integer | Controls the unwrapping of the phase <br> spectrum. <br> Valid values: <br> $1=$ enable phase unwrapping <br> $0=$ disable phase unwrapping <br> $(-\pi \leq p h a s e \leq+\pi)$ |
| $\mathbf{d t}$ | double-precision | Sampling period of the time-domain signal, <br> usually in seconds. <br> dt $=1 / f s$, where $f s$ is the sampling <br> frequency of the time-domain signal. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| ampSpectrum | double-precision array | Single-sided amplitude spectrum <br> magnitude in volts, root-mean-square, if the <br> input signal is in volts. If the input signal is <br> not in volts, the results are in input signal <br> units, root-mean-square. This array must be <br> at least $\mathbf{n} / 2$ elements long. |
| phaseSpectrum | double-precision array | Single-sided phase spectrum in radians. <br> This array must be at least $\mathbf{n} / 2$ <br> elements long. |
| df | double-precision | Points to the frequency interval, in hertz, if <br> $\mathbf{d t}$ is in seconds. <br> $\mathbf{d f}=1 /(\mathbf{n} \times \mathbf{d t})$ |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## ANOVA1Way

```
int status = ANOVA1Way (double y[], int level[], int n, int k, double *ssa,
    double *msa, double *f, double *sig, double *sse,
    double *mse, double *tss);
```


## Purpose

Takes an array of experimental observations you make at various levels of some factor, with at least one observation per factor, and performs a one-way analysis of variance (ANOVA) in the fixed effect model.

The one-way analysis of variance is a test to determine whether the level of the factor has an effect on the experimental outcome.

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Experimental observations. |
| level | integer array | The $i^{\text {th }}$ element tells in what level of the <br> factor the $i^{t h}$ observation falls. |
| $\mathbf{n}$ | integer | Total number of observations. |
| $\mathbf{k}$ | integer | Total number of levels of the factor. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| ssa | double-precision | Sum of squares as a result of the factor. |
| msa | double-precision | Mean square as a result of the factor. |
| $\mathbf{f}$ | double-precision | Calculated F-value. |
| sig | double-precision | Level of significance at which you must <br> reject the null hypothesis. |
| sse | double-precision | Sum of squares as a result of random <br> fluctuation. |
| mse | double-precision | Mean square as a result of random <br> fluctuation. |
| tss | Total sum of squares. |  |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Using This Function

## Factors and Levels

A factor is a way of categorizing data. You can categorize data into levels, beginning with level 0 . For example, if you perform a measurement on individuals, such as counting the number of sit-ups they can perform, one such categorization method is age. For age, you might have three levels, as shown in Table 2-1.

Table 2-1. Age Levels

| Level | Ages |
| :---: | :---: |
| 0 | 6 years to 10 years |
| 1 | 11 years to 15 years |
| 2 | 16 years to 20 years |

## General Method

Break up the total sum of squares tss, a measure of the total variation of the data from the overall population mean, into component sums of squares, which might be attributed to different sources.

You now have:

$$
t s s=s s a+s s e
$$

where $s s a$ is a measure of variation that is attributed to the factor sse is a measure of variation that is attributed to random fluctuation

Divide by appropriate numbers to obtain the averages msa and mse. If the factor causes much variation, msa will be larger relative to mse. The ratio $\mathbf{f}$ also will be larger relative to mse.

If the null hypothesis is true, the ratio $\mathbf{f}$ is taken from an F-distribution with $\mathbf{k}-1$ and $\mathbf{n}-\mathbf{k}$ degrees of freedom, from which you can calculate probabilities. Given a particular $\mathbf{f}, \mathbf{s i g}$ is the probability that sampling from this distribution results in a value larger than $f$.

## Statistical Model

ANOVA1Way expresses each experimental outcome as the sum of three parts while it performs the analysis of variance. Let $y_{i, m}$ be the $m^{\text {th }}$ observation from the $i^{\text {th }}$ level. Each observation is written as:

$$
y_{i, m}=\mu+\alpha_{i}+\varepsilon_{i, m}
$$

where $\mu$ is a standard effect
$\alpha_{i}$ is the effect of the $i^{\text {th }}$ level of the factor
$\varepsilon_{i, m}$ is a random fluctuation

## Assumptions

Assume that the populations of measurements at each level are normally distributed with mean $\alpha_{i}$ and variance $\sigma_{A}^{2}$. Assume that the means $\alpha_{i}$ sum to zero. Finally, assume that for each $i$ and $m, \varepsilon_{i, m}$ is normally distributed with mean 0 and variance $\sigma_{A}^{2}$.

## Hypothesis

Test the null hypothesis that $\alpha_{i}=0$ for $i=0,1, \ldots, k-1$, where $\mathbf{k}$ is the total number of levels. In other words, assume from the start that the levels have no effect on the experimental outcome, then look for evidence to the contrary.

## Testing the Hypothesis

ANOVA1Way generates a number $\mathbf{f}$ so that if the hypothesis is true, that number is from an F-distribution with $\mathbf{k}-1$ and $\mathbf{n}-\mathbf{k}$ degrees of freedom. ANOVA1Way also calculates the probability that a number taken from this F-distribution is larger than $\mathbf{f}$. This is the output parameter sig:

$$
\operatorname{sig}=\operatorname{prob}(x>f) \quad \text { where } x \text { is from } \mathrm{F}(k-1, n-k)
$$

Use the probability sig to determine when to reject the hypothesis by choosing a level of significance for the hypothesis. The level of significance determines how likely you are to reject the hypothesis when it is in fact true. Thus, the level of significance should be small, for example, 0.05 . Remember that the smaller the level of significance, the less likely you are to reject the hypothesis.

Reject the hypothesis when the output parameter sig is less than the level of significance you choose.

## Formulas

Let $y_{i, m}$ be the $m^{\text {th }}$ observation at the $i^{\text {th }}$ level for $m=0,1, \ldots, n_{i}$ and $i=0,1, \ldots, k$.
Let $n_{i}=$ the number of observations at the $i^{\text {th }}$ level.

$$
\begin{gathered}
Y_{i}=\frac{1}{n_{i}} \sum_{m=0}^{n_{i}-1} y_{i, m} \\
Y_{m}=\frac{1}{k} \sum_{i=0}^{k-1} y_{i, m} \\
Y=\frac{1}{n} \sum_{i=0}^{k-1} \sum_{m=0}^{n_{i}-1} y_{i, m} \\
T=n \times Y
\end{gathered}
$$

Then:

$$
\begin{gathered}
s s a=\sum_{i=0}^{k} \frac{Y_{i}^{2}}{n_{i}}-\frac{Y^{2}}{n} \\
m s e=\frac{s s a}{k-1} \\
s s e=\sum_{i=0}^{k-1} \sum_{m=0}^{n_{i}-1} y_{i, m}^{2}-\sum_{i=0}^{k} \frac{Y_{i}^{2}}{n_{i}} \\
m s e=\frac{s s e}{n-k} \\
t s s=\sum_{i=0}^{k-1} \sum_{m=0}^{n_{i}-1} y_{i, m}^{2}-\frac{Y^{2}}{n}
\end{gathered}
$$

$f=\frac{m s a}{m s e} \quad$ where $f$ is from an F-distribution with $k-1$ and $n-k$ degrees of freedom

## Example

Suppose that researchers want to know whether the amount of rainfall affects the yield of a crop. The factor, rainfall, is divided into three levels $(\mathbf{k}=3)$ as shown in Table 2-2.

Table 2-2. Rainfall Levels

| Level | Rainfall (Factor) |
| :---: | :--- |
| 0 | 2 inches |
| 1 | 3 inches |
| 2 | 4 inches |

The researchers set up 10 plots in various geographical locations chosen so that each plot receives a different amount of rainfall. Table 2-3 shows their results.

Table 2-3. Plot Production

| Level | Bushels Produced <br> from Each Plot |  |  |  |
| :---: | :--- | :--- | :--- | :--- |
| 0 | 128 | 122 | 126 | 124 |
| 1 | 140 | 141 | 143 |  |
| 2 | 120 | 118 | 123 |  |

To perform a one-way analysis using ANOVA1Way, you store all the numbers of bushels in a double-precision array $\mathbf{y}$ of size 10. The integer array level records the levels in which observations were made. For any particular $i$, you must set these arrays such that $y_{i}$ is the number of bushels a plot produces in the $i^{\text {th }}$ level. For example:

$$
\begin{gathered}
\text { level }_{i}=0 \\
y_{i}=128,122,126, \text { or } 124
\end{gathered}
$$

are valid combinations. Therefore, you can set up the input arrays $\mathbf{y}$ and level in this example for ANOVA1Way as follows:

$$
\begin{gathered}
y=128,122,126,124,140,141,143,120,118,123 \\
\text { level }=0,0,0,0,1,1,1,2,2,2
\end{gathered}
$$

Running the code in the following example produces:

$$
\operatorname{sig}=0.0000239
$$

For a level of significance such as 0.05 , the ANOVA1Way results show that the researchers must reject the hypothesis that the rainfall has no effect on the crop yield. In other words, the rainfall does affect the crop yield.

## Example

```
double y[10], ssa, msa, f, sig, sse, mse, tss;
int level[10];
int k;
int status;
k = 3; /* three levels for rainfall */
/* Read in recorded data y(10), level[10]. */
status = ANOVA1Way(y, level, 10, k, &ssa, &msa, &f, &sig, &sse, &mse,
        &tss);
```


## ANOVA2Way

```
int status = ANOVA2Way (double y[], int levelA[], int levelB[], int N,
    int L, int a, int b, void *info, double *sigA,
    double *sigB, double *sigAB);
```


## Purpose

Takes an array of experimental observations made at various levels of two factors and performs a two-way analysis of variance in any of the following models:

- Model 1: Fixed effects with no interaction and one observation per cell. $\mathbf{L}=1$ per specified levels $\mathbf{a}$ and $\mathbf{b}$ of the factors A and B, respectively.
- Model 2: Fixed effects with interaction and $\mathbf{L}>1$ observations per cell.
- Model 3: Either of the mixed-effects models, where one factor is taken to have a fixed effect but the other is taken to have a random effect, with interaction and $\mathbf{L}>1$ observations per cell.
- Model 4: Random effects with interaction and $\mathbf{L}>1$ observations per cell.

Any ANOVA looks for evidence that the factors, or interactions among the factors, have a significant effect on experimental outcomes. The method for finding significance varies among models.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Array of experimental data of <br> $\mathbf{N}=\|\mathbf{a}\| \times\|\mathbf{b}\| \times \mathbf{L}$ elements. |
| levelA | integer array | The $i^{\text {th }}$ element tells in what level of <br> factor A the $i^{\text {th }}$ observation falls. |
| levelB | integer array | The $i^{\text {th }}$ element tells in what level of <br> factor B the $i^{\text {th }}$ observation falls. |
| $\mathbf{N}$ | integer | Total number of observations. |
| $\mathbf{L}$ | integer | Number of observations per cell. |
| a | integer | Number of levels in factor A; negative if <br> A is a random effect. |
| $\mathbf{b}$ | integer | Number of levels in factor B; negative if <br> B is a random effect. |

## Output

| Name | Type | Description |
| :---: | :---: | :---: |
| info | double-precision 2D array | A 4-by-4 matrix as follows: $\left[\begin{array}{cccc} \text { ssa } & \text { dofa } & \text { msa } & f a \\ \text { ssb } & \text { dofb } & \text { msb } & f b \\ \text { ssab } & \text { dofab } & \text { msab } & \text { fab } \\ \text { sse } & \text { dofe } & \text { mse } & 0.0 \end{array}\right]$ <br> where ss designates sums of squares, $d o f$ designates degrees of freedom of $s s$, $m s$ designates mean squares, and $f$ designates F -distributions, depending on the statistical model. |
| sigA | double-precision | Level of significance at which you must reject hypothesis A. |
| sigB | double-precision | Level of significance at which you must reject hypothesis B. |
| sigAB | double-precision | Level of significance at which you must reject hypothesis AB. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Using This Function

## Factors, Levels, and Cells

A factor is a way of categorizing data. You can categorize data into levels, beginning with level 0 . For example, if you perform a measurement on individuals, such as counting the number of sit-ups they can perform, one such categorization method is age. For age, you might have three levels, as shown in Table 2-4.

Table 2-4. Age Levels

| Level | Age |
| :---: | :---: |
| 0 | 6 years to 10 years |
| 1 | 11 years to 15 years |
| 2 | 16 years to 20 years |

Another possible factor is eye color, with the levels shown in Table 2-5.
Table 2-5. Eye Color Levels

| Level | Eye Color |
| :---: | :--- |
| 0 | blue |
| 1 | brown |
| 2 | green |
| 3 | hazel |

In this example, an analysis of variance seeks evidence that the ages and eye color of the subjects have an effect on the number of sit-ups they perform.

A cell of data consists of all those experimental observations that fall in particular levels of the two factors. In this instance, a cell might consist of those observations made on hazel-eyed individuals between 11 years and 15 years old. The number of observations that fall in each cell must be a constant number $\mathbf{L}$ that does not vary between cells.

## Random and Fixed Effects

A factor is taken as a random effect when the factor has a large population of levels you want to draw conclusions about, but that you cannot sample at all levels. Levels are sampled at random in the hope of generalizing about all levels.

A factor is taken as a fixed effect when you can sample the factor from all levels you want to draw conclusions about.

The input parameters $\mathbf{a}$ and $\mathbf{b}$ represent the number of levels in factors A and B, respectively. If factor $A$ is random, set a to a negative value. If factor $B$ is random, set $\mathbf{b}$ to a negative value. If only one observation per cell exists, both $\mathbf{a}$ and $\mathbf{b}$ must be positive. Use model 1 as previously described.

## General Method

Each of the previous models breaks up the total sum of squares ( $t s s$ ), which is a measure of the total variation of the data from the overall population mean, into a number of component sums of squares. In model 1:

$$
t s s=s s a+s s b+s s e
$$

whereas in models 2 through 4 :

$$
t s s=s s a+s s b+s s a b+s s e
$$

Each component of the sums is a measure of variation attributed to a certain factor or interaction among the factors. The component $s s a$ is a measure of the variation as a result of factor A ; $s s b$ is a measure of the variation as a result of factor B ; $s s a b$ is a measure of the variation as a result of the interaction between factors A and B; and sse is a measure of the variation as a result of random fluctuation. Notice that there is no $s s a b$ term with model 1. Thus, no interaction exists.

If factor A has a strong effect on the experimental observations, msa is relatively large. You can look at specific ratios of these averages because you know how they are statistically distributed. You can therefore determine how likely it is that factor A is as relatively large as it is.

## Statistical Model

Let $y_{p, q, r}$ be the $r^{t h}$ observation at the $p^{t h}$ and $q^{t h}$ levels of A and B, respectively, where $r=0,1, \ldots, \mathbf{L}-1$.

In model 1, express each observation as the sum of four components so that:

$$
y_{p, q, r}=\mu+\alpha_{p}+\beta_{q}+\varepsilon_{p, q, r}
$$

where $\mu$ represents a standard effect present in each observation
$\alpha_{p}$ represents the effect of the $p^{t h}$ level of factor A
$\beta_{q}$ represents the effect of the $q^{\text {th }}$ level of factor B
$\varepsilon_{p, q, r}$ is a random fluctuation

In models 2 through 4, express each observation as the sum of five components so that:

$$
y_{p, q, r}=\mu+\alpha_{p}+\beta_{q}+(\alpha \beta)_{p, q}+\varepsilon_{p, q, r}
$$

where $\mu$ represents a standard effect present in each observation
$\alpha_{p}$ represents the effect of the $p^{t h}$ level of factor A
$\beta_{q}$ represents the effect of the $q^{\text {th }}$ level of factor B
$\varepsilon_{p, q, r}$ is a random fluctuation
$(\alpha \beta)_{p, q}$ represents the effect of the interaction between the $p^{\text {th }}$ level of factor A and the $q^{\text {th }}$ level of factor B

## Assumptions

- Assume that for each $p, q$, and $r, \varepsilon_{p, q, r}$ is normally distributed with mean 0 and variance $\sigma_{e}^{2}$.
- If a factor such as A is fixed, assume that the populations of measurements at each level are normally distributed with mean $\alpha_{p}$ and variance $\sigma_{A}^{2}$. All the populations at each of the levels have the same variance. In addition, assume that all the $\alpha_{p}$ means sum to zero. Make an analogous assumption for B.
- If a factor such as A is random, assume that the effect of the level of A itself, $\alpha_{p}$, is a random variable normally distributed with mean 0 and variance $\sigma_{A}^{2}$. Make an analogous assumption for B.
- If all the factors, such as A and B, associated with the effect of an interaction $(\alpha \beta)_{p, q}$ are fixed, assume that the populations of measurements at each level are normally distributed with mean $(\alpha \beta)_{p, q}$ and variance $\sigma_{A B}^{2}$. For any fixed $p$, the $(\alpha \beta)_{p, q}$ means sum to zero when summing over all $q$. Similarly, for any fixed $q$ the $(\alpha \beta)_{p, q}$ means sum to zero when summing over all $p$.
- If any of the factors, such as A and B, associated with the effect of an interaction $(\alpha \beta)_{p, q}$ are random, assume that the effect is a random variable normally distributed with mean 0 and variance $\sigma_{A B}^{2}$. If A is fixed but B is random, assume also that for any fixed $q$, the $(\alpha \beta)_{p, q}$ means sum to zero when summing over all $p$. Similarly, if B is fixed but A is random, assume also that for any fixed $p$, the $(\alpha \beta)_{p, q}$ means sum to zero when summing over all $q$.
- Assume that all effects taken to random variables are independent.


## Hypotheses

Each of the following hypotheses are different ways of stating that a factor or an interaction among factors has no effect on experimental outcomes. Start by assuming that there are no effects and then seek evidence to contradict these assumptions. The three hypotheses are as follows:

- For hypothesis A, $a_{p}=0$ for all levels of $p$ if factor A is fixed; $\sigma_{A}^{2}=0$ if factor A is random.
- For hypothesis $\mathrm{B}, \beta_{q}=0$ for all levels of $q$ if factor B is fixed; $\sigma_{B}^{2}=0$ if factor B is random.
- For hypothesis $\mathrm{AB},(\alpha \beta)_{p, q}=0$ for all levels of $p$ and $q$ if factors A and B are fixed; $\sigma_{A B}^{2}=0$ if either factor A or factor B is random. This does not apply to model 1 , where no interaction exists and the associated output parameters are superfluous.


## Testing the Hypotheses

For each hypothesis, ANOVA2Way generates a number so that if the hypothesis is true, that number is from a particular F-distribution.

For example, in model 1, $f a=m s a / m s e$, associated with hypothesis A, is from an F-distribution with $\mathbf{a}-1$ and $(\mathbf{a}-1) \times(\mathbf{b}-1)$ degrees of freedom, given that hypothesis A is true. In models 2 through $4, f a=m s a / m s e$, associated with hypothesis A , is from an F-distribution with $\mathbf{a}-1$ and $a b(\mathbf{L}-1)$ degrees of freedom, given that hypothesis A is true. ANOVA2Way calculates the probability that a number taken from a particular F-distribution is larger than the F-value. For example:

$$
\operatorname{sig} A=\operatorname{prob}(X>f a) \quad \text { where } X \text { is from } \mathrm{F}(a-1,(a-1)(b-1))
$$

Use the probabilities $\operatorname{sig} A, \operatorname{sigB}$, and $\operatorname{sigAB}$ to determine when to reject the associated hypotheses A, B, and AB by choosing a level of significance for each hypothesis. The level of significance determines how likely you are to reject the hypothesis when it is in fact true. Thus, the level of significance should be small, for example, 0.05 . Remember that the smaller the level of significance, the less likely you are to reject the hypothesis.

Reject a particular hypothesis when the associated output parameter $\operatorname{sig} \mathbf{A}, \operatorname{sig} \mathbf{B}$, or $\operatorname{sig} \mathbf{A B}$ is less than the level of significance you chose for that hypothesis. If A is a random effect, the chosen level of significance is 0.05 , and $\operatorname{sig} \mathbf{A}=0.03$, you must reject the hypothesis that $\sigma_{A}^{2}=0$ and conclude that factor $A$ has an effect on the experimental observations.

## Formulas

Let $y_{p, q, r}$ be the $r^{\text {th }}$ observation at the $p^{\text {th }}$ and $q^{\text {th }}$ levels of A and B, respectively, where $r=0,1, \ldots, \mathbf{L}-1$.

Let:

$$
\begin{aligned}
a a & =|a| \\
b b & =|b|
\end{aligned}
$$

$$
\begin{aligned}
T_{p, q} & =\sum_{r=0}^{L-1} y_{p, q, r} \\
T_{p} & =\sum_{q=0}^{b b-1} T_{p, q} \\
T_{q} & =\sum_{p=0}^{a a-1} T_{p, q}
\end{aligned}
$$

$T=$ the total sum of all observations:

$$
\begin{gathered}
A=\sum_{p=0}^{a a-1} \frac{T_{p}^{2}}{b b \times L} \\
B=\sum_{q=0}^{b b-1} \frac{T_{q}^{2}}{a a \times L} \\
S=\sum_{p=0}^{a a-1} \sum_{q=0}^{b b-1} \frac{T_{p, q^{2}}}{L} \\
C F=\frac{T^{2}}{a a \times b b \times L}
\end{gathered}
$$

Then:

$$
\begin{gathered}
s s a=A-C F \\
m s a=\frac{s s a}{a a-1}=\frac{s s a}{d o f a} \\
s s b=B-C F \\
m s b=\frac{s s b}{b b-1}=\frac{s s b}{d o f b} \\
s s a b=S-A-B-C F
\end{gathered}
$$

$$
\begin{aligned}
& m s a b=\frac{s s a b}{(a-1)(b-1)}=\frac{s s a b}{d o f a b} \\
& \text { sse }=T-S \\
& m s e=\frac{s s e}{a a \times b b \times(L-1)}=\frac{s s e}{d o f e} \\
& f a= \begin{cases}\frac{m s a}{m s e} & \text { if } B \text { is fixed } \\
\frac{m s a}{m s a b} & \text { if } B \text { is random }\end{cases} \\
& f b= \begin{cases}\frac{m s b}{m s e} & \text { if } A \text { is fixed } \\
\frac{m s b}{m s a b} & \text { if } A \text { is random }\end{cases} \\
& f a b=\frac{m s a b}{m s e}
\end{aligned}
$$

If:

$$
\begin{gathered}
f=\frac{m s_{1}}{m s_{2}} \\
m s_{1}=\frac{s s_{1}}{d o f_{1}} \\
m s_{2}=\frac{s s_{2}}{d o f_{2}}
\end{gathered}
$$

assume that $f$ is from an F-distribution with $d o f_{1}$ and $d o f_{2}$ degrees of freedom.

## Example

Suppose that researchers want to know how the amount of rainfall and the average temperature affect the yield of a crop. Each factor, rainfall and temperature, is divided into three levels as shown in Table 2-6 and Table 2-7.

Table 2-6. Rainfall Levels

| Level | Rainfall <br> (Factor A) |
| :---: | :--- |
| 0 | 2 inches |
| 1 | 3 inches |
| 2 | 4 inches |

Table 2-7. Temperature Levels

| Level | Temperature <br> (Factor B) |
| :---: | :---: |
| 0 | $76-80$ degrees |
| 1 | $81-85$ degrees |
| 2 | $86-90$ degrees |

A particular plot planted with the crop might appear in any one of the nine different combinations of these levels with the two factors. For example, one combination might be 2 inches of rain and an average temperature between 76 degrees and 80 degrees, recorded as $(0,0)$. Call these combinations cells.

The researchers set up 18 plots in various geographical locations chosen so that two plots fall in each of the nine cells. To measure the productivity of a particular plot, they record the crop production. Let rainfall be factor A and temperature be factor B. Table 2-8 shows their results.

Table 2-8. Plot Production

| $(\mathbf{A}, \mathbf{B})$ | Bushels Produced <br> from Each Plot |
| :---: | :---: |
| $(0,0)$ | 128122 |
| $(0,1)$ | 113108 |
| $(0,2)$ | 116116 |
| $(1,0)$ | 132129 |

Table 2-8. Plot Production (Continued)

| $(\mathbf{A}, \mathbf{B})$ | Bushels Produced <br> from Each Plot |
| :---: | :---: |
| $(1,1)$ | 119121 |
| $(1,2)$ | 126113 |
| $(2,0)$ | 118114 |
| $(2,1)$ | 141133 |
| $(2,2)$ | 121123 |

To perform a two-way analysis of variance in the fixed-effect model using ANOVA2Way, you store all the numbers of bushels in a double-precision array $\mathbf{y}$ of size 18. The integer arrays levelA and levelB record the cells in which observations were made. For any particular $i$, you set these arrays such that $y_{i}$ is the number of bushels a plot produces in the $\left(\operatorname{level} A_{i}, \operatorname{level} B_{i}\right)$ cell. For example:

$$
\begin{gathered}
\left(\text { level }_{i} \text {,level }_{i}\right)=(0,1) \\
y_{i}=113 \text { or } 108
\end{gathered}
$$

are valid combinations. Therefore, you can set up the input arrays $\mathbf{y}$, levelA, and levelB in this example for ANOVA2Way as follows:

$$
\begin{gathered}
y=128,122,113,108,116,132,129,119,121,126,113,118,114,141,133,121,123 \\
\text { levelA }=0,0,0,0,0,0,1,1,1,1,1,1,2,2,2,2,2,2 \\
\text { levelB }
\end{gathered}=0,0,1,1,2,2,0,0,1,1,2,2,0,0,1,1,2,24
$$

Running the code in the following example produces:

$$
\begin{gathered}
\operatorname{sig} A=0.026 \\
\operatorname{sig} B=0.203 \\
\operatorname{sig} A B=0.0018
\end{gathered}
$$

For a level of significance such as 0.05 , the ANOVA2Way results show that the researchers cannot reject the hypotheses that the combination of rainfall and temperature has any effect on the crop yield. In other words, the combination of rainfall and temperature has a significant effect on crop yield.

## Example

double y[18], sigA, sigB, sigAB, info[4][4];
int levelA[18], levelB[18];
int $L, a, b ;$
int status;
L $=2$; /* two observations per cell */
a $=3$; /* three levels for factor A, Rainfall */
$\mathrm{b}=3$; /* three levels for factor B, Temperature */
/* Read in recorded data y[18], levelA[18], levelB[18]. */
status $=$ ANOVA2Way (y, levelA, levelB, 18, L, $a, b, i n f o, \& s i g A$, \&sigB, \&sigAB);

## ANOVA3Way

```
int status = ANOVA3Way (double y[], int levelA[], int levelB[], int levelC[],
    int N, int L, int a, int b, int c, void *info,
    double *sigA, double *sigB, double *sigC,
    double *sigAB, double *sigAC, double *sigBC,
    double *sigABC);
```


## Purpose

Takes an array of experimental observations made at various levels of three factors and performs a three-way analysis of variance in any of the following models:

- Model 1: Fixed effects with interaction and $\mathbf{L}>1$ observations per cell.
- Model 2: Any of the six mixed-effects models, where one or two factors are taken to have fixed effects but the remaining factors are taken to have random effects, with interaction and $\mathbf{L}>1$ observations per cell.
- Model 3: Random effects with interaction and $\mathbf{L}>1$ observations per cell.

Any ANOVA looks for evidence that the factors, or interactions among the factors, have a significant effect on experimental outcomes. The method for finding significance varies among models.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Array of experimental data of <br> $\mathbf{N}=\|\mathbf{a}\| \times\|\mathbf{b}\| \times\|\mathbf{c}\| \times \mathbf{L}$ elements. |
| levelA | integer array | The $i^{\text {th }}$ element tells in what level of factor A <br> the $i^{\text {th }}$ observation falls. |
| levelB | integer array | The $i^{\text {th }}$ element tells in what level of factor B <br> the $i^{\text {th }}$ observation falls. |
| levelC | integer array | The $i^{\text {th }}$ element tells in what level of factor C <br> the $i^{\text {th }}$ observation falls. |
| $\mathbf{N}$ | integer | Total number of observations. |
| $\mathbf{L}$ | integer | Number of observations per cell. |


| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{b}$ | integer | Number of levels in factor B; negative if <br> B is a random effect. |
| $\mathbf{c}$ | integer | Number of levels in factor C; negative if <br> C is a random effect. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| info | 2D array | $\left.\begin{array}{llll}\text { An 8-by-4 matrix as follows: } \\ s s a & d o f a & m s a & f a \\ s s b & d o f b & m s b & f b \\ s s c & d o f c & m s c & f c \\ s s a b & d o f a b & m s a b & f a b \\ s s a c & d o f a c & m s a c & f a c \\ s s a b c & d o f a b c & \text { msabc } & \text { fabc } \\ s s e & d o f e & m s e & 0.0\end{array}\right]$ |
|  |  | where $s s$ designates sums of squares, <br> dof designates degrees of freedom of $s s$, <br> ms designates mean squares, and <br> $f$ designates F-distributions, depending on <br> the statistical model. |
| sigA | double-precision | Level of significance at which you must <br> reject hypothesis A. |
| sigB | double-precision | Level of significance at which you must <br> reject hypothesis B. |
| sigC | double-precision | Level of significance at which you must <br> reject hypothesis C. |
| sigAB | double-precision | Level of significance at which you must <br> reject hypothesis AB. |
| sigAC | double-precision | Level of significance at which you must <br> reject hypothesis AC. |


| Name | Type | Description |
| :--- | :--- | :--- |
| sigBC | double-precision | Level of significance at which you must <br> reject hypothesis BC. |
| sigABC | double-precision | Level of significance at which you must <br> reject hypothesis ABC. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Using This Function

## Factors, Levels, and Cells

A factor is a way of categorizing data. You can categorize data into levels, beginning with level 0 . For example, if you perform a measurement on individuals, such as counting the number of sit-ups they can perform, one such categorization method is age. For age, you might have three levels, as shown in Table 2-9.

Table 2-9. Age Levels

| Level | Age |
| :---: | :---: |
| 0 | 6 years to 10 years |
| 1 | 11 years to 15 years |
| 2 | 16 years to 20 years |

Another possible factor is eye color, with the levels shown in Table 2-10.
Table 2-10. Eye Color Levels

| Level | Eye Color |
| :---: | :---: |
| 0 | blue |
| 1 | brown |
| 2 | green |
| 3 | hazel |

A third factor might be height with levels in blocks of 10 cm . A cell of data consists of all those experimental observations that fall in particular levels of the three factors. In this
instance, a cell might consist of those observations made on hazel-eyed individuals between 11 years old and 15 years old who are between 151 cm and 160 cm tall. The number of observations that fall in each cell must be a constant number $\mathbf{L}$ that does not vary between cells.

## Random and Fixed Effects

A factor is taken as a random effect when the factor has a large population of levels you want to draw conclusions about, but that you cannot sample at all levels. Levels are sampled at random in the hope of generalizing about all levels.

A factor is taken as a fixed effect when the factor can be sampled from all levels you want to draw conclusions about.

The input parameters $\mathbf{a}, \mathbf{b}$, and $\mathbf{c}$ represent the number of levels in factors $\mathrm{A}, \mathrm{B}$, and C , respectively. If factor $A$ is random, set a to a negative value. In the same way, set $\mathbf{b}$ and $\mathbf{c}$ to negative values if B and C are random.

## General Method

Each of the previous models breaks up the total sum of squares ( $t s s$ ), which is a measure of the total variation of the data from the overall population mean, into a number of component sums of squares, so that:

$$
t s s=s s a+s s b+s s c+s s a b+s s a c+s s b c+s s a b c+s s e
$$

Each component in the sum is a measure of variation attributed to a certain factor or interaction among the factors. In this instance, $s s a$ is a measure of the variation as a result of factor A ; $s s b$ is a measure of the variation as a result of factor B ; $s s c$ is a measure of the variation as a result of factor C ; ssab is a measure of the variation as a result of the interaction between factors A and B ; and so on for $s s a c, s s b c$, and $s s a b c$. The variable sse is a measure of the variation as a result of random fluctuation.

If factor A has a strong effect on the experimental observations, msa is relatively large. You can look at specific ratios of these averages because you know how they are statistically distributed. You can therefore determine how likely it is that factor A is as relatively large as it is.

## Statistical Model

Let $y_{p, q, r, s}$ be the $s^{\text {th }}$ observation at the $p^{\text {th }}, q^{\text {th }}$, and $r^{\text {th }}$ levels of A, B, and C, respectively, where $s=0,1, \ldots, \mathbf{L}-1$.

Express each observation as the sum of eight components so that:

$$
y_{p, q, r, s}=\mu+\alpha_{p}+\beta_{q}+\Upsilon_{r}+(\alpha \beta)_{p, q}+(\alpha \Upsilon)_{p, r}+(\beta \Upsilon)_{q, r}+(\alpha \beta \Upsilon)_{p, q, r}+\varepsilon_{p, q, r, s}
$$

where $\quad \mu$ represents a standard effect present in each observation $\alpha_{p}, \beta_{q}$, and $\Upsilon_{r}$ are the effects of factors $\mathrm{A}, \mathrm{B}$, and C respectively $(\alpha \beta)_{p, q},(\alpha \Upsilon)_{p, r},(\beta \Upsilon)_{q, r}$, and $(\alpha \beta \Upsilon)_{p, q, r}$ are the effects of the corresponding interactions
$\varepsilon_{p, q, r, s}$ is a random fluctuation

## Assumptions

- Assume that for each $p, q, r$, and $s, \varepsilon_{p, q, r, s}$ is normally distributed with mean 0 and variance $\sigma_{e}^{2}$.
- If a factor such as A is fixed, assume that the populations of measurements at each level are normally distributed with mean $\alpha_{p}$ and variance $\sigma_{A}^{2}$. All the populations at each of the levels have the same variance. In addition, assume that all the $\alpha_{p}$ means sum to zero. Make analogous assumptions for B and C .
- If a factor such as A is random, assume that the effect of the level of A itself, $\alpha_{p}$, is a random variable normally distributed with mean 0 and variance $\sigma_{A}^{2}$. Make analogous assumptions for B and C .
- If all the factors, such as A and B, associated with the effect of an interaction $(\alpha \beta)_{p, q}$ are fixed, assume that the populations of measurements at each level are normally distributed with mean $(\alpha \beta)_{p, q}$ and variance $\sigma_{A B}^{2}$. For any fixed $p$, the $(\alpha \beta)_{p, q}$ means sum to zero when summing over all $q$. Similarly, for any fixed $q$, the $(\alpha \beta)_{p, q}$ means sum to zero when summing over all $p$.
- If any of the factors, such as $A$ and $B$, associated with the effect of an interaction $(\alpha \beta)_{p, q}$ are random, assume that the effect is a random variable normally distributed with mean 0 and variance $\sigma_{A B}^{2}$. If A is fixed but B is random, assume also that for any fixed $q$, the $(\alpha \beta)_{p, q}$ means sum to zero when summing over all $p$. Similarly, if B is fixed but A is random, assume also that for any fixed $p$, the $(\alpha \beta)_{p, q}$ means sum to zero when summing over all $q$.
- Assume that all effects taken to random variables are independent.


## Hypotheses

Each of the following hypotheses are different ways of stating that a factor or an interaction among factors has no effect on experimental outcomes. Start by assuming that there are no effects and then seek evidence to contradict these assumptions. The seven hypotheses are as follows:

- For hypothesis $\mathrm{A}, \alpha_{p}=0$ for all levels of $p$ if factor A is fixed; $\sigma_{A}^{2}=0$ if factor A is random.
- For hypothesis $\mathrm{B}, \beta_{q}=0$ for all levels of $q$ if factor B is fixed; $\sigma_{B}^{2}=0$ if factor B is random.
- For hypothesis $\mathrm{C}, \Upsilon_{r}=0$ for all levels of $r$ if factor C is fixed; $\sigma_{C}^{2}=0$ if factor C is random.
- For hypothesis $\mathrm{AB},(\alpha \beta)_{p, q}=0$ for all levels of $p$ and $q$ if factors A and B are fixed; $\sigma_{A B}^{2}=0$ if either factor A or B is random.
- For hypothesis AC, $(\alpha \Upsilon)_{p, r}=0$ for all levels of $p$ and $r$ if factors A and C are fixed; $\sigma_{A C}^{2}=0$ if either factor A or C is random.
- For hypothesis $\mathrm{BC},(\beta \Upsilon)_{q, r}=0$ for all levels of $q$ and $r$ if factors B and C are fixed; $\sigma_{B C}^{2}=0$ if either factor B or C is random.
- For hypothesis $\mathrm{ABC},(\alpha \beta \Upsilon)_{p, q, r}=0$ for all levels of $p, q$, and $r$ if factors $\mathrm{A}, \mathrm{B}$, and C are fixed; $\sigma_{A B C}^{2}=0$ if any of the factors $\mathrm{A}, \mathrm{B}$, or C are random.


## Testing the Hypotheses

For each hypothesis, ANOVA3Way generates a number so that if the hypothesis is true, that number is from a particular F-distribution.

For example, in the fixed-effects model, the number $f a=m s a / m s e$, associated with hypothesis A, is from an F-distribution with $\mathbf{a}-1$ and $a b c(\mathbf{L}-1)$ degrees of freedom, given that hypothesis A is true. ANOVA3Way calculates the probability that a number taken from a particular F-distribution is larger than the F-value. For example:

$$
\operatorname{sig} A=\operatorname{prob}(X>f a) \quad \text { where } X \text { is from } \mathrm{F}(a-1, a b c(L-1))
$$

Use the probabilities $\operatorname{sig} A, \operatorname{sig} B, \operatorname{sig} C, \operatorname{sig} A B, \operatorname{sig} A C, \operatorname{sigBC}$, and $\operatorname{sig} A B C$ to determine when to reject the associated hypotheses $\mathrm{A}, \mathrm{B}, \mathrm{C}, \mathrm{AB}, \mathrm{AC}, \mathrm{BC}$, and ABC by choosing a level of significance for each hypothesis. The level of significance determines how likely you are to reject the hypothesis when it is in fact true. Thus, the level of significance should be small, for example, 0.05 . Remember that the smaller the level of significance, the less likely you are to reject the hypothesis.

Reject a particular hypothesis when the associated output parameter $\operatorname{sig} A, \operatorname{sig} B, \operatorname{sig} C, \operatorname{sig} A B$, $\operatorname{sig} A C, \operatorname{sigBC}$, or $\operatorname{sig} \mathbf{A B C}$ is less than the level of significance you choose for that hypothesis. If A is a random effect, the chosen level of significance is 0.05 , and $\operatorname{sig} \mathbf{A}=0.03$, you must
reject the hypothesis that $\sigma_{A}^{2}=0$ and conclude that factor $A$ has an effect on the experimental observations.

With some models, no appropriate tests exist for certain hypotheses. In these cases, ANOVA3Way sets the output parameters directly involved with the testing of those hypotheses to -1.0 .

## Formulas

Let $y_{p, q, r, s}$ be the $s^{\text {th }}$ observation at the $p^{t h}, q^{t h}$, and $r^{\text {th }}$ levels of A, B, and C, respectively, where $s=0,1, \ldots, \mathbf{L}-1$.

Let:

$$
\begin{gathered}
a a=|a| \\
b b=|b| \\
c c=|c| \\
T_{p, q, r}=\sum_{s=0}^{L-1} y_{p, q, r, s} \\
T_{p, q}=\sum_{r=0}^{c c-1} T_{p, q, r} \\
T_{p, r}=\sum_{q=0}^{b b-1} T_{p, q, r} \\
T_{q, r}=\sum_{p=0}^{a a-1} T_{p, q, r} \\
T_{p}=\sum_{q=0}^{b b-1} T_{p, q}
\end{gathered}
$$

$$
\begin{aligned}
& T_{q}=\sum_{p=0}^{a a-1} T_{p, q} \\
& T_{r}=\sum_{p=0}^{a a-1} T_{p, r}
\end{aligned}
$$

$T=$ the total sum of all observations:

$$
\begin{aligned}
& A=\sum_{p=0}^{a a-1} \frac{T^{2}}{b b \times c c \times L} \\
& B=\sum_{q=0}^{b b-1} \frac{T_{p}^{2}}{a a \times c c \times L} \\
& C=\sum_{r=0}^{c c-1} \frac{T_{r}^{2}}{a a \times b b \times L} \\
& A B=\sum_{p=0}^{a a-1} \sum_{q=0}^{b b-1} \frac{T_{p, q}^{2}}{c c \times L} \\
& A C=\sum_{p=0}^{a a-1} \sum_{r=0}^{c c-1} \frac{T_{p, r}^{2}}{b b \times L} \\
& B C=\sum_{q=0}^{b b-1} \sum_{r=0}^{c c-1} \frac{T_{q, r}^{2}}{a a \times L} \\
& S=\sum_{p=0} \sum_{q=0}^{b a a} \sum_{r=0}^{b} \frac{T_{p, q, r}}{L} \\
& C F=\frac{T^{2}}{a a \times b b \times c c \times L}
\end{aligned}
$$

Then:

$$
\begin{aligned}
& s s a=A-C F \\
& m s a=\frac{s s a}{a a-1}=\frac{s s a}{d o f a} \\
& s s b=B-C F \\
& m s b=\frac{s s b}{b b-1}=\frac{s s b}{d o f b} \\
& s s c=C-C F \\
& m s c=\frac{s s c}{c c-1}=\frac{s s c}{d o f c} \\
& s s a b=A B-A-B+C F \\
& m s a b=\frac{s s a b}{(a a-1)(b b-1)}=\frac{s s a b}{d o f a b} \\
& \text { ssac }=A C-A-C+C F \\
& m s a c=\frac{s s a c}{(a a-1)(c c-1)}=\frac{s s a c}{d o f a c} \\
& s s b c=B C-B-C+C F \\
& m s b c=\frac{s s b c}{(b b-1)(c c-1)}=\frac{s s b c}{d o f b c} \\
& \text { ssabc }=S-A B-A C-B C+A+B+C-C F \\
& m s a b c=\frac{s s a b c}{(a a-1)(b b-1)(c c-1)}=\frac{s s a b c}{d o f a b c} \\
& m s e=\frac{s s e}{(a a \times b b \times c c)(L-1)}=\frac{s s e}{d o f e}
\end{aligned}
$$

$$
\left.\begin{array}{l}
f a= \begin{cases}\frac{m s a}{m s e} & \text { if } B \text { and } C \text { are fixed } \\
\frac{m s a}{m s a b} & \text { if } B \text { is random and } C \text { is fixed } \\
\frac{m s a}{m s a c} & \text { if } B \text { is fixed and } C \text { is random }\end{cases} \\
f b= \begin{cases}\frac{m s b}{m s e} & \text { if } A \text { and } C \text { are fixed } \\
\frac{m s b}{m s a b} & \text { if } A \text { is random and } C \text { is fixed } \\
\frac{m s b}{m s b c} & \text { if } A \text { is fixed and } C \text { is random }\end{cases} \\
f c= \begin{cases}\frac{m s c}{m s e} & \text { if } A \text { and } B \text { are fixed } \\
\frac{m s c}{m s a c} & \text { if } A \text { is random and } B \text { is fixed } \\
\frac{m s c}{m s b c} & \text { if } A \text { is fixed and } B \text { is random }\end{cases} \\
f a c= \begin{cases}\frac{m s a c}{m s e} & \text { if } B \text { is fixed } \\
\frac{m s a c}{m s a b c} & \text { if } B \text { is random }\end{cases} \\
f a b= \begin{cases}\frac{m s a b}{m s a b} & \text { if } C \text { is fixed }\end{cases} \\
f \text { if } C \text { is random }
\end{array}\right\}
$$

$$
\begin{gathered}
f b c= \begin{cases}\frac{m s b c}{m s e} & \text { if } A \text { is fixed } \\
\frac{m s b c}{m s a b c} & \text { if } A \text { is random }\end{cases} \\
f a b c=\frac{m s a b c}{m s e}
\end{gathered}
$$

If:

$$
\begin{gathered}
f=\frac{m s_{1}}{m s_{2}} \\
m s_{1}=\frac{s s_{1}}{d o f_{1}} \\
m s_{2}=\frac{s s_{2}}{d o f_{2}}
\end{gathered}
$$

assume that $f$ is from an F -distribution with $d o f_{1}$ and $d o f_{2}$ degrees of freedom.

## Example

Suppose that researchers want to know how the number of hours of sunlight, the amount of rainfall, and the average temperature affect the yield of a crop. Each factor, sunlight, rainfall, and temperature, is divided into three levels as shown in Tables 2-11, 2-12, and 2-13.

Table 2-11. Sunlight Levels

| Level | Sunlight <br> (Factor A) |
| :---: | :--- |
| 0 | 5 hours |
| 1 | 6 hours |
| 2 | 7 hours |

Table 2-12. Rainfall Levels

| Level | Rainfall <br> (Factor B) |
| :---: | :--- |
| 0 | 2 inches |
| 1 | 3 inches |
| 2 | 4 inches |

Table 2-13. Temperature Levels

| Level | Temperature <br> (Factor C) |
| :---: | :---: |
| 0 | $76-80$ degrees |
| 1 | $81-85$ degrees |
| 2 | $86-90$ degrees |

A particular plot planted with the crop might appear in any one of the 27 different combinations of these levels with the three factors. For example, one combination might be 6 hours of sunlight with 2 inches of rainfall and an average temperature between 76 degrees and 80 degrees, recorded as $(1,0,0)$. Call these combinations cells.

The researchers set up 54 plots in various geographical locations chosen so that two plots fall in each of the 27 cells. To measure the productivity of a particular plot, they record the crop production. Let sunlight be factor A , rainfall be factor B , and temperature be factor C .

Table 2-14 shows their results.
Table 2-14. Plot Production

| $(\mathbf{A}, \mathbf{B}, \mathbf{C})$ | Bushels Produced <br> from Each Plot |
| :---: | :---: |
| $(0,0,0)$ | 128122 |
| $(0,0,1)$ | 113108 |
| $(0,0,2)$ | 116116 |
| $(0,1,0)$ | 132129 |
| $(0,1,1)$ | 119121 |

Table 2-14. Plot Production (Continued)

| $(\mathbf{A}, \mathbf{B}, \mathbf{C})$ | Bushels Produced <br> from Each Plot |
| :---: | :---: |
| $(0,1,2)$ | 126113 |
| $(0,2,0)$ | 118114 |
| $(0,2,1)$ | 141133 |
| $(0,2,2)$ | 121123 |
| $(1,0,0)$ | 119118 |
| $(1,0,1)$ | 111115 |
| $(1,0,2)$ | 143140 |
| $(1,1,0)$ | 127129 |
| $(1,2,2)$ | 112113 |
| $(1,1,1)$ | 128120 |
| $(1,1,2)$ | 122121 |
| $(1,2,0)$ | 114115 |
| $(1,2,1)$ | 116113 |
| $(2,0,0)$ | 135131 |
| $(2,0,1)$ | 145145 |
| $(2,0,2)$ | 152147 |
| $(2,1,0)$ | 137141 |
| $(2,1,1)$ | 171171 |
| $(2,1,2)$ | 135131 |
| $(2,2,0)$ | 143144 |
| $(2,2,1)$ | 145147 |
| $(2,2,2)$ | 121123 |
|  |  |
| 121 |  |

To perform a three-way analysis of variance in the fixed-effect model using ANOVA3Way, you store all the numbers of bushels in a double-precision array $\mathbf{y}$ of size 54. The integer arrays levelA, levelB, and levelC record the cells in which observations were made. For any
particular $i$, you set these arrays such that $y_{i}$ is the number of bushels a plot produces in the (levelA $A_{i}$ levelB $_{i}$, levelC ${ }_{i}$ ) cell. For example:

$$
\begin{gathered}
\left(\text { level }_{i} \text {,level }_{i}, \text { level }_{i}\right)=(0,1,1) \\
y_{i}=119 \text { or } 121
\end{gathered}
$$

are valid combinations. Therefore, you can set up the input arrays $\mathbf{y}$, levelA, levelB, and levelC in this example for ANOVA3Way as follows:

$$
\begin{gathered}
y=128,122,113,108,116,116,132,129, \ldots \\
\text { levelA }=0,0,0,0,0,0,0,0, \ldots \\
\text { level } B=0,0,0,0,0,0,1,1, \ldots \\
\text { levelC }=0,0,1,1,2,2,0,0, \ldots
\end{gathered}
$$

Running the code in the following example produces:

$$
\begin{gathered}
\operatorname{sig} A=1.11 e^{-16} \\
\operatorname{sig} B=1.3 e^{-8} \\
\operatorname{sig} C=0.0072 \\
\operatorname{sig} A B=1.2 e^{-8} \\
\operatorname{sig} A C=2.0 e^{-4} \\
\operatorname{sig} B C=4.5 e^{-10} \\
\operatorname{sig} A B C=4.8 e^{-10}
\end{gathered}
$$

For a level of significance such as 0.05 , the ANOVA3Way results show that the researchers must reject the hypotheses that sunlight, rainfall, and temperature have no effect on the crop yield. In other words, all three factors have a significant effect on crop yield.

## Example

```
double y[54], sigA, sigB, sigC,sigAB, sigAC, sigBC, sigABC,info[8][4];
int levelA[54], levelB[54], levelC[54];
int L, a, b, c;
int status;
L = 2; /* two observations per cell */
a = 3; /* three levels for factor A, Sunlight */
b = 3; /* three levels for factor B, Rainfall */
c = 3; /* three levels for factor C, Temperature */
/* Read in recorded data y[54], levelA[54], levelB[54], and
levelc[54]. */
status = ANOVA3Way(y, levelA, levelB, levelC, 54, L, a, b, c, info,
    &sigA, &sigB, &sigC, &sigAB, &sigAC, &sigBC,
    &sigABC);
```


## ArbitraryWave

```
int status = ArbitraryWave (int n, double amp, double f, double *phase,
    double waveTable[], int tableSize, int interp,
    double x[]);
```


## Purpose

Generates an array that contains an arbitrary wave, with each cycle described by an interpolated version of the waveTable you specify. ArbitraryWave generates the output array $\mathbf{x}$ according to the following formula:

$$
x_{i}=a m p \times \operatorname{arb}(\text { phase }+f \times 360.0 \times i)
$$

where $\operatorname{arb}(p)=\mathrm{WT}(p$ modulo 360.0$)$
$f$ is frequency in cycles per sample
ArbitraryWave calculates $\mathrm{WT}(x)$ according to the following interpolation values:

$$
\mathrm{WT}(x)= \begin{cases}\text { waveTable }_{i x} & \text { for } \text { interp }=0 \\ \text { waveTable }_{i x}+d x\left(\text { waveTable }_{\left.(i x+1) \%_{\text {tableSize }}-\text { waveTable }_{i x}\right)} \text { for interp }=1\right.\end{cases}
$$

where $i x=($ int $) x$
$d x=x-($ int $) x$
(int) is the integral part of the variable $x$
You can use ArbitraryWave to simulate a continuous acquisition from an arbitrary wave function generator. The unit of the input phase is in degrees, and ArbitraryWave sets phase to ( $\mathbf{p h a s e}+\mathbf{f} \times 360.0 \times \mathbf{n}$ ) modulo 360.0 before it returns.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples to generate. |
| amp | double-precision | Amplitude of the generated signal. |
| $\mathbf{f}$ | double-precision | Frequency of the generated signal, in <br> normalized units of cycles/sample. |
| phase | double-precision | Points to the initial phase, in degrees, of the <br> generated signal. |


| Name | Type | Description |
| :--- | :--- | :--- |
| waveTable | double-precision array | Contains equally spaced samples of <br> one cycle of the generated signal. |
| tableSize | integer | Number of elements the waveTable <br> array contains. |
| interp | integer | Determines the type of interpolation to use <br> to generate the arbitrary wave signal from <br> the waveTable samples. <br> $0=$ no interpolation <br> $1=$ linear interpolation |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| phase | double-precision | Upon completion of ArbitraryWave, <br> phase points to the phase of the next <br> portion of the signal. Use this parameter in <br> the next call to ArbitraryWave to simulate <br> a continuous function generator. |
| $\mathbf{x}$ | double-precision array | Contains the generated arbitrary <br> wave signal. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## AutoPowerSpectrum

```
int status = AutoPowerSpectrum (double x[], int n, double dt,
    double autoSpectrum[], double *df);
```


## Purpose

Calculates the single-sided, scaled auto power spectrum of a time-domain signal. The auto power spectrum is defined as:

$$
\frac{\mathrm{FFT}(X) \mathrm{FFT}^{2}(X)}{n^{2}}
$$

where $\quad n$ is the number of points in the signal array $X$ * denotes a complex conjugate

AutoPowerSpectrum converts the auto power spectrum to a single-sided form.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Contains the time-domain signal. |
| $\mathbf{n}$ | integer | Number of elements in the input array. <br> $\mathbf{n}$ must be a power of 2. |
| $\mathbf{d t}$ | double-precision | Sampling period of the time-domain signal, <br> usually in seconds. <br> $\mathbf{d t}=1 / f s$, where $f s$ is the sampling <br> frequency of the time-domain signal. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| autoSpectrum | double-precision array | Single-sided amplitude spectrum magnitude <br> in volts, root-mean-square, if the input <br> signal is in volts. If the input signal is not in <br> volts, the results are in input signal units, <br> root-mean-square. This array must be at <br> least $\mathbf{n} / 2$ elements long. |
| df | double-precision | Points to the frequency interval, in hertz, if <br> $\mathbf{d t}$ is in seconds. <br> $\mathbf{d f}=1 /(\mathbf{n} \times \mathbf{d t})$ |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## BackSub

```
int status = BackSub (void *a, double y[], int n, double x[]);
```


## Purpose

Solves the linear equations $\mathbf{a} \times \mathbf{x}=\mathbf{y}$ by backward substitution. BackSub assumes $\mathbf{a}$ is an $\mathbf{n}$-by-n lower triangular matrix in which the diagonal elements all equal one. BackSub obtains $\mathbf{x}$ using the following formulas:

$$
\begin{array}{r}
x_{n-1}=\frac{y_{n-1}}{a_{n-1, n-1}} \\
x_{i}=\frac{y_{i}-\sum_{j=i+1}^{n-1} a_{i, j} x_{j}}{a_{i, j}}
\end{array}
$$

BackSub can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array. Use BackSub in conjunction with LU and ForwSub to solve linear equations.

Refer to the LU function description for more information.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{a}$ | double-precision <br> 2D array | Input matrix. |
| $\mathbf{y}$ | double-precision array | Input vector. |
| $\mathbf{n}$ | integer | Dimension size of $\mathbf{a}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Solution vector. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* to solve a linear equation A*x = y */
double A[10][10], x[10], y[10];
int p[10]; /* permutation vector */
int sign, n;
n = 10;
LU(A, n, p, &sign); /* LU decomposition of A */
ForwSub(A, y, n, x, p); /* forward substitution */
BackSub(A, x, n, x); /* backward substitution */
```


## Bessel_CascadeCoef

```
int status = Bessel_CascadeCoef (double fs, double fl, double fh,
    IIRFilterPtr filterInformation);
```


## Purpose

Generates the set of cascade form filter coefficients to implement an IIR filter as specified by the Bessel filter model.
filterInformation is the pointer to the filter structure that contains the filter coefficients and the internal filter information. You must allocate this structure by calling AllocIIRFilterPtr before calling this cascade IIR filter design function.

To redesign another filter, you should first call FreeIIRFilterPtr to free the present filter structure and then call AllocIIRFilterPtr with the new type and order parameters before you call Bessel_CascadeCoef.

If the type and order remain the same, you can call Bessel_CascadeCoef without calling FreeIIRFilterPtr and AllocIIRFilterPtr. In this case, you should properly reset the filtering operation for that structure by calling ResetIIRFilter before the first call to IIRCascadeFiltering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| fs | double-precision | Specifies the sampling frequency in hertz. |
| fl | double-precision | Specifies the desired lower cutoff frequency <br> of the filter in hertz. |
| fh | double-precision | Specifies the desired upper cutoff frequency <br> of the filter in hertz. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| filterInformation | IIRFilterPtr | Pointer to the filter structure that contains <br> the filter coefficients and the internal filter <br> information. |
|  |  | Refer to the AllocIIRFilterPtr <br> function description for more information <br> about the filter structure. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Design a cascade lowpass Bessel IIR filter. */
double fs, fl, fh, x[256], y[256];
int type, order, n;
IIRFilterPtr filterInfo;
n = 256;
fs = 1000.0;
fl = 200.0;
order = 5;
type = 0; /* lowpass */
Uniform(n, 17, x);
filterInfo = AllocIIRFilterPtr(type, order);
if(filterInfo!=0) {
    Bessel_CascadeCoef(fs, fl, fh, filterInfo);
    IIRCascadeFiltering(x, n, filterInfo, y);
    FreeIIRFilterPtr(filterInfo);
}
```


## Bessel_Coef

```
int status = Bessel_Coef (int type, int order, double fs, double fl,
    double fh, double a[], int na, double b[],
    int nb);
```


## Purpose

Generates the set of filter coefficients to implement an IIR filter as specified by the Bessel filter model. type has the valid values as shown in Table 2-15.

Table 2-15. Valid type Values

| Value | Description |
| :---: | :--- |
| 0 | lowpass filter; fh is not used |
| 1 | highpass filter; fh is not used |
| 2 | bandpass filter |
| 3 | bandstop filter |

$\mathbf{a}$ and $\mathbf{b}$ are the reverse and forward filter coefficients. Use IIRFiltering to achieve the actual filtering:

$$
y_{n}=\frac{1}{a_{0}}\left(\sum_{i=0}^{n b-1} b_{i} x_{n-i}-\sum_{i=1}^{n a-1} a_{i} y_{n-i}\right)
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| type | integer | Controls the filter type of the Bessel <br> IIR filter coefficients. |
| order | integer | Order of the IIR filter. |
| fs | double-precision | Sampling frequency in hertz. |
| fl | double-precision | Desired lower cutoff frequency of the filter <br> in hertz. |
| fh | double-precision | Desired higher cutoff frequency of the filter <br> in hertz. |


| Name | Type | Description |
| :--- | :--- | :--- |
| na | integer | Number of coefficients in the a coefficient <br> array. |
| nb | integer | Number of coefficients in the $\mathbf{b}$ coefficient <br> array. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| a | double-precision array | Array that contains the reverse coefficients <br> of the designed IIR filter. |
| b | double-precision array | Array that contains the forward coefficients <br> of the designed IIR filter. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## BkmanWin

```
int status = BkmanWin (double x[], int n);
```


## Purpose

Applies a Blackman window to the $\mathbf{x}$ input signal. The following formula defines the Blackman window:

$$
w_{i}=0.42-0.5 \cos \left(\frac{2 \pi i}{n}\right)+0.08 \cos \left(\frac{4 \pi i}{n}\right) \quad \text { for } i=0,1, \ldots, n-1
$$

BkmanWin obtains the output signal using the following formula:

$$
x_{i}=x_{i} \times w_{i} \quad \text { for } i=0,1, \ldots, n-1
$$

The function performs the window operation in place; that is, the windowed data $\mathbf{x}$ replaces the input data $\mathbf{x}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Windowed data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## BlkHarrisWin

```
int status = BlkHarrisWin (double x[], int n);
```


## Purpose

Applies a 3-term Blackman-Harris window to the input sequence $\mathbf{x}$. If $\mathbf{y}$ represents the output sequence, BlkHarrisWin obtains the elements of $\mathbf{y}$ using the formula:

$$
y_{i}=x_{i}\left(0.42323-0.49755 \cos \left(\frac{2 \pi i}{n}\right)+0.07922 \cos \left(\frac{4 \pi i}{n}\right)\right)
$$

where $n$ is the number of elements in $x$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Contains the input signal. |
| $\mathbf{n}$ | integer | Number of elements in the input array. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Contains the signal after BlkHarrisWin <br> applies the Blackman-Harris window. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Bw_BPF

```
int status = Bw_BPF (double x[], int n, double fs, double fl, double fh,
    int order, double y[]);
```


## Purpose

Filters the input array using a digital bandpass Butterworth filter. Bw_BPF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f l}$ | double-precision | Lower cutoff frequency. |
| $\mathbf{f h}$ | double-precision | Higher cutoff frequency. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random signal and filter it using a fifth-order bandpass
Butterworth filter. The pass band is from 200.0 to 300.0. */
double x[256], y[256], fs, fl, fh;
int n, order;
int status;
n = 256;
fs = 1000.0;
fl = 200.0;
fh = 300.0;
order = 5;
Uniform (n, 17, x);
status = Bw_BPF (x, n, fs, fl, fh, order, y);
```


## Bw_BSF

```
int status = Bw_BSF (double x[], int n, double fs, double fl, double fh,
```

    int order, double y[]);
    
## Purpose

Filters the input array using a digital bandstop Butterworth filter. Bw_BSF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f l}$ | double-precision | Lower cutoff frequency. |
| $\mathbf{f h}$ | double-precision | Higher cutoff frequency. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random signal and filter it using a fifth-order bandstop
Butterworth filter. The stop band is from 200.0 to 300.0. */
double x[256], y[256], fs, fl, fh;
int n, order;
int status;
n = 256;
fs = 1000.0;
fl = 200.0;
fh = 300.0;
order = 5;
Uniform (n, 17, x);
status = Bw_BSF (x, n, fs, fl, fh, order, y);
```


## Bw_CascadeCoef

```
int status = Bw_CascadeCoef (double fs, double fl, double fh,
    IIRFilterPtr filterInformation);
```


## Purpose

Generates the set of cascade form filter coefficients to implement an IIR filter as specified by the Butterworth filter model.
filterInformation is the pointer to the filter structure that contains the filter coefficients and the internal filter information. You must allocate this structure by calling AllocIIRFilterPtr before calling this cascade IIR filter design function.

To redesign another filter, you should first call FreeIIRFilterPtr to free the present filter structure and then call AllocIIRFilterPtr with the new type and order parameters before you call Bw_CascadeCoef.

If the type and order remain the same, you can call Bw_CascadeCoef without calling FreeIIRFilterPtr and AllocIIRFilterPtr. In this case, you should properly reset the filtering operation for that structure by calling ResetIIRFilter before the first call to IIRCascadeFiltering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| fs | double-precision | Specifies the sampling frequency in hertz. |
| fl | double-precision | Specifies the desired lower cutoff frequency <br> of the filter in hertz. |
| fh | double-precision | Specifies the desired upper cutoff frequency <br> of the filter in hertz. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| filterInformation | IIRFilterPtr | Pointer to the filter structure that contains <br> the filter coefficients and the internal filter <br> information. |
|  |  | Refer to the AllocIIRFilterPtr <br> function description for more information <br> about the filter structure. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Design a cascade lowpass Butterworth IIR filter. */
double fs, fl, fh, x[256], y[256];
int type, order, n;
IIRFilterPtr filterInfo;
n = 256;
fs = 1000.0;
fl = 200.0;
order = 5;
type = 0; /* lowpass */
Uniform(n, 17, x);
filterInfo = AllocIIRFilterPtr(type, order);
if(filterInfo!=0) {
    Bw_CascadeCoef(fs, fl, fh, filterInfo);
    IIRCascadeFiltering(x, n, filterInfo, y);
    FreeIIRFilterPtr(filterInfo);
}
```


## Bw_Coef

```
int status = Bw_Coef (int type, int order, double fs, double fl, double fh,
                        double a[], int na, double b[], int nb);
```


## Purpose

Generates the set of filter coefficients to implement an IIR filter as specified by the Butterworth filter model. type has the valid values as shown in Table 2-16.

Table 2-16. Valid type Values

| Value | Description |
| :---: | :--- |
| 0 | lowpass filter; fh is not used |
| 1 | highpass filter; $\mathbf{f h}$ is not used |
| 2 | bandpass filter |
| 3 | bandstop filter |

$\mathbf{a}$ and $\mathbf{b}$ are the reverse and forward filter coefficients. Use IIRFiltering to achieve actual filtering:

$$
y_{n}=\frac{1}{a_{0}}\left(\sum_{i=0}^{n b-1} b_{i} x_{n-i}-\sum_{i=1}^{n a-1} a_{i} y_{n-i}\right)
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| type | integer | Controls the filter type of the Butterworth <br> IIR filter coefficients. |
| order | integer | Order of the IIR filter. |
| fs | double-precision | Sampling frequency in hertz. |
| fl | double-precision | Desired lower cutoff frequency of the filter <br> in hertz. |
| fh | integer | Desired higher cutoff frequency of the filter <br> in hertz. |
| na | integer | Number of coefficients in the a coefficient <br> array. |
| nb | Number of coefficients in the $\mathbf{b}$ coefficient <br> array. |  |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| a | double-precision array | Array that contains the reverse coefficients <br> of the designed IIR filter. |
| b | double-precision array | Array that contains the forward coefficients <br> of the designed IIR filter. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Bw_HPF

int status $=$ Bw_HPF (double $x[]$, int $n$, double fs, double fc, int order, double y[]);

## Purpose

Filters the input array using a digital highpass Butterworth filter. Bw_HPF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f c}$ | double-precision | Cutoff frequency. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random signal and filter it using a fifth-order highpass
Butterworth filter. */
double x[256], y[256], fs, fc;
int n, order;
int status;
n = 256;
fs = 1000.0;
fc = 200.0;
order = 5;
Uniform (n, 17, x);
status = Bw_HPF (x, n, fs, fc, order, y);
```


## Bw_LPF

```
int status = Bw_LPF (double x[], int n, double fs, double fc, int order,
```

    double y[]);
    
## Purpose

Filters the input array using a digital lowpass Butterworth filter. Bw_LPF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f c}$ | double-precision | Cutoff frequency. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random signal and filter it using a fifth-order lowpass
Butterworth filter. */
double x[256], y[256], fs, fc;
int n, order;
int status;
n = 256;
fs = 1000.0;
fc = 200.0;
order = 5;
Uniform (n, 17, x);
status = Bw_LPF (x, n, fs, fc, order, y);
```


## CascadeToDirectCoef

```
int status = CascadeToDirectCoef (IIRFilterPtr filterInformation,
    double a[], int na, double b[], int nb);
```


## Purpose

Converts from the cascade IIR coefficients the filterInformation structure contains to direct-form IIR coefficients in arrays $\mathbf{a}$ and $\mathbf{b}$. These two arrays must be allocated in the same way as the old-style direct coefficient design functions, for example, Bw_Coef.

To redesign another filter, you should first call FreeIIRFilterPtr to free the present filter structure and then call AllocIIRFilterPtr with the new type and order parameters before you call CascadeToDirectCoef.

For lowpass and highpass type filters, the direct coefficient arrays must equal (order +1 ).
For bandpass and bandstop type filters, the direct coefficient arrays must equal ( $2 \times$ order +1 ).

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| filterInformation | IIRFilterPtr | Pointer to the filter structure that contains <br> the filter coefficients and the internal filter <br> information. <br> Refer to the AllocI IRFilterPtr <br> function description for more information <br> about the filter structure. |
| na | integer | Specifies the number of coefficients in the <br> $\mathbf{a}$ coefficient array. <br> na $=$ order +1 for low- or highpass <br> filters <br> na $=2 \times$ order +1 for bandpass or <br> bandstop filters |
| nb | integer | Specifies the number of coefficients in the <br> b coefficient array. <br> $\mathbf{n b}=$ order +1 for low- or highpass <br> filters <br> $\mathbf{n b}=2 \times$ order +1 for bandpass or <br> bandstop filters |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| a | double-precision array | Array that contains the reverse <br> coefficients of the direct-form IIR filter. |
| b | double-precision array | Array that contains the forward <br> coefficients of the direct-form IIR filter. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Ch_BPF

```
int status = Ch_BPF (double x[], int n, double fs, double fl, double fh,
    double ripple, int order, double y[]);
```


## Purpose

Filters the input array using a digital bandpass Chebyshev filter. Ch_BPF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f l}$ | double-precision | Lower cutoff frequency. |
| $\mathbf{f h}$ | double-precision | Higher cutoff frequency. |
| ripple | double-precision | Pass band ripples in decibels. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random signal and filter it using a fifth-order bandpass
Chebyshev filter. The pass band is from 200.0 to 300.0. */
double x[256], y[256], fs, fl, fh, ripple;
int n, order;
int status;
n = 256;
fs = 1000.0;
fl = 200.0;
fh = 300.0;
ripple = 0.5;
order = 5;
Uniform (n, 17, x);
status = Ch_BPF (x, n, fs, fl, fh, ripple, order, y);
```


## Ch_BSF

```
int status = Ch_BSF (double x[], int n, double fs, double fl, double fh,
    double ripple, int order, double y[]);
```


## Purpose

Filters the input array using a digital bandstop Chebyshev filter. Ch_BSF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f l}$ | double-precision | Lower cutoff frequency. |
| $\mathbf{f h}$ | double-precision | Higher cutoff frequency. |
| ripple | double-precision | Pass band ripples in decibels. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random signal and filter it using a fifth-order bandstop
Chebyshev filter. The stop band is from 200.0 to 300.0. */
double x[256], y[256], fs, fl, fh, ripple;
int n, order;
int status;
n = 256;
fs = 1000.0;
fl = 200.0;
fh = 300.0;
ripple = 0.5;
order = 5;
Uniform (n, 17, x);
status = Ch_BSF (x, n, fs, fl, fh, ripple, order, y);
```


## Ch_CascadeCoef

```
int status = Ch_CascadeCoef (double fs, double fl, double fh, double ripple,
                        IIRFilterPtr filterInformation);
```


## Purpose

Generates the set of cascade form filter coefficients to implement an IIR filter as specified by the Chebyshev filter model.
filterInformation is the pointer to the filter structure that contains the filter coefficients and the internal filter information. You must allocate this structure by calling AllocIIRFilterPtr before calling this cascade IIR filter design function.

To redesign another filter, you should first call FreeIIRFilterPtr to free the present filter structure and then call AllocIIRFilterPtr with the new type and order parameters before you call Ch_CascadeCoef.

If the type and order remain the same, you can call Ch_CascadeCoef without calling FreeIIRFilterPtr and AllocIIRFilterPtr. In this case, you should properly reset the filtering operation for that structure by calling ResetIIRFilter before the first call to IIRCascadeFiltering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| fs | double-precision | Specifies the sampling frequency in hertz. |
| fl | double-precision | Specifies the desired lower cutoff frequency <br> of the filter in hertz. |
| fh | double-precision | Specifies the desired upper cutoff frequency <br> of the filter in hertz. |
| ripple | double-precision | Specifies the amplitude of the stopband <br> ripple in decibels. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| filterInformation | IIRFilterPtr | Pointer to the filter structure that contains <br> the filter coefficients and the internal filter <br> information. <br> Refer to the AllocIIRFilterPtr <br> function description for more information <br> about the filter structure. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Design a cascade lowpass Chebyshev IIR filter. */
double fs, fl, fh, ripple, x[256], y[256];
int type, order, n;
IIRFilterPtr filterInfo;
n = 256;
fs = 1000.0;
fl = 200.0;
ripple = 0.5;
order = 5;
type = 0; /* lowpass */
Uniform(n, 17, x);
filterInfo = AllocIIRFilterPtr(type, order);
if(filterInfo!=0) {
    Ch_CascadeCoef(fs, fl, fh, ripple, filterInfo);
    IIRCascadeFiltering(x, n, filterInfo, y);
    FreeIIRFilterPtr(filterInfo);
}
```


## Ch_Coef

```
int status = Ch_Coef (int type, int order, double fs, double fl, double fh,
    double ripple, double a[], int na, double b[],
    int nb);
```


## Purpose

Generates the set of filter coefficients to implement an IIR filter as specified by the Chebyshev filter model. type has the valid values as shown in Table 2-17.

Table 2-17. Valid type Values

| Value | Description |
| :---: | :--- |
| 0 | lowpass filter; fh is not used |
| 1 | highpass filter; fh is not used |
| 2 | bandpass filter |
| 3 | bandstop filter |

$\mathbf{a}$ and $\mathbf{b}$ are the reverse and forward filter coefficients. Use IIRFiltering to achieve the actual filtering:

$$
y_{n}=\frac{1}{a_{0}}\left(\sum_{i=0}^{n b-1} b_{i} x_{n-i}-\sum_{i=1}^{n a-1} a_{i} y_{n-i}\right)
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| type | integer | Controls the filter type of the Chebyshev IIR <br> filter coefficients. |
| order | integer | Order of the IIR filter. |
| fs | double-precision | Sampling frequency in hertz. |
| fl | double-precision | Desired lower cutoff frequency of the filter <br> in hertz. |
| fh | double-precision | Desired higher cutoff frequency of the filter <br> in hertz. |


| Name | Type | Description |
| :--- | :--- | :--- |
| ripple | double-precision | Amplitude of the stopband ripple in <br> decibels. |
| na | integer | Number of coefficients in the a coefficient <br> array. |
| nb | integer | Number of coefficients in the $\mathbf{b}$ coefficient <br> array. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| a | double-precision array | Array that contains the reverse coefficients <br> of the designed IIR filter. |
| b | double-precision array | Array that contains the forward coefficients <br> of the designed IIR filter. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Ch_HPF

```
int status = Ch_HPF (double x[], int n, double fs, double fc, double ripple,
                        int order, double y[]);
```


## Purpose

Filters the input array using a digital highpass Chebyshev filter. Ch_HPF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f c}$ | double-precision | Cutoff frequency. |
| ripple | double-precision | Pass band ripples in decibels. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random signal and filter it using a fifth-order highpass
Chebyshev filter. */
double x[256], y[256], fs, fc, ripple;
int n, order;
int status;
n = 256;
fs = 1000.0;
fc = 200.0;
ripple = 0.5;
order = 5;
Uniform (n, 17, x);
status = Ch_HPF (x, n, fs, fc, ripple, order, y);
```


## Ch_LPF

```
int status = Ch_LPF (double x[], int n, double fs, double fc, double ripple,
    int order, double y[]);
```


## Purpose

Filters the input array using a digital lowpass Chebyshev filter. Ch_LPF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f c}$ | double-precision | Cutoff frequency. |
| ripple | double-precision | Pass band ripples in decibels. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random signal and filter it using a fifth-order lowpass
Chebyshev filter. */
double x[256], y[256], fs, fc, ripple;
int n, order;
int status;
n = 256;
fs = 1000.0;
fc = 200.0;
ripple = 0.5;
order = 5;
Uniform (n, 17, x);
status = Ch_LPF (x, n, fs, fc, ripple, order, y);
```


## CheckPosDef

```
int status = CheckPosDef (void *A, int n, int *positiveDefinite);
```


## Purpose

Checks if the real, square input matrix is positive definite. A real, square matrix is positive definite if and only if it is symmetric and the quadratic form

$$
x^{T} A x>0
$$

is true for all nonzero vectors $x$. For more information on positive definite matrices, refer to Matrix Computations by G.H. Golub and C.F. VanLoan.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | double-precision <br> 2D array | Input square matrix. |
| $\mathbf{n}$ | integer | Number of elements in one dimension of <br> the matrix. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| positiveDefinite | integer | 1 if the input matrix is positive definite; <br> 0 otherwise. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Chirp

```
int status = Chirp (int n, double amp, double f1, double f2, double x[]);
```


## Purpose

Generates an array that contains a chirp pattern. Chirp generates the output array $\mathbf{x}$ according to the following formula:

$$
x_{i}=a m p \times \sin \left(\left(\frac{a}{2} i+b\right) i\right)
$$

where $\quad a=\frac{2 \pi \times(f 2-f 1)}{n}$
$b=2 \pi \times f 1$
and where $\quad f 1$ is the beginning frequency in cycles per sample $f 2$ is the ending frequency in cycles per sample

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples to generate. |
| amp | double-precision | Amplitude of the resulting signal. |
| f1 | double-precision | Beginning frequency of the resulting signal <br> in normalized units of cycles/sample. |
| f2 | double-precision | Ending frequency of the resulting signal in <br> normalized units of cycles/sample. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Contains the generated chirp pattern. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Cholesky

```
int status = Cholesky (void *A, int n, void *R);
```


## Purpose

Calculates the Cholesky factorization of a real, symmetric positive definite input matrix. If the input matrix is not positive definite, Cholesky returns an error.

The following formula defines the Cholesky factorization of an $\mathbf{n}$-by- $\mathbf{n}$ symmetric positive definite matrix $\mathbf{A}$ :

$$
A=R^{T} R
$$

where $\quad R$ is an upper triangular matrix of dimensions $\mathbf{n}$-by-n $R^{T}$ is the transpose of $R$

Cholesky factorization is similar to LU factorization for symmetric positive definite matrices. If the matrix in your application is positive definite, use Cholesky factorization rather than LU factorization for the following reasons:

- The algorithm is well defined.
- Numerical stability does not require pivoting.
- Cholesky factorization requires about half the programming time and less memory than LU factorization.


## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | double-precision <br> 2D array | Input square, positive definite matrix. |
| $\mathbf{n}$ | integer | Number of elements in one dimension of <br> the matrix. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{R}$ | double-precision <br> 2D array | Result matrix of the Cholesky <br> decomposition. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Clear1D

```
int status = Clear1D (double x[], int n);
```


## Purpose

Sets the elements of the $\mathbf{x}$ array to 0.0 .

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Cleared array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Clip

```
int status = Clip (double x[], int n, double upper, double lower, double y[]);
```


## Purpose

Clips the input array values. The range of the resulting output array is [lower : upper]. Clip obtains the $i^{\text {th }}$ element of the resulting array using the following formula:

$$
y_{i}= \begin{cases}\text { upper } & \text { if } x_{i}>\text { upper } \\ \text { lower } & \text { if } x_{i}<\text { lower } \\ x_{i} & \text { otherwise }\end{cases}
$$

Clip can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| upper | double-precision | Upper limit. |
| lower | double-precision | Lower limit. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Clipped array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## ConditionNumber

int status = ConditionNumber (void *A, int $n$, int m, int normType, double *C);

## Purpose

Calculates the condition number of a real input matrix A. The normType parameter indicates what type of norm to use to calculate the condition number. The input matrix $\mathbf{A}$ does not need to be square when normType is 2 -norm.

The following formula defines the condition number $\mathbf{c}$ of a matrix $\mathbf{A}$ :

$$
c=\|A\|_{p} \times\left\|A^{-1}\right\|_{p} \quad \text { where }\|A\|_{p} \text { is the p-norm of the matrix } A
$$

The normType value defines the p-norm behavior. For a 2-norm normType, $\mathbf{c}$ is the ratio of the largest singular value of $\mathbf{A}$ to the smallest singular value of $\mathbf{A}$.

The condition number of a matrix indicates how near to singular the matrix is. A matrix with a large condition number is nearly singular, and a matrix with a condition number close to one is far from singular. The condition number of a matrix is always greater than or equal to one, and it can help assess the accuracy of a solution to a linear system of equations and matrix inversion.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| A | double-precision <br> 2D array | Input matrix. If normType is 2-norm, the <br> matrix can be square or rectangular; <br> otherwise, it must be square. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in $\mathbf{A}$. |
| normType | integer | Type of p-norm function to use to calculate <br> the condition number. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{c}$ | double-precision | Condition number of the matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The normType parameter indicates what type of norm to use to calculate the condition number. Table 2-18 shows valid norm type values.

Table 2-18. Valid Norm Type Values

| Norm Type | Value | Meaning |
| :--- | :---: | :--- |
| 2-norm | 0 | Largest singular value of $\mathbf{A}$. |
| 1-norm | 1 | Largest column sum of $\mathbf{A .}$ |
| Frobenius-norm | 2 | Square root of the sum of the diagonal elements of $\mathbf{A}^{\mathbf{T}} \mathbf{A}$, <br> where $\mathbf{A}^{\mathbf{T}}$ is the complex conjugate transpose of $\mathbf{A}$. |
| Infinite-norm | 3 | Largest row sum of $\mathbf{A}$. |

## Contingency_Table

```
int status = Contingency_Table (int s, int k, int void *y, double *Test_Stat,
    double *Sig);
```


## Purpose

Creates a contingency table in which to classify and tally objects of experimentation according to two schemes of categorization. Use Contingency_Table to perform a test of homogeneity or a test of independence.

雨 Note For both tests, the math is identical. It is not necessary to specify which test you apply. The only difference is in the hypothesis you test.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{s}$ | integer | Number of random samples in the test of <br> homogeneity or the number of categories in <br> the first categorization scheme in the test of <br> independence. |
| $\mathbf{k}$ | integer | Number of categories in the test of <br> homogeneity or the number of categories in <br> the second categorization scheme in the test <br> of independence. |
| $\mathbf{y}$ | integer 2D array | Contingency table, indexed as an s-by-k <br> matrix. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| Test_Stat | double-precision | Use to calculate Sig. If the hypothesis <br> is true, Test_Stat is known to come <br> from a chi-square distribution with <br> $(\mathbf{s}-1) \times(\mathbf{k}-1)$ degrees of freedom. |
| Sig | double-precision | Level of significance at which you must <br> reject the hypothesis. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Using This Function

A contingency table is a table in which you can classify and tally objects of experimentation according to two schemes of categorization. For example, if the objects of experimentation are individuals, one scheme might be political affiliation: Know-Nothing, Tory, Whig, Mugwump, and so on. Another scheme might be to classify individuals according to how they vote on an issue.

## Chi-Square Test of Homogeneity

Take a random sample of a fixed size from each of the categories in one categorization scheme for the chi-square test of homogeneity. For each of the samples, categorize the objects of experimentation according to the second scheme and tally them. For example, you might pick 100 Know-Nothings, 100 Whigs, 100 Tories, and 100 Mugwumps. Count the number of individuals who vote a certain way for each category. This produces the contingency table shown in Table 2-19.

Table 2-19. Contingency Table

| Category | Yes | No | Undecided |
| :--- | :---: | :---: | :---: |
| Know-Nothing | 36 | 24 | 40 |
| Whig | 12 | 53 | 35 |
| Tory | 61 | 11 | 28 |
| Mugwump | 83 | 3 | 14 |

Notice that the sum of each of the rows equals 100.
Test the hypothesis that the populations from which you take each sample are identically distributed with respect to the second categorization scheme. For example, you can test the hypothesis that the four samples of politically affiliated individuals are distributed identically with respect to the way they vote. If this hypothesis is true, it means that a Mugwump you select at random is just as likely to vote yes as a Whig you select at random.

## Chi-Square Test of Independence

Take only one sample from the total population for the chi-square test of independence.
Categorize each object of experimentation and tally them in the two categorization schemes. If you select 500 individuals, for example, you might receive the results shown in Table 2-20.

Table 2-20. Contingency Table

| Category | Yes | No | Undecided |
| :--- | :---: | :---: | :---: |
| Know-Nothing | 18 | 15 | 18 |
| Whig | 55 | 93 | 38 |
| Tory | 101 | 83 | 20 |
| Mugwump | 16 | 31 | 12 |

Notice that the sum of each row is different but that the total number of individuals tallied is 500 .

Test the hypothesis that the categorization schemes are independent. For example, if you choose a person at random and he or she is a Mugwump, the hypothesis states that his or her political affiliation has no effect on how he or she votes on the issue you select.

## Testing the Hypothesis

Whichever test you use, you must choose a level of significance. This determines how likely you are to reject a true hypothesis. Thus, the level of significance should be small, for example, 0.05.

The output parameter $\mathbf{S i g}$ is the level of significance at which you reject the hypothesis:

$$
\text { Sig }=\operatorname{prob}(\chi \geq \text { Test_Stat })
$$

where $\quad \chi$ is a random variable from the chi-square distribution with $(s-1) \times(k-1)$ degrees of freedom

If $\mathbf{S i g}$ is less than the level of significance, you must reject the hypothesis.

## Formulas

Let $y_{p, q}$ be the number of occurrences in the $(p, q)^{t h}$ cell of the contingency table for $p=0,1, \ldots,(\mathbf{s}-1)$ and $q=0,1, \ldots,(\mathbf{k}-1)$.

Let:

$$
\begin{gathered}
y_{p}=\sum_{q=0}^{k-1} y_{p, q} \\
y_{q}=\sum_{p=0}^{s-1} y_{p, q} \\
y=\sum_{p=0}^{s-1} \sum_{q=0}^{k-1} y_{p, q} \\
e_{p, q}=\frac{y_{p} \times y_{q}}{y} \\
\text { Test_Stat }=\sum_{p=0}^{s-1} \sum_{q=0}^{k-1} \frac{\left[y_{p, q}-e_{p, q}\right]^{2}}{e_{p, q}}
\end{gathered}
$$

## Example

```
/* Generate a random contingency table. Because rows will not have
identical sums, use the chi-square test of independence. */
int s=10, k=10, y[10][10], i, j, status;
double Test_Stat, Sig, temp[1];
for(i=0; i<s; i++)
    for(j=0; j<k; j++)
    {
            WhiteNoise (1, 5, 17, temp);
            temp[0] += 6.0;
            y[i][j] = (int) temp[0];
        }
status = Contingency_Table (s, k, y, &Test_Stat, &Sig);
```


## Convolve

```
int status = Convolve (double x[], int n, double y[], int m, double cxy[]);
```


## Purpose

Finds the convolution of the $\mathbf{x}$ and $\mathbf{y}$ input arrays. Convolve obtains the convolution using the following formula:

$$
c x y_{i}=\sum_{k=a}^{b} x_{k} \times y_{i-k}
$$

$$
\text { where } \begin{aligned}
y & =0, b=i \quad \text { for } 0 \leq i<m \\
a & =i-m+1, b=i \quad \text { for } m \leq i<n \\
a & =i-m+1, b=n-1 \quad \text { for } n \leq i \leq n+m-1
\end{aligned}
$$

## [ 3 Note <br> This formula description assumes that $\mathbf{m} \leq \mathbf{n}$. For $\mathbf{m}>\mathbf{n}$, exchange ( $\mathbf{x}, \mathbf{y}$ ) and

 ( $\mathbf{m}, \mathbf{n}$ ) in the previous equations.
## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | $\mathbf{x}$ input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{y}$ | double-precision array | $\mathbf{y}$ input array. |
| $\mathbf{m}$ | integer | Number of elements in $\mathbf{y}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| cxy | double-precision array | Convolution array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Using This Function

The size of the output array must be at least ( $\mathbf{n}+\mathbf{m}-1$ ) elements long. This algorithm executes more efficiently if the sizes of the input arrays are a power of two.

## Example

```
/* Generate two arrays with random numbers and find their
convolution. */
double x[256], y[256], cxy[512];
int n, m;
n = 256;
m = 256;
Uniform (n, 17, x);
Uniform (m, 17, y);
Convolve (x, n, y, m, cxy);
```


## Copy1D

```
int status = Copy1D (double x, int n, double y[]);
```


## Purpose

Copies the elements of the $\mathbf{x}$ array. Use Copy1D to duplicate arrays for in place operations.

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Duplicated array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Correlate

```
int status = Correlate (double x[], int n, double y[], int m, double rxy[]);
```


## Purpose

Finds the correlation of the input arrays. Correlate obtains the correlation using the following formula:

$$
\begin{gathered}
r x y_{i}=\sum_{k=0}^{m-1} x_{k+n-1-i} y_{k} \\
y_{j}=0 \quad \text { when } j<0 \text { or } j \geq m \\
x_{i}=0 \quad \text { when } j<0 \text { or } j \geq n
\end{gathered}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | $\mathbf{y}$ input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x .}$ |
| $\mathbf{y}$ | double-precision array | $\mathbf{y}$ input array. |
| $\mathbf{m}$ | integer | Number of elements in $\mathbf{y}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| rxy | double-precision array | Correlation array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Using This Function

The size of the output array must be at least $(\mathbf{n}+\mathbf{m}-1)$ elements long.

## Example

```
/* Generate two arrays with random numbers and find their
correlation. */
double x[256], y[256], cxy[512];
int n, m;
n = 256;
m = 256;
Uniform (n, 17, x);
Uniform (m, 17, y);
Correlate (x, n, y, m, cxy);
```


## CosTaperedWin

```
int status = CosTaperedWin (double x[], int n);
```


## Purpose

Applies a cosine-tapered window to the input sequence $\mathbf{x}$. If $\mathbf{y}$ represents the output sequence, CosTaperedWin obtains the elements of $\mathbf{y}$ from the formula:

$$
y_{i}= \begin{cases}0.5 x_{i}\left(1-\cos \left(\frac{2 \pi i}{n}\right)\right) & i=0,1, \ldots, m-1 \\ x_{i} & i=m, m+1, n-m-1 \\ 0.5 x_{i}\left(1-\cos \left(\frac{2 \pi i}{n}\right)\right) & i=n-m, n-m+1, n-1\end{cases}
$$

where $m=\operatorname{round}\left(\frac{n}{10}\right)$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Contains the input signal. |
| $\mathbf{n}$ | integer | Number of elements in the input array. |

## Output

| Name | Type | Description |
| :---: | :---: | :--- |
| $\mathbf{x}$ | double-precision array | Contains the signal after CosTaperedWin <br> applies the cosine-tapered window. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CrossPowerSpectrum

```
int status = CrossPowerSpectrum (double x[], double y[], int n, double dt,
    double magSxy[], double phaseSxy[], double *df);
```


## Purpose

Calculates the single-sided, scaled cross power spectrum of two time-domain signals. The following formula defines the cross power spectrum:

$$
S x y=\frac{\operatorname{FFT}(y) \operatorname{FFT}(x)}{n^{2}} \quad \text { where } n \text { is the number of points in arrays } x \text { and } y
$$

magSxy and phaseSxy are single-sided magnitude and phase spectra of Sxy.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Time-domain signal $\mathbf{x .}$ |
| $\mathbf{y}$ | double-precision array | Time-domain signal $\mathbf{y}$. |
| $\mathbf{n}$ | integer | Number of elements in the input array. <br> $\mathbf{n}$ must be a power of 2. |
| $\mathbf{d t}$ | double-precision | Sampling period of the time-domain signal, <br> usually in seconds. $\mathbf{d t}=1 / f s$, where $f s$ is <br> the sampling frequency of the time-domain <br> signal. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| magSxy | double-precision array | Single-sided magnitude cross power <br> spectrum between signals $\mathbf{x}$ and $\mathbf{y}$ in volts <br> rms square if the input signals are in volts. If <br> the input signals are not in volts, the results <br> are in input signal units rms square. This <br> array must be at least $\mathbf{n} / 2$ elements long. |
| phaseSxy | double-precision array | Single-sided phase cross spectrum in <br> radians, showing the difference between the <br> phases of signal $\mathbf{y}$ and signal $\mathbf{x}$. This array <br> must be at least $\mathbf{n} / 2$ elements long. |
| df | double-precision | Points to the frequency interval, in hertz, if <br> $\mathbf{d t}$ is in seconds. $\mathbf{d f}=1 /(\mathbf{n} \times \mathbf{d t})$ |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CrossSpectrum

```
int status = CrossSpectrum (double x[], double y[], int n, double realSxy[],
    double imagSxy[]);
```


## Purpose

Calculates the double-sided cross power spectrum, $S x y$, of the input sequences $\mathbf{x}$ and $\mathbf{y}$ according to the following formula:

$$
S x y=\frac{\mathrm{FFT}^{*}(x) \mathrm{FFT}(y)}{n^{2}}
$$

where $n$ is the number of samples in both input sequences
$\mathrm{FFT}^{*}(x)$ is the complex conjugate of $\mathrm{FFT}(x)$
$\mathbf{n}$ must be a power of two. CrossSpect rum copies the input sequences to internal buffers before it calculates the FFTs. The output arrays are the real and imaginary parts of the cross spectrum.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Time-domain signal $\mathbf{x}$. |
| $\mathbf{y}$ | double-precision array | Time-domain signal $\mathbf{y}$. |
| $\mathbf{n}$ | integer | Number of elements in the input arrays. <br> $\mathbf{n}$ must be a power of 2. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| realSxy | double-precision array | Real part of the double-sided cross power <br> spectrum between signals $\mathbf{x}$ and $\mathbf{y}$. The size <br> of this array must be $\mathbf{n}$. |
| imagSxy | double-precision array | Imaginary part of the double-sided cross <br> power spectrum between signals $\mathbf{x}$ and $\mathbf{y}$. <br> The size of this array must be $\mathbf{n}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxAdd

```
int status = CxAdd (double xr, double xi, double yr, double yi, double *zr,
    double *zi);
```


## Purpose

Adds two complex numbers, $x$ and $y$. CxAdd obtains the resulting complex number, $z$, using the following formulas:

$$
\begin{aligned}
z r & =x r+y r \\
z i & =x i+y i
\end{aligned}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision | Real part of $x$. |
| $\mathbf{x i}$ | double-precision | Imaginary part of $x$. |
| $\mathbf{y r}$ | double-precision | Real part of $y$. |
| $\mathbf{y i}$ | double-precision | Imaginary part of $y$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z r}$ | double-precision pointer | Real part of $z$. |
| $\mathbf{z i}$ | double-precision pointer | Imaginary part of $z$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxAdd1D

```
int status = CxAdd1D (double xr[], double xi[], double yr[], double yi[],
    int n, double zr[], double zi[]);
```


## Purpose

Adds two 1D complex arrays, $\mathbf{x}$ and $\mathbf{y}$. CxAdd1D obtains the $i^{\text {th }}$ element of the resulting complex array, $\mathbf{z}$, using the following formulas:

$$
\begin{aligned}
z r_{i} & =x r_{i}+y r_{i} \\
z i_{i} & =x i_{i}+y i_{i}
\end{aligned}
$$

CxAdd1D can perform the operations in place; that is, the input and output complex arrays can be the same.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision array | Real part of $\mathbf{x .}$ |
| $\mathbf{x i}$ | double-precision array | Imaginary part of $\mathbf{x .}$ |
| $\mathbf{y r}$ | double-precision array | Real part of $\mathbf{y}$. |
| $\mathbf{y i}$ | double-precision array | Imaginary part of $\mathbf{y}$. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z r}$ | double-precision array | Real part of $\mathbf{z}$. |
| $\mathbf{z i}$ | double-precision array | Imaginary part of $\mathbf{z}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxCheckPosDef

```
int status = CxCheckPosDef (void *A, int n, int *positiveDefinite);
```


## Purpose

Checks if the complex, square input matrix $\mathbf{A}$ is positive definite. A complex, square matrix is positive definite if and only if it is symmetric and the quadratic form

$$
x^{H} A x>0
$$

is true for all nonzero vectors $x$. For more information on positive definite matrices, refer to Matrix Computations by G.H. Golub and C.F. VanLoan.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | ComplexNum 2D array | Input complex, square matrix. |
| $\mathbf{n}$ | integer | Number of elements in one dimension of <br> the matrix. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| positiveDefinite | integer | 1 if the input matrix is positive definite; <br> 0 otherwise. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```


## CxCholesky

```
int status = CxCholesky (void *A, int n, void *R);
```


## Purpose

Calculates the Cholesky factorization of a complex, symmetric positive definite input matrix. If the input matrix is not positive definite, CxCholesky returns an error.

The following formula defines the Cholesky factorization of an $\mathbf{n}$-by- $\mathbf{n}$ symmetric positive definite matrix $\mathbf{A}$ :

$$
A=R^{H} R
$$

where $\quad R$ is an upper triangular matrix of dimensions $\mathbf{n}$-by-n
$R^{H}$ is the complex conjugate transpose of $R$
Cholesky factorization is similar to LU factorization for symmetric positive definite matrices. If the matrix in your application is positive definite, use Cholesky factorization rather than LU factorization for the following reasons:

- The algorithm is well defined.
- Numerical stability does not require pivoting.
- Cholesky factorization requires about half the programming time and less memory than LU factorization.


## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| A | ComplexNum 2D array | Input complex, square matrix. |
| $\mathbf{n}$ | integer | Number of elements in one dimension of <br> the matrix. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{R}$ | ComplexNum 2D array | Result matrix of the Cholesky <br> decomposition. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```


## CxConditionNumber

```
int status = CxConditionNumber (void *A, int n, int m, int normType,
    double *C);
```


## Purpose

Calculates the condition number of a complex input matrix $\mathbf{A}$. The normType parameter indicates what type of norm to use to calculate the condition number. The input matrix $\mathbf{A}$ does not need to be square when normType is 2-norm.

The following formula defines the condition number $\mathbf{c}$ of a matrix $\mathbf{A}$ :

$$
c=\|A\|_{p} \times\left\|A^{-1}\right\|_{p} \quad \text { where }\|A\|_{p} \text { is the p-norm of the matrix } A
$$

The normType value defines the type of norm. For a 2-norm normType, $\mathbf{c}$ is the ratio of the largest singular value of $\mathbf{A}$ to the smallest singular value of $\mathbf{A}$.

The condition number of a matrix indicates how near to singular the matrix is. A matrix with a large condition number is nearly singular, and a matrix with a condition number close to one is far from singular. The condition number of a matrix is always greater than or equal to one, and it can help assess the accuracy of a solution to a linear system of equations and matrix inversion.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | ComplexNum 2D array | Input complex, square matrix. If normType <br> is 2-norm, the matrix can be square or <br> rectangular; otherwise, it must be square. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in A. |
| normType | integer | Type of p-norm function to use to calculate <br> the condition number. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{c}$ | double-precision | Complex condition number of the matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```

The normType parameter indicates what type of norm to use to calculate the condition number. Table 2-21 shows valid norm type values.

Table 2-21. Valid Norm Type Values

| Norm Type | Value | Meaning |
| :--- | :---: | :--- |
| 2-norm | 0 | Largest singular value of $\mathbf{A}$. |
| 1-norm | 1 | Largest column sum of $\mathbf{A .}$ |
| Frobenius-norm | 2 | Square root of the sum of the diagonal elements of $\mathbf{A}^{\mathbf{H}} \mathbf{A}$, <br> where $\mathbf{A}^{\mathbf{H}}$ is the complex conjugate transpose of $\mathbf{A}$. |
| Infinite-norm | 3 | Largest row sum of $\mathbf{A}$. |

## CxDeterminant

```
int status = CxDeterminant (void *A, int n, int matrixType, ComplexNum *det);
```


## Purpose

Calculates the complex determinant of a square, complex input matrix $\mathbf{A}$. The input matrix can be upper or lower triangular, general, or positive definite. For an upper or lower triangular matrix, the determinant equals the product of the diagonal elements of the matrix. For a positive definite matrix, CxDeterminant first calculates the Cholesky factorization of the input matrix and then calculates the determinant as the square of the determinant of $\mathbf{R}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| A | ComplexNum 2D array | Input complex, square matrix. |
| $\mathbf{n}$ | integer | Number of elements in one dimension of the <br> matrix. |
| matrixType | integer | Type of the input matrix. Choose the matrix <br> type correctly because it significantly <br> affects the speed of computation. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| det | ComplexNum | Complex determinant. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```

Table 2-22 shows valid matrix type values.
Table 2-22. Valid Matrix Type Values

| Matrix Type | Value |
| :--- | :---: |
| General matrix | 0 |
| Positive definite | 1 |
| Upper triangular | 2 |
| Lower triangular | 3 |

## CxDiv

```
int status = CxDiv (double xr, double xi, double yr, double yi, double *zr,
    double *zi);
```


## Purpose

Divides two complex numbers, $x$ and $y$. CxDiv obtains the resulting complex number, $z$, using the following formulas:

$$
\begin{aligned}
& z r=\frac{(x r \times y r+x i \times y i)}{y r^{2}+y i^{2}} \\
& z i=\frac{(x i \times y r-x r \times y i)}{y r^{2}+y i^{2}}
\end{aligned}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision | Real part of $x$. |
| $\mathbf{x i}$ | double-precision | Imaginary part of $x$. |
| $\mathbf{y r}$ | double-precision | Real part of $y$. |
| $\mathbf{y i}$ | double-precision | Imaginary part of $y$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z r}$ | double-precision | Real part of $z$. |
| $\mathbf{z i}$ | double-precision | Imaginary part of $z$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxDiv1D

```
int status = CxDiv1D (double xr[], double xi[], double yr[], double yi[],
    int n, double zr[], double zi[]);
```


## Purpose

Divides two 1D complex arrays, $\mathbf{x}$ and $\mathbf{y}$. CxDiv1D obtains the $i^{\text {th }}$ element of the resulting complex array, $\mathbf{z}$, using the following formulas:

$$
\begin{aligned}
& z r_{i}=\frac{\left(x r_{i} \times y r_{i}+x i_{i} \times y i_{i}\right)}{y r_{i}^{2}+y i_{i}^{2}} \\
& z i_{i}=\frac{\left(x i_{i} \times y r_{i}-x r_{i} \times y i_{i}\right)}{y r_{i}^{2}+y i_{i}^{2}}
\end{aligned}
$$

$\mathbf{z r}$ can be in place with $\mathbf{~ x r ; ~} \mathbf{z i}$ can be in place with $\mathbf{x i}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision array | Real part of $\mathbf{x}$. |
| $\mathbf{x i}$ | double-precision array | Imaginary part of $\mathbf{x}$. |
| $\mathbf{y r}$ | double-precision array | Real part of $\mathbf{y}$. |
| $\mathbf{y i}$ | double-precision array | Imaginary part of $\mathbf{y}$. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z r}$ | double-precision array | Real part of $\mathbf{z}$. |
| $\mathbf{z i}$ | double-precision array | Imaginary part of $\mathbf{z}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxDotProduct

```
int status = CxDotProduct (ComplexNum x[], ComplexNum y[], int n,
    ComplexNum *dotProduct);
```


## Purpose

Calculates the dot product of the complex input arrays $\mathbf{x}$ and $\mathbf{y}$. Use the following formula to obtain the dot product $d$ :

$$
d=\sum_{i=0}^{n-1} x_{i} \times y_{i}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | ComplexNum array | First complex input vector. |
| $\mathbf{y}$ | ComplexNum array | Second complex input vector. |
| $\mathbf{n}$ | integer | Number of elements in each vector. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| dotProduct | ComplexNum | Complex dot product. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```


## CxEigenValueVector

```
int status = CxEigenValueVector (void *A, int n, int matrixType,
    int outputChoice, ComplexNum eigenValues[],
    void *eigenVectors);
```


## Purpose

Calculates the eigenvalues $\lambda$ and the corresponding eigenvectors $\mathbf{x}$ of a complex, square input matrix $\mathbf{A}$. The following formula defines the eigenvalues and the corresponding eigenvectors:

$$
A x=\lambda x
$$

The matrixType parameter indicates the type of the input matrix. The input matrix can be a general or a Hermitian matrix. The outputChoice parameter determines what CxEigenValueVector calculates. Depending on your application, you can choose to calculate just the eigenvalues or to calculate both the eigenvalues and the eigenvectors. The eigenValues output parameter is a 1 D array of $\mathbf{n}$ complex numbers. The eigenVectors output parameter is an $\mathbf{n}$-by-n, complex matrix ( 2 D array) that contains the eigenvectors of the input matrix. Each $i^{t h}$ column of this matrix is the eigenvector that corresponds to the $i^{\text {th }}$ component of the eigenValues. Each eigenvector is normalized so that its largest component equals one.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| A | ComplexNum 2D array | Input complex, square matrix. |
| $\mathbf{n}$ | integer | Number of elements in one dimension of <br> the matrix. |
| matrixType | integer | Pass 0 for general matrix; 1 for Hermitian <br> matrix. Choose the matrix type correctly <br> because it significantly affects the speed of <br> computation. |
| outputChoice | integer | Pass 0 for eigenvalues only; 1 for both <br> eigenvalues and eigenvectors. |

Output

| Name | Type | Description |
| :--- | :--- | :--- |
| eigenValues | ComplexNum array | Result eigenvalues of the input matrix. |
| eigenVectors | ComplexNum 2D array | Result eigenvectors of the input matrix. <br> You can pass NULL if outputChoice is 0. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```


## CxExp

```
int status = CxExp (double xr, double xi, double *yr, double *yi);
```


## Purpose

Calculates the exponential of a complex number, $x$. CxExp obtains the resulting complex number, $y$, using the following formula:

$$
(y r, y i)=e^{(x r, x i)}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision | Real part of $x$. |
| $\mathbf{x i}$ | double-precision | Imaginary part of $x$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y r}$ | double-precision | Real part of $y$. |
| $\mathbf{y i}$ | double-precision | Imaginary part of $y$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxGenInvMatrix

```
int status = CxGenInvMatrix (void *A, int n, int matrixType, void *B);
```


## Purpose

Calculates the inverse of a complex, square input matrix $\mathbf{A}$. If $\mathbf{B}$ denotes the inverse of the matrix A:
$A B=I \quad$ where $I$ is the identity matrix
The input matrix can be an upper or lower triangular matrix, a general, square matrix, or a positive definite matrix. You can save significant computation time if you properly specify the input matrix type.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| A | ComplexNum 2D array | Input complex, square matrix. |
| $\mathbf{n}$ | integer | Number of elements in one dimension of <br> the matrix. |
| matrixType | integer | Type of the input matrix. Choose the matrix <br> type correctly because it significantly <br> affects the speed of computation. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| B | ComplexNum 2D array | Result complex inverse matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
double real;
double imaginary;
    } ComplexNum;
```

Table 2-23 shows valid matrix type values.
Table 2-23. Valid Matrix Type Values

| Matrix Type | Value |
| :--- | :---: |
| General matrix | 0 |
| Positive definite | 1 |
| Upper triangular | 2 |
| Lower triangular | 3 |

## CxGenLinEqs

```
int status = CxGenLinEqs (void *A, int n, int m, ComplexNum y[],
    int matrixType, ComplexNum x[]);
```


## Purpose

Solves for the unknown vector $\mathbf{x}$ in the linear system of equations:

$$
\begin{equation*}
A x=y \tag{2-1}
\end{equation*}
$$

where $A$ is the complex input matrix $y$ is the known vector on the right side

The input matrix can be square or rectangular. The number of elements in $\mathbf{y}$ must equal the number of rows in the matrix $\mathbf{A}$.

CxGenLinEqs calculates the solution using the Singular Value Decomposition technique.
In the case of non-singular, square matrices, in which no row or column is a linear combination of any other row or column, CxGenLinEqs solves for the unique solution $\mathbf{x}$.

Two possibilities exist in the case of rectangular matrices. If the number of rows is greater than the number of columns, the system has more equations than unknowns and is an overdetermined system. Because the solution that satisfies Equation (2-1) might not exist, this procedure finds the least square solution $\mathbf{x}$, which minimizes $\|A\|_{2}$. If the number of rows is less than the number of columns, the system has more unknowns than equations and is an underdetermined system. It might have infinite solutions that satisfy Equation (2-1). This procedure calculates the minimum 2-norm solution.

If the input matrix is rank deficient, CxGenLinEqs returns a warning.
The matrixType parameter specifies the type of the input matrix. The input matrix can be an upper or lower triangular matrix, a general matrix, or a positive definite matrix.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | ComplexNum 2D array | Input complex matrix. The matrix can be <br> square or rectangular. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in $\mathbf{A}$. |


| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | ComplexNum array | Complex array that contains the set of <br> known vector coefficients. |
| matrixType | integer | Type of the input matrix. Choose the matrix <br> type correctly because it significantly <br> affects the speed of computation. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | ComplexNum array | Solution to the linear system of equations. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |
|  |  | If the input matrix is rank-deficient, |
|  |  | cxGenLinEqs returns the warning |
|  |  | code 20001. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```

Table 2-24 shows valid matrix type values.
Table 2-24. Valid Matrix Type Values

| Matrix Type | Value |
| :--- | :---: |
| General matrix | 0 |
| Positive definite | 1 |
| Upper triangular | 2 |
| Lower triangular | 3 |

## CxLinEv1D

```
int status = CxLinEv1D (double xr[], double xi[], int n, double ar,
    double ai, double br, double bi, double yr[],
    double yi[]);
```


## Purpose

Performs a complex linear evaluation of a 1D complex array, $\mathbf{x}$ and $\mathbf{y}$. CxLinEv1D obtains the $i^{\text {th }}$ element of the resulting complex array, $\mathbf{z}$, using the following formulas:

$$
\begin{aligned}
& y r_{i}=a r \times x r_{i}-a i \times x i_{i}+b r \\
& y i_{i}=a r \times x i_{i}+a i \times x r_{i}+b i
\end{aligned}
$$

CxLinEv1D can perform the operations in place; that is, the input and output complex arrays can be the same.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision array | Real part of $\mathbf{x}$. |
| $\mathbf{x i}$ | double-precision array | Imaginary part of $\mathbf{x}$. |
| $\mathbf{n}$ | integer | Number of elements. |
| $\mathbf{a r}$ | double-precision | Real part of $a$. |
| $\mathbf{a i}$ | double-precision | Imaginary part of $a$. |
| $\mathbf{b r}$ | double-precision | Real part of $b$. |
| $\mathbf{b i}$ | double-precision | Imaginary part of $b$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y r}$ | double-precision array | Real part of $\mathbf{y}$. |
| $\mathbf{y i}$ | double-precision array | Imaginary part of $\mathbf{y}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

```
int status = CxLn (double xr, double xi, double *yr, double *yi);
```


## Purpose

Calculates the natural logarithm of a complex number, $x$. CxLn obtains the resulting complex number, $y$, using the following formula:

$$
(y r, y i)=\log _{e}(x r, x i) \quad \text { where } e=2.178 \ldots
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision | Real part of $x$. |
| $\mathbf{x i}$ | double-precision | Imaginary part of $x$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y r}$ | double-precision | Real part of $y$. |
| $\mathbf{y i}$ | double-precision | Imaginary part of $y$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxLog

```
int status = CxLog (double xr, double xi, double *yr, double *yi);
```


## Purpose

Calculates the logarithm (base 10) of a complex number, $x$. CxLog obtains the resulting complex number, $y$, using the following formula:

$$
(y r, y i)=\log _{10}(x r, x i)
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision | Real part of $x$. |
| $\mathbf{x i}$ | double-precision | Imaginary part of $x$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y r}$ | double-precision | Real part of $y$. |
| $\mathbf{y i}$ | double-precision | Imaginary part of $y$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxLU

```
int status = CxLU (void *A, int n, int p[], int *sign);
```


## Purpose

Performs an LU decomposition on the complex, square matrix $\mathbf{A}$ :

$$
P A=L U
$$

where $\quad L$ is an $\mathbf{n}$-by-n lower triangular matrix with all diagonal elements equal to one $U$ is an upper triangular matrix
$P$ is an identity matrix with some rows exchanged based on the information in the permutation vector $\mathbf{p}$

On output, the $\mathbf{U}$ matrix occupies the upper triangular part of the input matrix, including the diagonal elements, and the $\mathbf{L}$ matrix occupies the lower part.

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | ComplexNum 2D array | Square matrix to factorize. |
| $\mathbf{n}$ | integer | Number of elements in one dimension of <br> the matrix. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | ComplexNum 2D array | LU factorized matrix. |
| $\mathbf{p}$ | integer array | Permutation vector. |
| sign | integer | Row exchange indicator. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

After CxLU executes, $\mathbf{p}$ contains possible row exchange information. sign helps calculate the determinant. sign $=0$ indicates that there is no such exchange or that there is an even number of such exchanges. sign $=1$ indicates that there is an odd number of such exchanges.

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```


## CxMatrixMul

```
int status = CxMatrixMul (void *X, void *Y, int n, int k, int m, void *Z);
```


## Purpose

Multiplies two matrices. Use the following formula to obtain the output matrix:

$$
z_{i, j}=\sum_{p=0}^{k-1} x_{i, p} \times y_{p, j} \quad \text { for } i=0,1,2, \ldots, n-1 ; j=0,1,2, \ldots, m-1
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{X}$ | ComplexNum 2D array | First matrix to multiply. |
| $\mathbf{Y}$ | ComplexNum 2D array | Second matrix to multiply. |
| $\mathbf{n}$ | integer | Number of rows in $\mathbf{X}$. |
| $\mathbf{k}$ | integer | Number of columns in $\mathbf{X}$, and number of <br> rows in $\mathbf{Y}$. |
| $\mathbf{m}$ | integer | Number of columns in $\mathbf{Y}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{Z}$ | ComplexNum 2D array | Result of the matrix multiplication. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```

Confirm that the matrix sizes are valid for matrix multiplication. You must meet the following size constraints:

- Input matrix $\mathbf{X}$ must be $\mathbf{n}$ by $\mathbf{k}$.
- Input matrix $\mathbf{Y}$ must be $\mathbf{k}$ by $\mathbf{m}$.
- Output matrix $\mathbf{Z}$ must be $\mathbf{n}$ by $\mathbf{m}$.


## CxMatrixNorm

```
int status = CxMatrixNorm (void *A, int n, int m, int normType, double *norm);
```


## Purpose

Calculates the norm of the complex input matrix $\mathbf{A}$. The input matrix can be square or rectangular. The norm of a matrix is a scalar that gives some measure of the size of the elements in the matrix. It is similar to the concept of magnitude or absolute value for scalar numbers.

There are different ways to calculate the norm of a matrix. The normType parameter indicates which type of norm to use to calculate the norm.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | ComplexNum 2D array | Input complex matrix. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in A. |
| normType | integer | Type of norm to calculate. Refer to the <br> following Parameter Discussion section. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| norm | double-precision | Calculated norm of the input matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```

The normType parameter indicates what type of norm to use to calculate the condition number. Table 2-25 shows valid norm type values.

Table 2-25. Valid Norm Type Values

| Norm Type | Value | Meaning |
| :--- | :---: | :--- |
| 2-norm | 0 | Largest singular value of $\mathbf{A}$. |
| 1-norm | 1 | Largest column sum of $\mathbf{A}$. |
| Frobenius-norm | 2 | Square root of the sum of the diagonal elements of $\mathbf{A}^{\mathbf{T}} \mathbf{A}$, <br> where $\mathbf{A}^{\mathbf{T}}$ is the complex conjugate transpose of $\mathbf{A}$. |
| Infinite-norm | 3 | Largest row sum of $\mathbf{A}$. |

## CxMatrixRank

```
int status = CxMatrixRank (void *A, int n, int m, double tolerance,
    int *rank);
```


## Purpose

Calculates the rank of the complex input matrix $\mathbf{A}$. The input matrix can be square or rectangular.

The maximum number of linearly independent rows or columns of the matrix defines the rank of a matrix. The rank is always less than or equal to the minimum of the number of rows and columns of the matrix. If the rank equals this minimum value, the matrix is a full-rank matrix. Otherwise, it is a rank-deficient matrix.

The rank of a matrix can be calculated in a number of ways. CxMatrixRank first calculates the singular values of the input matrix and then calculates the rank as the number of singular values of the input matrix that are larger than the input tolerance.

You must specify the input tolerance as a positive number close to machine precision. If the matrix in your application is a full-rank matrix, any small value of tolerance gives the same rank. If the matrix in your application is a rank-deficient matrix, different values of tolerance can result in different values of rank.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | ComplexNum 2D array | Input complex matrix. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in A. |
| tolerance | double-precision | Tolerance value. Refer to the following <br> Parameter Discussion section. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| rank | integer | Rank of the input matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```

Matrix rank is the number of singular values in the input matrix that are larger than the tolerance. Set tolerance close to eps, which is the smallest possible double-precision, floating-point number.

## CxMuI

```
int status = CxMul (double xr, double xi, double yr, double yi, double *zr,
    double *zi);
```


## Purpose

Multiplies two complex numbers, $x$ and $y$. CxMul obtains the resulting complex number, $z$, using the following formulas:

$$
\begin{aligned}
& z r=x r \times y r-x i \times y i \\
& z i=x r \times y i+x i \times y r
\end{aligned}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision | Real part of $x$. |
| $\mathbf{x i}$ | double-precision | Imaginary part of $x$. |
| $\mathbf{y r}$ | double-precision | Real part of $y$. |
| $\mathbf{y i}$ | double-precision | Imaginary part of $y$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z r}$ | double-precision | Real part of $z$. |
| $\mathbf{z i}$ | double-precision | Imaginary part of $z$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxMul1D

```
int status = CxMul1D (double xr[], double xi[], double yr[], double yi[],
    int n, double zr[], double zi[]);
```


## Purpose

Multiplies two 1D complex arrays, $\mathbf{x}$ and $\mathbf{y}$. CxMul1D obtains the $i^{\text {th }}$ element of the resulting complex array, $\mathbf{z}$, using the following formulas:

$$
\begin{aligned}
& z r_{i}=x r_{i} \times y r_{i}-x i_{i} \times y i_{i} \\
& z i_{i}=x r_{i} \times y i_{i}+x i_{i} \times y r_{i}
\end{aligned}
$$

CxMul1D can perform the operations in place; that is, the input and output complex arrays can be the same.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision array | Real part of $\mathbf{x}$. |
| $\mathbf{x i}$ | double-precision array | Imaginary part of $\mathbf{x}$. |
| $\mathbf{y r}$ | double-precision array | Real part of $\mathbf{y}$. |
| $\mathbf{y i}$ | double-precision array | Imaginary part of $\mathbf{y}$. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z r}$ | double-precision array | Real part of $\mathbf{z}$. |
| $\mathbf{z i}$ | double-precision array | Imaginary part of $\mathbf{z}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxOuterProduct

```
int status = CxOuterProduct (ComplexNum x[], int nx, ComplexNum y[], int ny,
    void *outerProduct);
```


## Purpose

Calculates the outer product of the complex input vectors $\mathbf{x}$ and $\mathbf{y}$.
Let $x_{i}$ represent the elements of the nx-element vector $\mathbf{x}$ for $i=0,1,2, \ldots, n x-1$.
Let $y_{j}$ represent the elements of the ny-element vector $\mathbf{y}$ for $j=0,1,2, \ldots, n y-1$.
The outer product of these two vectors is a matrix $\mathbf{O}$ of dimensions $\mathbf{n}$-by- $\mathbf{m}$, where the $(i, j)^{\text {th }}$ element of $\mathbf{O}$ is given by:

$$
o_{i, j}=x_{i} \times y_{j}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | ComplexNum array | Input complex vector $\mathbf{x}$. |
| $\mathbf{n x}$ | integer | Number of elements in $\mathbf{x .}$ |
| $\mathbf{y}$ | ComplexNum array | Input complex vector $\mathbf{y}$. |
| $\mathbf{n y}$ | integer | Number of elements in $\mathbf{y}$. |

## Output

| Name | Type | Description |
| :---: | :---: | :--- |
| outerProduct | ComplexNum 2D array | Calculated outer product matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```


## CxPolyRoots

```
int status = CxPolyRoots (double x[], int n, ComplexNum y[]);
```


## Purpose

Calculates the roots of a polynomial. The polynomial coefficients must be real numbers. The roots can be complex or real. The number of roots of the polynomial equals one less than the number of coefficients of the polynomial. Consider the following example:

$$
P(x)=x^{2}+7 x+10
$$

For this example, the elements of the input array $\mathbf{x}$ are $[1,7,10]$. The parameter $\mathbf{n}$ represents the number of coefficients, which is three. The output parameter $\mathbf{y}$ contains the roots of this polynomial, which are $[-5,-2]$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Array of polynomial coefficients. The <br> highest order coefficient is the first element <br> in the array. |
| $\mathbf{n}$ | integer | Number of coefficients in $\mathbf{x}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | ComplexNum array | Array of polynomial roots. Contains $\mathbf{n}-1$ <br> elements. The roots can be complex. Real <br> roots have a zero imaginary part. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```


## CxPow

```
int status = CxPow (double xr, double xi, double a, double *yr, double *yi);
```


## Purpose

Calculates the power of a complex number, $x$. CxPow obtains the resulting complex number, $y$, using the following formula:

$$
(y r, y i)=(x r, x i)^{a}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision | Real part of $x$. |
| $\mathbf{x i}$ | double-precision | Imaginary part of $x$. |
| $\mathbf{a}$ | double-precision | Exponent. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y r}$ | double-precision | Real part of $y$. |
| $\mathbf{y i}$ | double-precision | Imaginary part of $y$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxPseudoInverse

```
int status = CxPseudoInverse (void *A, int n, int m, double tolerance,
    void *B);
```


## Purpose

Calculates the generalized inverse of the complex input matrix $\mathbf{A}$. The input matrix can be square or rectangular. The dimensions of the input matrix $\mathbf{A}$ are $\mathbf{n}-$ by-m. The dimensions of the output matrix (inverse) B are m-by-n.

Note In the case of rectangular matrices with $\mathbf{n}<\mathbf{m}$ (number of rows less than number of columns), take the complex conjugate transpose of the input matrix before you pass it to CxPseudoInverse. The actual pseudoinverse is then the complex conjugate transpose of the matrix calculated by CxP seudo Inverse.

CxPseudoInverse uses the Singular Value Decomposition (SVD) technique. Define the pseudoinverse of a scalar $s$ to be $1 / s$ if $s$ does not equal zero, and zero otherwise. Similarly, define the pseudoinverse of a diagonal matrix by transposing the matrix and then taking the scalar pseudoinverse of each entry. If $A^{\dagger}$ denotes the pseudoinverse of a matrix $\mathbf{A}$ whose singular value decomposition is given by:

$$
A=U S V^{T}
$$

then:

$$
A^{\dagger}=U S^{\dagger} V^{T}
$$

where $S^{\dagger}$ is the pseudoinverse of the diagonal matrix $S$ that contains the singular values of $A$
The pseudoinverse exists for square and rectangular matrices. If the input matrix is square and nonsingular, the pseudoinverse is the same as the general matrix inverse.
$\square$ Note Do not use CxPseudo Inverse to calculate the inverse of a square matrix because it takes more time. Use CxGenInvMatrix instead.

The tolerance parameter must be a small positive number close to machine precision. CxPseudoInverse sets all singular values of the input matrix smaller than tolerance equal to zero.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | ComplexNum 2D array | Input complex matrix. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in A. |
| tolerance | double-precision | Tolerance value. Refer to the following <br> Parameter Discussion section. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| B | ComplexNum 2D array | Calculated pseudoinverse matrix. <br> It is $\mathbf{m}$-by-. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```

The value of tolerance determines the level of accuracy in your final solution. Set tolerance close to eps, which is the smallest possible double-precision, floating-point number.

## CxQR

```
int status = CxQR (void *A, int n, int m, void *Q, void *R);
```


## Purpose

Calculates the QR factorization of the complex input matrix $\mathbf{A}$. The input matrix can be square or rectangular.

The following formula defines the QR factorization of a $\mathbf{n}$-by-m matrix $\mathbf{A}$ such that:

$$
A=Q R
$$

where $\quad Q$ is an orthogonal matrix of dimensions $\mathbf{n}$-by-n
$R$ is an upper triangular matrix of dimensions $\mathbf{n}$-by-m
In general, QR factorization can be calculated in many different ways. In CxQR , QR factorization uses the Householder algorithm. You can use QR factorization to solve linear systems with more equations than unknowns.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | ComplexNum 2D array | Input complex matrix. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in $\mathbf{A}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{Q}$ | ComplexNum 2D array | Calculated orthogonal matrix of the <br> QR factorization. |
| $\mathbf{R}$ | ComplexNum 2D array | Calculated upper triangular matrix of the <br> QR factorization. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
```

    \} ComplexNum;
    
## CxRecip

```
int status = CxRecip (double xr, double xi, double *yr, double *yi);
```


## Purpose

Finds the reciprocal of a complex number, $x$. CxRecip obtains the resulting complex number, $y$, using the following formulas:

$$
\begin{aligned}
& y r=\frac{x r}{x r^{2}+x i^{2}} \\
& y i=\frac{-x i}{x r^{2}+x i^{2}}
\end{aligned}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision | Real part of $x$. |
| $\mathbf{x i}$ | double-precision | Imaginary part of $x$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y r}$ | double-precision | Real part of $y$. |
| $\mathbf{y i}$ | double-precision | Imaginary part of $y$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxSpecialMatrix

```
int status = CxSpecialMatrix (int matrixType, int m, ComplexNum x[], int nx,
    ComplexNum y[], int ny, void *Z);
```


## Purpose

Generates a special type of complex matrix depending on the value of matrixType. There are five possible matrix types: Identity, Diagonal, Toeplitz, Vandermonde, and Companion. Table 2-26 shows each matrix type and its behavior.

Table 2-26. Matrix Types and Behaviors

| Matrix Type | Behavior |
| :--- | :--- |
| Identity | CxSpecialMatrix generates an $\mathbf{m}$-by-m identity matrix. |$|$| CxSpecialMatrix generates an $\mathbf{n x}$-by-nx diagonal matrix with |
| :--- |
| diagonal elements that are the elements of $\mathbf{x}$. |

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| matrixType | integer | Type of matrix to generate. Refer to the <br> following Parameter Discussion section. |
| $\mathbf{m}$ | integer | Number of rows and columns to generate <br> when matrixType is Identity matrix. |
| $\mathbf{x}$ | ComplexNum array | Complex vector used to generate a Diagonal <br> matrix, Toeplitz matrix, Vandermonde <br> matrix, or Companion matrix. |
| $\mathbf{n x}$ | integer | Number of elements in vector $\mathbf{x}$. |
| $\mathbf{y}$ | ComplexNum array | Second vector to use to generate the <br> Toeplitz matrix. |
| $\mathbf{n y}$ | integer | Number of elements in vector $\mathbf{y}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{Z}$ | integer | Generated matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```

Table 2-27 shows valid matrix type values.
Table 2-27. Valid Matrix Type Values

| Matrix Type | Value |
| :--- | :---: |
| Identity matrix | 0 |
| Diagonal matrix | 1 |
| Toeplitz matrix | 2 |
| Vandermonde matrix | 3 |
| Companion matrix | 4 |

## CxSqrt

```
int status = CxSqrt (double xr, double xi, double *yr, double *yi);
```


## Purpose

Calculates the square root of a complex number, $x$. CxSqrt obtains the resulting complex number, $y$, using the following formula:

$$
(y r, y i)=(x r, x i)^{1 / 2}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision | Real part of $x$. |
| $\mathbf{x i}$ | double-precision | Imaginary part of $x$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y r}$ | double-precision | Real part of $y$. |
| $\mathbf{y i}$ | double-precision | Imaginary part of $y$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxSub

```
int status = CxSub (double xr, double xi, double yr, double yi, double *zr,
    double *zi);
```


## Purpose

Subtracts two complex numbers, $x$ and $y$. The resulting complex number, $z$, is obtained using the following formulas:

$$
\begin{aligned}
z r & =x r-y r \\
z i & =x i-y i
\end{aligned}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision | Real part of $x$. |
| $\mathbf{x i}$ | double-precision | Imaginary part of $x$. |
| $\mathbf{y r}$ | double-precision | Real part of $y$. |
| $\mathbf{y i}$ | double-precision | Imaginary part of $y$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z r}$ | double-precision | Real part of $z$. |
| $\mathbf{z i}$ | double-precision | Imaginary part of $z$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxSub1D

```
int status = CxSub1D (double xr[], double xi[], double yr[], double yi[],
    int n, double zr[], double zi[]);
```


## Purpose

Subtracts two 1D complex arrays, $\mathbf{x}$ and $\mathbf{y}$. CxSub1D obtains the $i^{\text {th }}$ element of the resulting complex array, $\mathbf{z}$, using the formulas:

$$
\begin{aligned}
z r_{i} & =x r_{i}-y r_{i} \\
z i_{i} & =x i_{i}-y i_{i}
\end{aligned}
$$

CxSub1D can perform the operations in place; that is, the input and output complex arrays can be the same.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x r}$ | double-precision array | Real part of $\mathbf{x}$. |
| $\mathbf{x i}$ | double-precision array | Imaginary part of $\mathbf{x}$. |
| $\mathbf{y r}$ | double-precision array | Real part of $\mathbf{y .}$ |
| $\mathbf{y i}$ | double-precision array | Imaginary part of $\mathbf{y}$. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z r}$ | double-precision array | Real part of $\mathbf{z}$. |
| $\mathbf{z i}$ | double-precision array | Imaginary part of $\mathbf{z}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## CxSVD

```
int status = CxSVD (void *A, int n, int m, void *U, ComplexNum s[], void *V);
```


## Purpose

Calculates the Singular Value Decomposition (SVD) factorization of the complex input matrix $\mathbf{A}$. The input matrix can be square or rectangular.

The following formula defines the SVD factorization of an $\mathbf{n}-$ by $-\mathbf{m}$ matrix $\mathbf{A}$ :

$$
A=U S V^{H}
$$

where $\quad U$ is an orthogonal matrix of dimensions $\mathbf{n}$-by-m
$V$ is an orthogonal matrix of dimensions $\mathbf{m}$-by-m
$S$ is a diagonal matrix of dimensions $\mathbf{m}$-by-m
$V^{H}$ represents the complex conjugate transpose of $\mathbf{V}$. The diagonal elements of $S$ are called the singular values of $\mathbf{A}$ and are arranged in descending order. CxSVD stores the diagonal elements of $S$ in the output array $\mathbf{s}$.

The Singular Value Decomposition is an eigenvalue-like decomposition for rectangular matrices. You can use it to calculate the condition number of a matrix or to solve linear, least square problems. SVD is useful for ill-conditioned or rank-deficient problems because it can detect small singular values.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | ComplexNum 2D array | Input complex matrix. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in $\mathbf{A}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{U}$ | ComplexNum 2D array | The $\mathbf{n}$-by-m orthogonal matrix <br> SVD factorization generates. |
| $\mathbf{s}$ | ComplexNum array | Array that contains the singular values <br> of $\mathbf{A}$, in descending order. |
| $\mathbf{V}$ | ComplexNum 2D array | The $\mathbf{m}$-by-m orthogonal matrix <br> SVD factorization generates. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```


## CxSVDS

```
int status = CxSVDS (void *A, int n, int m, ComplexNum s[]);
```


## Purpose

CxSVDS is similar to CxSVD, but it calculates only the singular values that result from the Singular Value Decomposition factorization of the complex input matrix. The input matrix can be square or rectangular.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | ComplexNum 2D array | Input complex matrix. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in $\mathbf{A}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{s}$ | ComplexNum array | Array that contains the singular values of $\mathbf{A}$, <br> in descending order. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```


## CxTrace

```
int status = CxTrace (void *A, int n, int m, ComplexNum *trace);
```


## Purpose

Calculates the trace of a complex matrix. The trace of a matrix is the sum of all its diagonal elements.

CxTrace uses the following formula to obtain trace $t$ :

$$
t=\sum_{i=0}^{k-1} a_{i, i} \quad \text { where } k=\min (n, m)
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | ComplexNum 2D array | Input complex matrix. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in $\mathbf{A}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| trace | ComplexNum | Sum of the diagonal elements of $\mathbf{A}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
```

    double real;
    double imaginary;
    \} ComplexNum;
    
## CxTranspose

```
int status = CxTranspose (void *A, int n, int m, void *B);
```


## Purpose

Calculates the complex conjugate transpose of a 2D, complex input matrix. The following formula defines the $(i, j)^{t h}$ element of the resulting matrix:

$$
b_{i, j}=a_{j, i}^{*} \quad \text { where } * \text { denotes a complex conjugate }
$$

If $z=x+j \times y$ is a complex number, then $x-j \times y$ is the complex conjugate of $z$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | ComplexNum 2D array | Input complex matrix. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in $\mathbf{A}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| B | ComplexNum | Calculated complex conjugate transpose <br> matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
```

    double real;
    double imaginary;
    \} ComplexNum;
    
## Decimate

```
int status = Decimate (double x[], int n, int dFact, int ave, double y[]);
```


## Purpose

Decimates the input sequence $\mathbf{x}$ by the decimating factor. If $\mathbf{y}$ represents the decimated output sequence, Decimate obtains the elements of the sequence $y$ using the following formula:

$$
y_{i}= \begin{cases}x_{i \times d F a c t} & \text { ave }=0 \\ \frac{1}{d F a c t} \sum_{k=0}^{d F a c t-1} x_{i \times d F a c t+k} & \text { ave }=1\end{cases}
$$

where $i=0,1,2, \ldots$, size -1
size $=\operatorname{int}(n / d F a c t)$ is the size of the output sequence

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Contains the input array to decimate. |
| $\mathbf{n}$ | integer | Number of elements in the input array. |
| $\mathbf{d F a c t}$ | integer | Amount by which to decimate $\mathbf{x}$ to form $\mathbf{y}$. |
| ave | integer | Specifies whether to use averaging in <br> decimating $\mathbf{x}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Contains the output array, which is $\mathbf{x}$ <br> decimated by the dFact. The size of this <br> array must be int $(\mathbf{n} / \mathbf{d F a c t})$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Deconvolve

```
int status = Deconvolve (double y[], int ny, double x[], int nx, double h[]);
```


## Purpose

Calculates the deconvolution of $\mathbf{y}$ with $\mathbf{x}$. Deconvolve assumes $\mathbf{y}$ to be the result of the convolution of $\mathbf{x}$ with some system response. The function realizes the deconvolution operation using Fourier transform pairs. Deconvolve obtains the output sequence $\mathbf{h}$ using the following formula:

$$
h=\operatorname{InvFFT}\left(\frac{Y(f)}{X(f)}\right)
$$

where $X(f)$ is the Fourier transform of $\mathbf{x}$
$Y(f)$ is the Fourier transform of $\mathbf{y}$
InvFFT( ) is the inverse Fourier transform

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Input array to deconvolve with $\mathbf{x}$. |
| $\mathbf{n y}$ | integer | Number of elements in $\mathbf{y}$. |
| $\mathbf{x}$ | double-precision array | Input array with which to deconvolve $\mathbf{y}$. |
| $\mathbf{n x}$ | integer | Number of elements in $\mathbf{x}$. <br> $\mathbf{n x} \leq \mathbf{n y}$ |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{h}$ | double-precision array | Output array that is $\mathbf{y}$ deconvolved with $\mathbf{x}$. <br> This array must be $(\mathbf{n y}-\mathbf{n x}+1)$ <br> elements long. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Determinant

```
int status = Determinant (void *x, int n, double *det);
```


## Purpose

Finds the determinant of an $\mathbf{n}$-by-n 2D input matrix.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision <br> 2D array | Input matrix. |
| $\mathbf{n}$ | integer | Dimension size of input matrix. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| det | double-precision | Determinant. |

Note $\quad$ The input matrix must be an n-by-n square matrix.

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Difference

```
int status = Difference (double x[], int n, double dt, double xInit,
    double xFinal, double y[]);
```


## Purpose

Finds the discrete difference of the input array. Difference obtains the $i^{\text {th }}$ element of the resulting array using the following formula:

$$
y_{i}=\frac{x_{i+1}-x_{i-1}}{2 d t} \quad \text { where } x_{-1}=x \text { Int and } x_{n}=x \text { Final }
$$

Difference can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{d t}$ | double-precision | Sampling interval. |
| $\mathbf{x I n i t}$ | double-precision | Initial condition. |
| xFinal | double-precision | Final condition. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Differentiated array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* Generate an array with random numbers and differentiate it. */ double x[200], y[200];

```
double dt, xInit, xFinal;
```

    int \(n\);
    \(\mathrm{n}=200\);
    \(d t=0.001 ;\)
    xInit \(=-0.5\);
    xFinal \(=-0.25\);
    Uniform ( \(\mathrm{n}, 17\), x );
    Integrate ( \(x, n, d t, x I n i t, x F i n a l, y)\);
    
## Div1D

```
int status = Div1D (double x[], double y[], int n, double z[]);
```


## Purpose

Divides two 1D arrays, $\mathbf{x}$ and $\mathbf{y}$. Div1D obtains the $i^{\text {th }}$ element of the output array, $\mathbf{z}$, using the following formula:

$$
z_{i}=\frac{x_{i}}{y_{i}}
$$

Div1D can perform the operation in place; that is, $\mathbf{z}$ can be the same array as either $\mathbf{x}$ or $\mathbf{y}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | $\mathbf{x}$ input array. |
| $\mathbf{y}$ | double-precision array | $\mathbf{y}$ input array. |
| $\mathbf{n}$ | integer | Number of elements to divide. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z}$ | double-precision array | Result array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Div2D

```
int status = Div2D (void *x, void *y, int n, int m, void *z);
```


## Purpose

Divides two 2D arrays. Div2D obtains the $(i, j)^{\text {th }}$ element of the output array using the following formula:

$$
z_{i, j}=\frac{x_{i, j}}{y_{i, j}}
$$

Div2D can perform the operation in place; that is, $\mathbf{z}$ can be the same array as either $\mathbf{x}$ or $\mathbf{y}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision <br> 2D array | $\mathbf{x}$ input array. |
| $\mathbf{y}$ | double-precision <br> 2D array | $\mathbf{y}$ input array. |
| $\mathbf{n}$ | integer | Number of elements in first dimension. |
| $\mathbf{m}$ | integer | Number of elements in second dimension. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z}$ | double-precision <br> 2D array | Result array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## DotProduct

```
int status = DotProduct (double x[], double y, int n, double *dotProd);
```


## Purpose

Calculates the dot product of the $\mathbf{x}$ and $\mathbf{y}$ input arrays. DotProduct obtains the dot product using the following formula:

$$
\text { dotProd }=x \bullet y=\sum_{i=0}^{n-1} x_{i} \times y_{i}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | $\mathbf{x}$ input vector. |
| $\mathbf{y}$ | double-precision array | $\mathbf{y}$ input vector. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| dotProd | double-precision | Dot product. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Elp_BPF

```
int status = Elp_BPF (double x[], int n, double fs, double fl, double fh,
    double ripple, double atten, int order,
    double y[]);
```


## Purpose

Filters the input array using a digital bandpass elliptic filter. Elp_BPF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f l}$ | double-precision | Lower cutoff frequency. |
| $\mathbf{f h}$ | double-precision | Higher cutoff frequency. |
| ripple | double-precision | Pass band ripples in decibels. |
| atten | double-precision | Stop band attenuation in decibels. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random signal and filter it using a fifth-order bandpass
elliptic filter. The pass band is from 200.0 to 300.0. */
double x[256], y[256], fs, fl, fh, ripple, atten;
int n, order;
n = 256;
fs = 1000.0;
fl = 200.0;
fh = 300.0;
ripple = 0.5;
atten = 40.0;
order = 5;
Uniform (n, 17, x);
Elp_BPF (x, n, fs, fl, fh, ripple, atten, order, y);
```


## Elp_BSF

```
int status = Elp_BSF (double x[], int n, double fs, double fl, double fh,
    double ripple, double atten, int order,
    double y[]);
```


## Purpose

Filters the input array using a digital bandstop elliptic filter. Elp_BSF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f l}$ | double-precision | Lower cutoff frequency. |
| $\mathbf{f h}$ | double-precision | Higher cutoff frequency. |
| ripple | double-precision | Pass band ripples in decibels. |
| atten | double-precision | Stop band attenuation in decibels. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random signal and filter it using a fifth-order bandstop
elliptic filter. The stop band is from 200.0 to 300.0. */
double x[256], y[256], fs, fl, fh, ripple, atten;
int n, order;
n = 256;
fs = 1000.0;
fl = 200.0;
fh = 300.0;
ripple = 0.5;
atten = 40.0;
order = 5;
Uniform (n, 17, x);
Elp_BSF (x, n, fs, fl, fh, ripple, atten, order, y);
```


## Elp_CascadeCoef

```
int status = Elp_CascadeCoef (double fs, double fl, double fh, double ripple,
    double atten, IIRFilterPtr filterInformation);
```


## Purpose

Generates the set of cascade form filter coefficients to implement an IIR filter as specified by the elliptic (or Cauer) filter model.
filterInformation is the pointer to the filter structure that contains the filter coefficients and the internal filter information. You must allocate this structure by calling AllocIIRFilterPtr before calling this cascade IIR filter design function.

To redesign another filter, you should first call FreeIIRFilterPtr to free the present filter structure and then call AllocIIRFilterPtr with the new type and order parameters before you call Elp_CascadeCoef.

If the type and order remain the same, you can call Elp_CascadeCoef without calling FreeIIRFilterPtr and AllocIIRFilterPtr. In this case, you should properly reset the filtering operation for that structure by calling ResetIIRFilter before the first call to IIRCascadeFiltering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| fs | double-precision | Specifies the sampling frequency in hertz. |
| fl | double-precision | Specifies the desired lower cutoff frequency <br> of the filter in hertz. |
| fh | double-precision | Specifies the desired upper cutoff frequency <br> of the filter in hertz |
| ripple | double-precision | Specifies the amplitude of the stop band <br> ripple in decibels. |
| atten | double-precision | Specifies the stop band attenuation, in <br> decibels, of the IIR filter to design. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| filterInformation | IIRFilterPtr | Pointer to the filter structure that contains <br> the filter coefficients and the internal filter <br> information. |
|  |  | Refer to the AllocI IRFilterPtr <br> function description for more information <br> about the filter structure. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Design a cascade lowpass elliptic IIR filter. */
double fs, fl, fh, ripple, atten, x[256], y[256];
int type, order, n;
IIRFilterPtr filterInfo;
n = 256;
fs = 1000.0;
fl = 200.0;
ripple = 0.5;
atten = 40.0;
order = 5;
type = 0; /* lowpass */
Uniform(n, 17, x);
filterInfo = AllocIIRFilterPtr(type, order);
if(filterInfo!=0) {
    Elp_CascadeCoef(fs, fl, fh, ripple, atten, filterInfo);
    IIRCascadeFiltering(x, n, filterInfo, y);
    FreeIIRFilterPtr(filterInfo);
}
```


## Elp_Coef

```
int status = Elp_Coef (int type, int order, double fs, double fl, double fh,
    double ripple, double atten, double a[], int na,
    double b[], int nb);
```


## Purpose

Generates the set of filter coefficients to implement an IIR filter as specified by the elliptic (or Cauer) filter model. type has the valid values as shown in Table 2-28.

Table 2-28. Valid type Values

| Value | Description |
| :---: | :--- |
| 0 | lowpass filter; fh is not used |
| 1 | highpass filter; fh is not used |
| 2 | bandpass filter |
| 3 | bandstop filter |

$\mathbf{a}$ and $\mathbf{b}$ are the reverse and forward filter coefficients. Use IIRFiltering to achieve the actual filtering:

$$
y_{n}=\frac{1}{a_{0}}\left(\sum_{i=0}^{n b-1} b_{i} x_{n-i}-\sum_{i=1}^{n a-1} a_{i} y_{n-i}\right)
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| type | integer | Controls the filter type of the elliptic <br> IIR filter coefficients. |
| order | integer | Order of the IIR filter. |
| fs | double-precision | Sampling frequency in hertz. |
| fl | double-precision | Desired lower cutoff frequency of the filter <br> in hertz. |
| fh | double-precision | Desired higher cutoff frequency of the filter <br> in hertz. |


| Name | Type | Description |
| :--- | :--- | :--- |
| ripple | double-precision | Amplitude of the stop band ripple <br> in decibels. |
| atten | double-precision | Stop band attenuation, in decibels, of the <br> IIR filter to be designed. |
| na | integer | Number of coefficients in the a coefficient <br> array. |
| nb | integer | Number of coefficients in the $\mathbf{b}$ coefficient <br> array. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| a | double-precision array | Array that contains the reverse coefficients <br> of the designed IIR filter. |
| b | double-precision array | Array that contains the forward coefficients <br> of the designed IIR filter. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Elp_HPF

```
int status = Elp_HPF (double x[], int n, double fs, double fc, double ripple,
    double atten, int order, double y[]);
```


## Purpose

Filters the input array using a digital highpass elliptic filter. Elp_HPF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| fc | double-precision | Cutoff frequency. |
| ripple | double-precision | Pass band ripples in decibels. |
| atten | double-precision | Stop band attenuation in decibels. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random signal and filter it using a fifth-order highpass
elliptic filter. */
double x[256], y[256], fs, fc, ripple, atten;
int n, order;
n = 256;
fs = 1000.0;
fc = 200.0;
ripple = 0.5;
atten = 40.0;
order = 5;
Uniform (n, 17, x);
Elp_HPF (x, n, fs, fc, ripple, atten, order, y);
```


## Elp_LPF

```
int status = Elp_LPF (double x[], int n, double fs, double fc, double ripple,
    double atten, int order, double y[]);
```


## Purpose

Filters the input array using a digital lowpass elliptic filter. Elp_LPF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| fc | double-precision | Cutoff frequency. |
| ripple | double-precision | Pass band ripples in decibels. |
| atten | double-precision | Stop band attenuation in decibels. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random signal and filter it using a fifth-order lowpass
elliptic filter. */
double x[256], y[256], fs, fc, ripple, atten;
int n, order;
n = 256;
fs = 1000.0;
fc = 200.0;
ripple = 0.5;
atten = 40.0;
order = 5;
Uniform (n, 17, x);
Elp_LPF (x, n, fs, fc, ripple, atten, order, y);
```


## Equi_Ripple

```
int status = Equi_Ripple (int bands, double A[], double wts[], double fs,
    double cutoffs[], int type, int n, double coef[],
    double *delta);
```


## Purpose

Designs a multiband FIR linear phase filter, a differentiator, or a Hilbert Transform using the Parks-McClellan algorithm. The frequency response in each band has equal ripples that you can adjust by a weighting factor. Equi_Ripple generates only the filter coefficients; it does not actually perform data filtering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| bands | integer | Number of bands of the filter. |
| A | double-precision array | Desired frequency response magnitude of <br> each band. |
| wts | double-precision array | Weighting factor for each band. |
| fs | double-precision | Sampling frequency. |
| cutoffs | double-precision array | End frequencies of each band. |
| type | integer | Filter type. |
| $\mathbf{n}$ | integer | Filter length. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Filter coefficients. |
| delta | double-precision | Normalized ripple size. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

Generally, when type $=1$ and bands $\geq 2$, Equi_Ripple designs a multiband filter. When type $=2$, bands $=1$, and $\mathbf{n}$ is even, Equi_Ripple designs a differentiator. When type $=3$, bands $=1$, and $\mathbf{n}$ is even, Equi_Ripple designs a Hilbert Transform. For more information, refer to Digital Filter Design by Parks and Burrus or "A computer program for designing optimum FIR linear phase digital filters," by McClellan, et al., IEEE Transactions on Audio and Electroacoustics.

## Using This Function

Although Equi_Ripple is the most flexible way to design an FIR linear phase filter, it has more complex parameters and requires some digital signal processing (DSP) knowledge. You might find it more convenient to use EquiRpl_LPF, EquiRpl_HPF, EquiRpl_BPF, and EquiRpl_BSF. These functions, which provide lowpass, highpass, bandpass, and bandstop FIR filters with equal weighting factors in all bands, are special cases of Equi_Ripple with simplified parameters.

For more information about windowing, refer to the About Windowing section in Chapter 1, Advanced Analysis Library Overview.

## Example 1

```
/* Design a 24-point lowpass filter and filter the incoming signal. */
double x[256], coef[24], y[280], fs, delta;
double A[2]; /* array of frequency responses */
double wts[2]; /* array of weighting factors */
double cutoffs[4]; /* frequency points */
int n, m;
int bands; /* number of bands */
int type; /* filter type */
bands = 2; /* one pass band and one stop band */
fs = 1000.0; /* sampling frequency */
A[0] = 1.0; /* 1 for the pass band */
A[1] = 0.0; /* 0 for the stop band */
wts[0] = 1.0; /* weighting factor for the pass band */
wts[1] = 1.0; /* weighting factor for the stop band */
cutoffs[0] = 0.0;
cutoffs[1] = 300.0; /* the first stop band [0, 300.0] */
cutoffs[2] = 400.0;
cutoffs[3] = 500.0; /* the pass band [400, 500] */
type = 1; /* multiple band filter */
n = 24; /* filter length */
m = 256;
Equi_Ripple (bands, A, wts, fs, cutoffs, type, n, coef, &delta);
Convolve (coef, n, x, m, y); /* Convolve the filter with the signal. */
```


## Example 2

/* Design a 31 -point bandpass filter and filter the incoming signal. */
double x[256], coef[55], y[287], fs, delta;
double A[3]; /* array of frequency responses */
double wts[3]; /* array of weighting factors */
double cutoffs[6]; /* frequency points */
int $n, m$;
int bands; /* number of bands */
int type;
/* filter type */
bands $=3$;
$\mathrm{fs}=1000.0$;
/* one pass band and two stop bands */
/* sampling frequency */
$\mathrm{A}[0]=0.0 ; \quad / * 0$ for the first stop band */
$A[1]=1.0 ; \quad / * 1$ for the stop band */
A[2] $=0.0 ; \quad / * 0$ for second stop band */
wts[0] = 10.0; /* weighting factor for the first stop band */
wts [1] = 1.0; /* weighting factor for the pass band */
wts[2] = 4.0; /* weighting factor for the second stop band */
cutoffs[0] $=0.0$;
cutoffs[1] = 200.0; /* the first stop band [0, 200.0] */
cutoffs[2] $=250.0$;
cutoffs [3] $=350.0 ; \quad / *$ the pass band [250, 350] */
cutoffs[4] $=400.0$;
cutoffs[5] = 500.0; /* the second stop band */
type $=1$; /* multiple band filter */
$\mathrm{n}=31$; /* filter length */
$m=256 ;$
Equi_Ripple (bands, A, wts, fs, cutoffs, type, n, coef, \&delta);
Convolve (coef, $n, x, m, y) ; / *$ Convolve the filter with the signal. */

## Example 3

```
/* Design a 30-point differentiator. */
double coef[30], fs, delta;
double A[1]; /* array of frequency responses */
double wts[1]; /* array of weighting factors */
double cutoffs[2]; /* frequency points */
int n;
int bands; /* number of bands */
int type; /* filter type */
bands = 1; /* one pass band and one stop band */
fs = 1000.0; /* sampling frequency */
A[0] = 1.0; /* 1 for the band */
wts[0] = 1.0; /* weighting factor for the band */
cutoffs[0] = 0.0;
cutoffs[1] = 500.0; /* the entire frequency range */
type = 2; /* differentiator */
n = 30; /* filter length */
Equi_Ripple (bands, A, wts, fs, cutoffs, type, n, coef, &delta);
```


## Example 4

```
/* Design a 20-point Hilbert transform. */
double coef[20], fs, delta;
double A[1]; /* array of frequency responses */
double wts[1];
double cutoffs[2];
    /* array of weighting factors */
    /* frequency points */
int n;
int bands;
/* number of bands */
int type;
bands = 1;
fs = 1000.0;
A[0] = 1.0;
wts[0] = 1.0;
cutoffs[0] = 100.0;
cutoffs[1] = 500.0;
type = 3; /* Hilbert transform */
n = 20; /* filter length */
Equi_Ripple (bands, A, wts, fs, cutoffs, type, n, coef, &delta);
```


## EquiRpl_BPF

```
int status = EquiRpl_BPF (double fs, double f1, double f2, double f3,
    double f4, int n, double coef[], double *delta);
```


## Purpose

Designs a bandpass FIR linear phase filter using the Parks-McClellan algorithm. EquiRpl_BPF is a special case of the general Parks-McClellan algorithm. EquiRpl_BPF generates only the filter coefficients; it does not actually perform data filtering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| fs | double-precision | Sampling frequency. |
| f1 | double-precision | Cutoff frequency 1. |
| f2 | double-precision | Cutoff frequency 2. |
| f3 | double-precision | Cutoff frequency 3. |
| f4 | double-precision | Cutoff frequency 4. |
| n | integer | Filter length. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Filter coefficients. |
| delta | double-precision | Normalized ripple size. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

There are two stop bands and one pass band. The first stop band is [0,f1], and the second stop band is $[\mathbf{f 4}, \mathbf{f s} / 2]$. The pass band is $[\mathbf{f} 2, \mathbf{f 3}] . \mathbf{f 1}, \mathbf{f} \mathbf{2}, \mathbf{f 3}$, and $\mathbf{f 4}$ must be in ascending order. Refer to the Equi_Ripple function description for more information.

## Example

/* Design a 51-point bandpass filter and filter the incoming signal. */ double $x[256], \operatorname{coef}[25], y[301], f s, f 1, f 2, f 3, f 4$, delta;
int $n, m ;$
fs = 1000.0; /* sampling frequency */
$\mathrm{f} 1=200.0$;
/* the first stop band [0, 200] */
$\mathrm{f} 2=250.0$;
f3 $=350.0 ; \quad / *$ the pass band $[250,350]$ */
f4 4 400.0; $/ *$ the second stop band [400, 500] */
n = 51; /* filter length */
$m=256 ;$
EquiRpl_BPF (fs, f1, f2, f3, f4, n, coef, \&delta);
Convolve (coef, $n, x, m, y) ; / *$ Convolve the filter with the signal. */

## EquiRpl_BSF

```
int status = EquiRpl_BSF (double fs, double f1, double f2, double f3,
    double f4, int n, double coef[], double *delta);
```


## Purpose

Designs a bandstop FIR linear phase filter using the Parks-McClellan algorithm. EquiRpl_BSF is a special case of the general Parks-McClellan algorithm. EquiRpl_BSF generates only the filter coefficients; it does not actually perform data filtering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| fs | double-precision | Sampling frequency. |
| f1 | double-precision | Cutoff frequency 1. |
| f2 | double-precision | Cutoff frequency 2. |
| f3 | double-precision | Cutoff frequency 3. |
| f4 | double-precision | Cutoff frequency 4. |
| n | integer | Filter length. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Filter coefficients. |
| delta | double-precision | Normalized ripple size. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

There are two pass bands and one stop band. The first pass band is $[0, \mathbf{f 1}]$, and the second pass band is $[\mathbf{f 4}, \mathbf{f s} / 2]$. The stop band is $[\mathbf{f} \mathbf{2}, \mathbf{f 3}] . \mathbf{f 1}, \mathbf{f} \mathbf{2}, \mathbf{f 3}$, and $\mathbf{f 4}$ must be in ascending order. Refer to the Equi_Ripple function description for more information.

## Example

/* Design a 51-point bandstop filter and filter the incoming signal. */ double $x[256], \operatorname{coef}[25], y[301], f s, f 1, f 2, f 3, f 4$, delta;
int $n, m ;$
fs = 1000.0; /* sampling frequency */
$\mathrm{f} 1=200.0$;
/* the first pass band [0, 200] */
$\mathrm{f} 2=250.0$;
f3 = 350.0; /* the stop band [250, 350] */
f4 4 400.0; $\quad / *$ the second pass band [400, 500] */
n = 51; /* filter length */
$m=256 ;$
EquiRpl_BSF (fs, f1, f2, f3, f4, n, coef, \&delta);
Convolve (coef, $n, x, m, y) ; / *$ Convolve the filter with the signal. */

## EquiRpl_HPF

```
int status = EquiRpl_HPF (double fs, double f1, double f2, int n,
    double coef [], double *delta);
```


## Purpose

Designs a highpass FIR linear phase filter using the Parks-McClellan algorithm.
EquiRpl_HPF is a special case of the general Parks-McClellan algorithm. EquiRpl_HPF generates only the filter coefficients; it does not actually perform data filtering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| fs | double-precision | Sampling frequency. |
| f1 | double-precision | Cutoff frequency 1. |
| f2 | double-precision | Cutoff frequency 2. |
| n | integer | Filter length. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Filter coefficients. |
| delta | double-precision | Normalized ripple size. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

There is one stop band and one pass band. The stop band is [0,f1], and the pass band is [ $\mathbf{f 2}$, fs/2]. Refer to the Equi_Ripple function description for more information.

## Example

/* Design a 25-point highpass filter and filter the incoming signal. */ double $x[256], \operatorname{coef}[25], y[281], f s, f 1, f 2$, delta;
int $n, m ;$
fs $=1000.0 ; \quad / *$ sampling frequency */
f1 = 300.0; $/ *$ the stop band $[0,300]$ */
f2 $=400.0 ; \quad / *$ the pass band $[400,500]$ */
n $=25$; /* filter length */
$m=256 ;$
EquiRpl_HPF (fs, f1, f2, n, coef, \&delta);
Convolve (coef, $n, x, m, y) ; / *$ Convolve the filter with the signal. */

## EquiRpl_LPF

```
int status = EquiRpl_LPF (double fs, double f1, double f2, int n,
    double coef[], double *delta);
```


## Purpose

Designs a lowpass FIR linear phase filter using the Parks-McClellan algorithm. EquiRpl_LPF is a special case of the general Parks-McClellan algorithm. EquiRpl_LPF generates only the filter coefficients; it does not actually perform data filtering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| fs | double-precision | Sampling frequency. |
| f1 | double-precision | Cutoff frequency 1. |
| f2 | double-precision | Cutoff frequency 2. |
| n | integer | Filter length. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Filter coefficients. |
| delta | double-precision | Normalized ripple size. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

There is one pass band and one stop band. The pass band is [ $0, \mathbf{f} 1]$, and the stop band is [f2, fs/2]. Refer to the Equi_Ripple function description for more information.

## Example

/* Design a 25-point lowpass filter and filter the incoming signal. */ double x[256], coef[25], y[281], fs, f1, f2, delta;
int $n, m ;$
fs $=1000.0 ; \quad$ /* sampling frequency */
f1 = 300.0; /* the pass band [0, 300] */
f2 = 400.0; /* the stop band [400, 500] */
n = 25; /* filter length */
m = 256;
EquiRpl_LPF (fs, f1, f2, n, coef, \&delta);
Convolve (coef, $n, ~ x, m, y) ; / *$ Convolve the filter with the signal. */

## ExBkmanWin

```
int status = ExBkmanWin (double x[], int n);
```


## Purpose

Applies an exact Blackman window to the input sequence $\mathbf{x}$. If $\mathbf{y}$ represents the output sequence, ExBkmanWin obtains the elements of $\mathbf{y}$ using the following formula:

$$
y_{i}=x_{i}\left(a_{0}-a_{1} \times \cos \left(\frac{2 \pi i}{n}\right)+a_{2} \times \cos \left(\frac{4 \pi i}{n}\right)\right) \quad \text { for } i=0, \ldots, n-1
$$

$$
\text { where } \begin{aligned}
a_{0} & =\frac{7938.0}{18608.0} \\
a_{1} & =\frac{9240.0}{18608.0} \\
a_{2} & =\frac{1430.0}{18608.0}
\end{aligned}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Contains the input signal. |
| $\mathbf{n}$ | integer | Number of elements in the input array. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Contains the signal after ExBkmanWin <br> applies the exact Blackman window. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## ExpFit

```
int status = ExpFit (double x[], double y[], int n, double z[], double *a,
    double *b, double *mse);
```


## Purpose

Finds the coefficient values that best represent the exponential fit of the data points $(\mathbf{x}, \mathbf{y})$ using the least squares method. ExpFit obtains the $i^{\text {th }}$ element of the output array using the following formula:

$$
z_{i}=a e^{b x_{i}}
$$

ExpFit obtains the mean squared error (mse) using the following formula:

$$
m s e=\frac{\sum_{i=0}^{n-1}\left|z_{i}-y_{i}\right|^{2}}{n} \quad \text { where } n \text { is the number of sample points }
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | $\mathbf{x}$ values. |
| $\mathbf{y}$ | double-precision array | $\mathbf{y}$ values. |
| $\mathbf{n}$ | integer | Number of sample points. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z}$ | double-precision array | Best exponential fit. |
| $\mathbf{a}$ | double-precision | Amplitude. |
| $\mathbf{b}$ | double-precision | Exponential constant. |
| $\mathbf{m s e}$ | double-precision | Mean squared error. |

[^0]
## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* Generate an exponential pattern and find the best exponential fit. */
double x[200], y[200], z[200];
double first, last, $a, b, a m p, ~ d e c a y, ~ m s e ; ~$
int $n$;
$\mathrm{n}=200$;
first $=0.0$;
last $=1.99 \mathrm{E} 2$;
Ramp (n, first, last, x); /* $x[i]=i$ */
$a=3.5 ;$
$\mathrm{b}=-2.75$;
for (i=0; i<n; i++)
$y[i]=a{ }^{\star} \exp \left(b^{*} x[i]\right)$;
/* Find the best exponential fit in z. */
ExpFit (x, y, $n, z, \quad$ \&amp, \&decay, \&mse);

## ExpWin

```
int status = ExpWin (double x[], int n, double final);
```


## Purpose

Applies an exponential window to the input sequence $\mathbf{x}$. If $\mathbf{y}$ represents the output sequence, ExpWin obtains the elements of $\mathbf{y}$ using the following formula:

$$
y_{i}=x_{i} e^{a i}
$$

where $a=\frac{\ln (f)}{n-1}$
$f$ is the final value
$n$ is the number of elements in $x$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | On input, $\mathbf{x}$ contains the input signal. |
| $\mathbf{n}$ | integer | Number of elements in the input array. |
| final | double-precision | Final value of the exponential window <br> function. |

## Output

| Name | Type | Description |
| :---: | :---: | :--- |
| $\mathbf{x}$ | double-precision array | On output, $\mathbf{x}$ contains the signal after <br> ExpWin applies the exponential window. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## F_Dist

```
int status = F_Dist (double f, int n, int m, double *p);
```


## Purpose

Calculates the one-sided probability $\mathbf{p}$ :

$$
p=\operatorname{prob}(F \leq f)
$$

where $F$ is a random variable from the F -distribution with $\mathbf{n}$ and $\mathbf{m}$ degrees of freedom

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{f}$ | double-precision | $-\infty<\mathbf{f}<\infty$. |
| $\mathbf{n}$ | integer | Degrees of freedom. |
| $\mathbf{m}$ | integer | Degrees of freedom. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{p}$ | double-precision | Probability $(0 \leq \mathbf{p}<1)$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
double \(x, p ;\)
int \(n, m ;\)
\(\mathrm{x}=-123.456\);
\(\mathrm{n}=6\);
m = 7;
F_Dist (x, n, m, \&p);
/* Now \(\mathrm{p}=0\) because F -distributed variables are non-negative. */
```


## FFT

```
int status = FFT (double x[], double y[], int n);
```


## Purpose

Calculates the Fast Fourier Transform of the complex data. Let $X=x+j y$ be the complex array:

$$
Y=\mathrm{FFT}(X)
$$

FFT can perform the operation in place and overwrite the input arrays $\mathbf{x}$ and $\mathbf{y}$. Refer to the About the Fast Fourier Transform (FFT) section in Chapter 1, Advanced Analysis Library Overview.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Real part of complex array. |
| $\mathbf{y}$ | double-precision array | Imaginary part of complex array. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Real part of FFT. |
| $\mathbf{y}$ | double-precision array | Imaginary part of FFT. |

## [. 3 Note n must be a power of two.

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* Generate two arrays with random numbers and calculate the Fast Fourier Transform. */
double $x[256], y[256] ;$
int $n$;
$\mathrm{n}=256$;
Uniform ( $\mathrm{n}, 17, \mathrm{x}$ );
Uniform ( $\mathrm{n}, 17, \mathrm{y}$ );
$\operatorname{FFT}(\mathrm{x}, \mathrm{y}, \mathrm{n})$;

## FHT

```
int status = FHT (double x[], int n);
```


## Purpose

Calculates the Fast Hartley Transform using the following formula:

$$
X_{k}=\sum_{i=0}^{n-1} x_{i} \operatorname{cas}\left(\frac{2 \pi i k}{n}\right)
$$

where $X_{k}$ is the $k^{\text {th }}$ point of the FHT

$$
\operatorname{cas}(k)=\cos (k)+\sin (k)
$$

FHT can perform the operation in place and overwrite the $\mathbf{x}$ input array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Array to transform. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Hartley Transform. |

4. 3 Note n must be a power of two.

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* Generate an array with random numbers and calculate its Fast Hartley Transform. */

```
double x[256];
```

    int \(n\);
    \(\mathrm{n}=256\);
    Uniform (n, 17, x);
    FHT ( \(x, n\) );
    
## FIR_Coef

```
int status = FIR_Coef (int type, double fs, double fl, double fh, int taps,
    double coef[]);
```


## Purpose

Generates a set of FIR filter coefficients based on the window design method. FIR_Coef returns the coefficients as the truncated impulse response of an ideal frequency response of the selected filter type. type has the valid values shown in Table 2-29.

Table 2-29. Valid type Values

| Value | Description |
| :---: | :--- |
| 0 | lowpass filter; $\mathbf{f h}$ is not used |
| 1 | highpass filter; $\mathbf{f h}$ is not used |
| 2 | bandpass filter |
| 3 | bandstop filter |

Use Convolve to achieve the actual filtering:

$$
\mathrm{y}_{n}=\sum_{i=0}^{\text {taps }-1} \operatorname{coef}_{i} \times x_{n-1}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| type | integer | Controls the filter type of the FIR filter <br> coefficients to design. |
| fs | double-precision | Sampling frequency in hertz. |
| $\mathbf{f l}$ | double-precision | Desired lower cutoff frequency in hertz. |
| $\mathbf{f h}$ | double-precision | Desired upper cutoff frequency in hertz. |
| taps | integer | Desired length of the FIR filter. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Calculated output window FIR filter <br> coefficients. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## FlatTopWin

```
int status = FlatTopWin (double x[], int n);
```


## Purpose

Applies a flat top window to the input sequence $\mathbf{x}$. If $\mathbf{y}$ represents the output sequence, FlatTopWin obtains the elements of $\mathbf{y}$ using the following formula:

$$
y_{i}=x_{i}\left(0.2810639-0.5208972 \cos \left(\frac{2 \pi i}{n}\right)+0.1980399 \cos \left(\frac{4 \pi i}{n}\right)\right)
$$

where $n$ is the number of elements in $x$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | On input, $\mathbf{x}$ contains the input signal. |
| $\mathbf{n}$ | integer | Number of elements in the input array. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | On output, $\mathbf{x}$ contains the signal after <br> Flat TopWin applies the flat top window. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## ForceWin

```
int status = ForceWin (double x[], int n, double duty);
```


## Purpose

Applies a force window to the input sequence $\mathbf{x}$ :

$$
x_{i}= \begin{cases}x_{i} & 0 \leq i \leq \operatorname{int}\left(\left(\frac{d u t y}{100}\right) \times n\right) \\ 0 & \text { elsewhere }\end{cases}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | On input, $\mathbf{x}$ contains the input signal. |
| $\mathbf{n}$ | integer | Number of elements in the input array. |
| duty | double-precision | Duty cycle, in percent, of the force window. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | On output, $\mathbf{x}$ contains the signal after <br> ForceWin applies the force window. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## ForwSub

```
int status = ForwSub (void *a, double y[], int n, double x[], int p[]);
```


## Purpose

Solves the linear equations $a \times x=y$ by forward substitution. ForwSub assumes a to be an n-by-n lower triangular matrix with all diagonal elements equal to one. ForwSub obtains $\mathbf{x}$ using the following formulas:

$$
\begin{gathered}
x_{0}=y_{0} \\
x_{i}=y_{i}-\sum_{j=0}^{i-1} a_{i, j} \times x_{j} \quad \text { for } i=1,2, \ldots, n-1
\end{gathered}
$$

ForwSub can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{a}$ | double-precision <br> 2D array | Input matrix. |
| $\mathbf{y}$ | double-precision array | Input vector. |
| $\mathbf{n}$ | integer | Dimension size of $\mathbf{a .}$. |
| $\mathbf{p}$ | integer array | Permutation vector. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Solution vector. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Using This Function

Use ForwSub in conjunction with LU and BackSub to solve linear equations. ForwSub obtains the parameter $\mathbf{p}$ from LU. If you are not using LU, set $p_{i}=i$.

Refer to the Lu function description for more information.

## Example

```
/* to solve a linear equation A*x = y */
double A[10][10], x[10], y[10];
int p[10]; /* permutation vector */
int sign, n;
n = 10;
LU (A, n, p, &sign); /* LU decomposition of A */
ForwSub (A, y, n, x, p); /* forward substitution */
BackSub (A, x, n, x); /* backward substitution */
```


## FreeAnalysisMem

```
void FreeAnalysisMem (void *pointer);
```


## Purpose

Frees the memory that PeakDetector allocated internally for the output arguments.

## Parameter

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| pointer | void pointer | Pointer to memory to free. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following code example shows how to use FreeAnalysisMem in conjunction with

```
PeakDetector.
```

```
main()
{
    double *x = NULL;
    double *amplitudes = NULL;
    double *locations = NULL;
    double *secondDerivatives = NULL;
    int err = 0;
    int xSize;
    /* Insert code here to determine xSize. */
    x = (double *)malloc (xSize * sizeof(double));
    err = PeakDetector(x, xSize, 0.01, 3, 0, 1, 0, &count, &locations,
    &amplitudes, &secondDerivatives);
    /* Memory is allocated internally by the PeakDetector function for
    the locations, amplitudes and second_derivatives outputs. Use the
    FreeAnalysisMem function to free this memory. */
    FreeAnalysisMem(locations);
    FreeAnalysisMem(amplitudes);
    FreeAnalysisMem(secondDerivatives);
}
```


## FreelIRFilterPtr

```
int status = FreeIIRFilterPtr (IIRFilterPtr filterInformation);
```


## Purpose

Frees the IIR cascade filter structure and all internal arrays.

## Parameter

Input

| Name | Type | Description |
| :---: | :--- | :--- |
| filterInformation | IIRFilterPtr | Pointer to the filter structure that contains <br> the filter coefficients and the internal filter <br> information. |
|  |  | Refer to the AllocIIRFilterPtr <br> function description for more information <br> about the filter structure. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## GaussNoise

```
int status = GaussNoise (int n, double sDev, int seed, double noise[]);
```


## Purpose

Generates an array of random Gaussian numbers distributed with expected zero mean value and the standard deviation you specify.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples. |
| sDev | double-precision | Standard deviation you specify. |
| seed | integer | Initial seed value. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| noise | double-precision array | Gaussian noise pattern. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Using This Function

You specify the expected standard deviation of the pattern GaussNoise returns. The expected mean value is zero; that is, the noise array values are expected to be centered about zero. When seed $\geq 0$, GaussNoise generates a new random sequence using the seed value. When seed $<0$, the previously generated random sequence continues.

## Example

```
/* The following code generates an array of random Gaussian
distributed numbers. */
double x[20], sDev;
int n;
n = 20;
sDev = 5.0;
GaussNoise (n, sDev, 17, x);
```


## GenCosWin

```
int status = GenCosWin (double x[], int n, double a[], int na);
```


## Purpose

Applies a general cosine window to the input sequence $\mathbf{x}$. If $\mathbf{y}$ represents the output sequence, GenCosWin obtains the elements of $\mathbf{y}$ using the following formula:

$$
y_{i}=x_{i} \sum_{k=0}^{n a-1}(-1)^{k} a_{k} \cos \left(\frac{2 \pi k i}{n}\right)
$$

where $a$ is the array of coefficients $n a$ is the number of coefficients $n$ is the number of elements in $x$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | On input, $\mathbf{x}$ contains the input signal. |
| $\mathbf{n}$ | integer | Number of elements in the input array. |
| $\mathbf{a}$ | double-precision array | General cosine coefficient array. |
| na | integer | Number of elements in $\mathbf{a}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | On output, $\mathbf{x}$ contains the signal after <br> GenCosWin applies the general cosine <br> window. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## GenDeterminant

```
int status = GenDeterminant (void *A, int n, int matrixType, double *det);
```


## Purpose

Calculates the determinant of the real, square input matrix A. In contrast to Determinant, GenDeterminant allows you to specify the type of matrix type with the matrixType parameter. The input matrix can be upper or lower triangular, general, or positive definite.

For upper or lower triangular matrices, the determinant equals the product of the diagonal elements of the matrix. For a positive definite matrix, GenDeterminant first calculates the Cholesky factorization of the input matrix and then calculates the determinant as the square of the determinant of the upper triangular matrix $\mathbf{R}$. Refer to the Cholesky function description for more information.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| A | double-precision <br> 2D array | Input square matrix. |
| $\mathbf{n}$ | integer | Number of elements in one dimension of <br> the matrix. |
| matrixType | integer | Type of the matrix. Choose the matrix type <br> correctly because it significantly affects the <br> speed of computation. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| det | double | Determinant of the input matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

Table 2-30 shows valid matrix type values.
Table 2-30. Valid Matrix Type Values

| Matrix Type | Value |
| :--- | :---: |
| General matrix | 0 |
| Positive definite | 1 |
| Upper triangular | 2 |
| Lower triangular | 3 |

## GenEigenValueVector

```
int status = GenEigenValueVector (void *A, int n, int outputChoice,
    ComplexNum eigenValues[], void *eigenVectors);
```


## Purpose

Calculates the eigenvalues $\lambda$ and the corresponding eigenvectors $\mathbf{x}$ of a real, square input matrix $\mathbf{A}$. The following formula defines the eigenvalues and the corresponding eigenvectors:

$$
A x=\lambda x
$$

Although the input matrix is real, the eigenvalues and the eigenvectors can be complex if the matrix is not symmetric.

The outputChoice parameter determines what to calculate. Depending on your application, you can choose to calculate just the eigenvalues or to calculate both the eigenvalues and the eigenvectors.

The eigenValues output parameter is a 1D, complex array of $\mathbf{n}$ elements. The eigenVectors output parameter is an $\mathbf{n}$-by-n, complex matrix (2D array). Each $i^{t h}$ column of this matrix is the eigenvector that corresponds to the $i^{\text {th }}$ component of the eigenValues. Each eigenvector is normalized so that its largest component equals one.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| A | double-precision <br> 2D array | Input square matrix. |
| $\mathbf{n}$ | integer | Number of elements in one dimension of <br> the matrix. |
| outputChoice | integer | Pass 0 for eigenvalues only; 1 for both <br> eigenvalues and eigenvectors. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| eigenValues | ComplexNum array | Resulting eigenvalues of the input matrix. |
| eigenVectors | ComplexNum 2D array | Resulting eigenvectors of the input matrix. <br> You can pass NULL if outputChoice is 0. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The following C typedef statement defines the ComplexNum structure:

```
typedef struct {
    double real;
    double imaginary;
    } ComplexNum;
```


## GenInvMatrix

```
int status = GenInvMatrix (void *A, int n, int matrixType, void *B);
```


## Purpose

Calculates the inverse of the real, square input matrix $\mathbf{A}$. If $\mathbf{B}$ denotes the inverse of the matrix A:

$$
A B=I \quad \text { where } I \text { is the identity matrix }
$$

In contrast to InvMatrix, GenInvMatrix allows you to specify the type of the input matrix with the matrixType parameter. The input matrix can be an upper or lower triangular matrix, a general, square matrix, or a positive definite matrix. You can save significant computation time if you properly specify the type of the matrix.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | double-precision <br> 2D array | Input square matrix. |
| $\mathbf{n}$ | integer | Number of elements in one dimension of <br> the matrix. |
| matrixType | integer | Type of the matrix. Choose the matrix type <br> correctly because it significantly affects the <br> speed of computation. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| B | double-precision <br> 2D array | Calculated inverse matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

Table 2-31 shows valid matrix type values.
Table 2-31. Valid Matrix Type Values

| Matrix Type | Value |
| :--- | :---: |
| General matrix | 0 |
| Positive definite | 1 |
| Upper triangular | 2 |
| Lower triangular | 3 |

## GenLinEqs

```
int status = GenLinEqs (void *A, int n, int m, double y[], int matrixType,
    double x[]);
```


## Purpose

Solves for the unknown vector $\mathbf{x}$ in the linear system of equations:

$$
\begin{equation*}
A x=y \tag{2-2}
\end{equation*}
$$

where $A$ is the real input matrix $y$ is the known vector on the right side

The input matrix can be square or rectangular. The number of elements in $\mathbf{y}$ must equal the number of rows in the matrix $\mathbf{A}$.

GenLinEqs calculates the solution using the Singular Value Decomposition technique.
In the case of non-singular, square matrices, in which no row or column is a linear combination of any other row or column, GenLinEqs solves for the unique solution $\mathbf{x}$.

Two possibilities exist in the case of rectangular matrices. If the number of rows is greater than the number of columns, the system has more equations than unknowns and is an overdetermined system. Because the solution that satisfies the Equation (2-2) might not exist, this procedure finds the least square solution $\mathbf{x}$, which minimizes $\|A\|_{2}$. If the number of rows is less than the number of columns, the system has more unknowns than equations and is an underdetermined system. It might have infinite solutions that satisfy Equation (2-2). This procedure calculates the minimum 2-norm solution.

If the input matrix is rank-deficient, GenLinEqs returns a warning.
The matrixType parameter specifies the type of the input matrix. The input matrix can be an upper or lower triangular matrix, a general matrix, or a positive definite matrix.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | double-precision <br> 2D array | Input matrix. The matrix can be square or <br> rectangular. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in $\mathbf{A}$. |


| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Complex array that contains the set of <br> known vector coefficients. |
| matrixType | integer | Type of the input matrix. Choose the matrix <br> type correctly because it significantly <br> affects the speed of computation. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Solution to the linear system of equations. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. If the <br> input matrix is rank-deficient, GenLinEqs <br> returns the warning code 20001. |

## Parameter Discussion

Table 2-32 shows valid matrix type values.
Table 2-32. Valid Matrix Type Values

| Matrix Type | Value |
| :--- | :---: |
| General matrix | 0 |
| Positive definite | 1 |
| Upper triangular | 2 |
| Lower triangular | 3 |

## GenLSFit

```
int status = GenLSFit (void *H, int n, int k, double y[], double stdDev[],
    int algorithm, double z[], double b[],
    double covar[], double *mse);
```


## Purpose

Finds the best fit $\mathbf{k}$-dimensional plane and the set of linear coefficients using the least chi-squares method for observation data sets:

$$
\left(x_{i, 0}, x_{i, 1}, \ldots, x_{i, k-1}, x_{i}\right)
$$

where $i=0,1, \ldots, n-1$
$n=$ the number of your observation data sets

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{H}$ | double-precision <br> 2D array | An $\mathbf{n}$-by-k matrix that contains the <br> observation data $\left(x_{i, 0}, x_{i, 1}, \ldots, x_{i, k-1}\right)$ for <br> $i=0,1, \ldots, \mathbf{n}-1$, where $\mathbf{n}$ is the number <br> of rows in $\mathbf{H}, \mathbf{k}$ is the number of columns <br> in $\mathbf{H}$. |
| $\mathbf{n}$ | integer | Number of rows of $\mathbf{H}$ as well as the number <br> of elements in $\mathbf{y}$. |
| $\mathbf{k}$ | integer | Number of columns of $\mathbf{H}$ as well as the <br> number of elements in $\mathbf{b}$. |
| $\mathbf{y}$ | double-precision array | Number of elements in $\mathbf{y}$ should equal the <br> number of rows in $\mathbf{H}$. |


| Name | Type | Description |
| :---: | :---: | :---: |
| stdDev | double-precision array | Standard deviation $\sigma_{i}$ for data point $\left(x_{i}, y_{i}\right)$. If they are equal or if you do not know, pass an empty array, and GenLSFit ignores this parameter. The size of this array should equal $n$. |
| algorithm | integer | Algorithm to use to solve the multiple linear regression model. The algorithm has six selections: <br> $0=$ SVD <br> 1 = Givens <br> 2 = Givens2 <br> 3 = Householder <br> $4=\mathrm{LU}$ decomposition <br> 5 = Cholesky algorithm |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z}$ | double-precision array | Fitted data GenLSFit calculates by using <br> the coefficients $\mathbf{b}$. |
| $\mathbf{b}$ | double-precision array | Set of coefficients that minimize $\chi^{2}$, which <br> Equation (2-3) defines. |
| covar | double-precision <br> 2D array | Matrix of covariances with $\mathbf{k}$-by-k <br> elements. $c_{j, k}$ is the covariance between $b_{j}$ <br> and $b_{k}$, and $c_{j, j}$ is the variance of $b_{j}$. If you <br> pass an empty array for covar, GenLSFit <br> does not calculate this matrix. |
| mse | double-precision | Mean squared error. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Using This Function

You can use GenLSFit to solve multiple linear regression problems and to solve for the linear coefficients in a multiple-function equation.

The general least squares linear fit problem can be described as follows. Given a set of observation data, find a set of coefficients that fit the linear model:

$$
\begin{align*}
& y_{i}=b_{0} x_{i, 0}+\ldots+b_{k-1} x_{i, k-1} \\
= & \sum_{j=0}^{k-1} b_{j} x_{i, j} \quad \text { for } i=0,1, \ldots, n-1 \tag{2-3}
\end{align*}
$$

where $b$ is the set of coefficients
$n$ is the number of elements in $y$ and the number of rows of $H$
$k$ is the number of elements in $b$
$x_{i, j}$ is your observation data, which $H$ contains

$$
H=\left[\begin{array}{cccc}
x_{0,0} & x_{0,1} & \ldots & x_{0, k-1} \\
x_{1,0} & x_{1,1} & & x_{1, k-1} \\
\vdots & & & \\
x_{n-10} & x_{n-12} & & x_{n-1, k-1}
\end{array}\right]
$$

You can write Equation (2-2) as $Y=H B$.
The previous discussion leads to a multiple linear regression model, which uses several variables:

$$
x_{i, 0}, x_{i, 1}, \ldots, x_{i, k-1}
$$

to predict one variable $y_{i}$. In contrast, LinFit, ExpFit, and PolyFit are all based on a single predictor variable, which uses one variable to predict another variable.

In most cases, we have more observation data than coefficients. The formulas in Equation (2-3) might not produce the solution. The fit problem becomes to find the coefficients $B$ that minimize the difference between the observed data, $y_{i}$, and the predicted value:

$$
z_{i}=\sum_{j=0}^{k-1} b_{j} x_{i, j}
$$

GenLSFit uses the least chi-squares plane method to obtain the coefficients in Equation (2-3), that is, finding the solution, $B$, which minimizes the following quantity:

$$
\begin{equation*}
\chi^{2}=\sum_{i=0}^{n-1}\left(\frac{y_{i}-z_{i}}{\sigma_{i}}\right)^{2}=\sum_{i=0}^{n-1}\left(\frac{y_{i}-\sum_{i=0}^{k-1} b_{j} x_{i, j}}{\sigma_{i}}\right)^{2}=\left|H_{0} B-Y_{0}\right|^{2} \tag{2-4}
\end{equation*}
$$

where $h_{0_{i, j}}=\frac{x_{i, j}}{\sigma_{i}}, y_{0_{i}}=\frac{y_{i}}{\sigma_{i}} \quad$ for $i=0,1, \ldots, n-1 \quad$ for $j=0,1, \ldots, k-1$
In Equation (2-4), $\sigma_{i}$ is the standard deviation, stdDev. If the measurement errors are independent and normally distributed with constant standard deviation $\sigma_{i}=\sigma$, Equation (2-4) is also the least squares estimation.

There are different ways to minimize $\chi^{2}$. One way to minimize $\chi^{2}$ is to set the partial derivatives of $\chi^{2}$ to zero with respect to $b_{0}, b_{1}, \ldots, b_{k-1}$ :

$$
\left\{\begin{array}{c}
\frac{\partial \chi^{2}}{\partial b_{0}}=0 \\
\frac{\partial \chi^{2}}{\partial b_{1}}=0 \\
\cdot \\
\cdot \\
\cdot \\
\frac{\partial \chi^{2}}{\partial b_{k-1}}=0
\end{array}\right.
$$

The previous equations can be written as:

$$
\begin{equation*}
H_{0}^{T} H_{0} B=H_{0}^{T} Y \tag{2-5}
\end{equation*}
$$

$H_{0}^{T}$ is the transposition of $H_{0}$.
Equations (2-5) and (2-4) are also called normal equations of the least squares problems.
You can solve them using LU or Cholesky factorization algorithms, but the solution from the normal equations is susceptible to round-off error.

The preferred way to minimize $\chi^{2}$ is to find the least squares solution of the equations:

$$
H_{0} B=Y_{0}
$$

You can use QR or Singular Value Decomposition factorization to find the solution, B. For QR factorization, you can choose Householder, Givens, or Givens2, also called fast Givens.

Different algorithms can give you different precision. In some cases, if one algorithm cannot solve the equation, perhaps another algorithm can. You can try different algorithms to find the one best suited to your data.

GenLSFit calculates the covariance matrix covar as follows:

$$
\text { covar }=\left(H_{0}^{T} H_{0}\right)^{-1}
$$

The best fitted curve $\mathbf{z}$ is given by the following formula:

$$
z_{i}=\sum_{j=0}^{k-1} b_{j} x_{i, j}
$$

GenLSFit obtains the mse using the following formula:

$$
m s e=\frac{1}{n} \sum_{i=0}^{n-1}\left(\frac{y_{i}-z_{i}}{\sigma_{i}}\right)^{2}
$$

You can think of the polynomial fit that has a single predictor variable as a special case of multiple regression. If the observation data sets are $\left(x_{i}, y_{i}\right)$ where $i=0,1, \ldots, n-1$, the model for polynomial fit is as follows:

$$
\begin{equation*}
y_{i}=\sum_{j=0}^{k-i} b_{j} x_{i}^{j}=b_{0}+b_{1} x_{i}+b_{2} x_{i}^{2}+\ldots+b_{k-1} x_{i}^{k-1} \quad \text { where } i=0,1,2, \ldots, n-1 \tag{2-6}
\end{equation*}
$$

Comparing Equations (2-3) and (2-6) shows that $x_{i j}=x_{i}^{j}$. In other words:

$$
x_{i 0}=x_{i}^{0}, x_{i, 1}=x_{i}, x_{i, 2}=x_{i}^{2}, \ldots, x_{i, k-1}=x_{i}^{k-1}
$$

In this case, you can build $\mathbf{H}$ as follows:

$$
H=\left[\begin{array}{ccccc}
1 & x_{0} & x_{0}^{2} & \ldots & x_{0}^{k-1} \\
1 & x_{1} & x_{1}^{2} & & x_{1}^{k-1} \\
\vdots & & & & \\
1 & x_{n-1} & x_{n-1} & & x_{n-1}^{k-1}
\end{array}\right]
$$

Instead of using $x_{i, j}=x_{i}^{j}$, you can choose another function formula to fit the data sets $\left(x_{i}, y_{i}\right)$. In general, you can select $x_{i, j}=f_{j}\left(x_{i}\right)$. Here, $f_{j}\left(x_{i}\right)$ is the function model that you choose to fit your observation data. In polynomial fit, $f_{j}\left(x_{i}\right)=x_{i}^{j}$.

In general, you can build $\mathbf{H}$ as follows:

$$
H=\left[\begin{array}{lllll}
f_{0}\left(x_{0}\right) & f_{1}\left(x_{0}\right) & f_{2}\left(x_{0}\right) & \cdots & f_{k-1}\left(x_{0}\right) \\
f_{0}\left(x_{1}\right) & f_{1}\left(x_{1}\right) & f_{2}\left(x_{1}\right) & & f_{k-1}\left(x_{1}\right) \\
\vdots & & & & \\
f_{0}\left(x_{n-1}\right) & f_{1}\left(x_{n-1}\right) & f_{2}\left(x_{n-1}\right) & & f_{k-1}\left(x_{n-1}\right)
\end{array}\right]
$$

Your fit model is:

$$
y_{i}=b_{0} f_{0}(x)+b_{1} f_{1}(x)+\ldots+b_{k-1} f_{k-1}(x)
$$

The following two examples show how to use GenLSFit. The first example uses the function to perform multiple regression analysis based entirely on tabulated observation data. The second solves for the linear coefficients in a multiple-function equation.

## Example: Predicting Cost

Suppose you want to estimate the total cost, in dollars, of a production of baked scones using the quantity produced, $X_{1}$, and the price of one pound of flour, $X_{2}$. To keep things simple, the following five data points form the sample data table shown in Table 2-33.

Table 2-33. Sample Data Table

| Cost (dollars) <br> $\mathbf{Y}$ | Quantity <br> $\mathbf{X}_{\mathbf{1}}$ | Flour Price <br> $\mathbf{X}_{\mathbf{2}}$ |
| :---: | :---: | :---: |
| $\$ 150$ | 295 | $\$ 3.00$ |
| $\$ 75$ | 100 | $\$ 3.20$ |
| $\$ 120$ | 200 | $\$ 3.10$ |
| $\$ 300$ | 700 | $\$ 2.80$ |
| $\$ 50$ | 60 | $\$ 2.50$ |

You want to estimate the coefficients to the following formula:

$$
Y=b_{0}+b_{1} X_{1}+b_{2} X_{2}
$$

The only parameters you must build are $\mathbf{H}$ (observation matrix) and $\mathbf{y}$ arrays. Each column of $\mathbf{H}$ is the observed data for each independent variable: The first column is one because the coefficient $b_{0}$ is not associated with any independent variable.

Fill in $\mathbf{H}$ as follows:

$$
H=\left[\begin{array}{ccc}
1 & 295 & 3 \\
1 & 100 & 3.20 \\
1 & 200 & 3.10 \\
1 & 700 & 280 \\
1 & 60 & 250
\end{array}\right]
$$

The following code is based on this example.

```
/* example of predicting cost using GenLSFit */
int k, n, algorithm, status;
double H[5][3], y[5], z[5], b[3], X1[5], X2[5], mse;
double *stdDev=0, *covar=0; /* Define empty arrays; the function will
                                    ignore these parameters. */
n = 5;
k = 3;
/* Read in data for X1, X2, and y. */
.
/* Construct matrix H. */
for(i=0;i<n;i++) {
    H[i][0] = 1; /* Fill in the first column of H. */
    H[i][1] = X1[i]; /* Fill in the second column of H. */
    H[i][2] = X2[i]; /* Fill in the third column of H. */
}
algorithm = 0; /* Use SVD algorithm. */
status = GenLSFit(H, n, k, y, stdDev, algorithm, z, b, covar, &mse);
```


## Example: Linear Combinations

Suppose that you have samples from a transducer, $\mathbf{y}$ values, and you want to solve for the coefficients of the model:

$$
y=b_{0}+b_{1} \sin (\omega x)+b_{2} \cos (\omega x)+b_{3} x^{3}
$$

To build $\mathbf{H}$, set each column to the independent functions evaluated at each $x$ value. Assuming there are $100 x$ values, $\mathbf{H}$ would be the following array:

$$
H=\left[\begin{array}{cccc}
1 & \sin \left(\omega x_{0}\right) & \cos \left(\omega x_{0}\right) & x_{0} \\
1 & \sin \left(\omega x_{1}\right) & \cos \left(\omega x_{1}\right) & x_{1}^{2} \\
1 & \sin \left(\omega x_{2}\right) & \cos \left(\omega x_{2}\right) & x_{2}^{2} \\
\vdots & & & \\
1 & \sin \left(\omega x_{99}\right) & \cos \left(\omega x_{99}\right) & x_{99}^{2}
\end{array}\right]
$$

The following code is based on this example.

```
/* example of linear combinations using GenLSFit */
int i, k, n, algorithm, status;
double H[100][4], y[100], z[100], b[4], x[100], mse, w;
double *stdDev=0, *covar=0; /* Define empty arrays, the function will
                                    ignore these parameters. */
n = 100;
k = 4;
w = 0.2;
/* Read in data for x and y. */
.
•
/* Construct matrix H. */
for(i=0;i<n;i++) {
    H[i][0] = 1; /* Fill in the first column of H. */
    H[i][1] = sin(w*x[i]);/* Fill in the second column of H. */
    H[i][2] = cos(w*x[i]);/* Fill in the third column of H. */
    H[i][3] = pow(x[i],3);/* Fill in the fourth column of H. */
}
algorithm = 0; /* Use SVD algorithm. */
status = GenLSFit(H, n, k, y, stdDev, algorithm, z, b, covar, &mse);
```


## GenLSFitCoef

```
int status = GenLSFitCoef (void *H, int n, int k, double y[], double b[],
                        int algorithm);
```


## Purpose

Finds the set of linear fit coefficients, which describe the linear curve that best represents the input data GenLSFitCoef uses to obtain the least squares solution technique. The general form of the $\mathbf{k}$-dimension linear fit is as follows:

Let $i=0,1, \ldots, n$ be your $i^{\text {th }}$ observation
$x_{i, j}, \ldots, x_{i, k-1}$ be $k-1$ observed $x$ points
$y_{i}$ be observed $y$ points
Compose the $\mathbf{H}$ matrix as follows:

$$
H=\left[\begin{array}{lllll}
1 & x_{0,1} & x_{0,2} & \ldots & x_{0, k-1} \\
1 & x_{1,1} & x_{1,2} & & x_{1, k-1} \\
\vdots & & & & \\
1 & x_{n-1,1} & x_{n-1,2} & & x_{n-1, k-1}
\end{array}\right]
$$

GenLSFitCoef obtains the general LS linear fit coefficient $b_{k}$ by minimizing the quantity:

$$
Q=\sum_{i=0}^{n-1}\left(y_{i}-z_{i}\right)^{2}=\sum_{i=0}^{n-1}\left(y_{i}-b_{0}-\sum_{j=1}^{k-1} b_{j} x_{i, j}\right)^{2}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{H}$ | double-precision <br> 2D array | Input matrix that represents the formula you <br> use to fit the data set $(x, y) . H_{i, j}$ are the <br> function values of $x_{i}$. |
| $\mathbf{n}$ | integer | Number of rows of $\mathbf{H}$, as well as the number <br> of elements in $\mathbf{y}$. |
| $\mathbf{k}$ | integer | Number of columns of $\mathbf{H}$, as well as the <br> number of elements in $\mathbf{b}$. |


| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Array that contains the $\mathbf{y}$-coordinates of the <br> $(x, y)$ data sets to fit. |
| algorithm | integer | Algorithm to use to solve the multiple linear <br> regression model. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| $\mathbf{b}$ | double-precision array | Contains the set of linear coefficients that <br> best fit the multiple linear regression model <br> in a least squares sense. The size of this <br> array must be at least $\mathbf{k}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The algorithm has the valid selection as shown in Table 2-34.
Table 2-34. Valid Algorithm Selections

| Selection | Description |
| :---: | :--- |
| 0 | Singular value decomposition (default) |
| 1 | Givens decomposition |
| 2 | Square root free Givens decomposition |
| 3 | Household transformation |
| 4 | LU decomposition |
| 5 | Cholesky decomposition |

Each algorithm might offer different precision depending on the input data. Given the coefficient vector $\mathbf{b}$ and $\mathbf{H}$, GenLSF itCoef can calculate the fitted data $z_{i}$ by a simple matrix multiplication:

$$
Z=H \times b
$$

and can calculate the mean squared error by:

$$
m s e=\frac{1}{n} \sum_{i=0}^{n-1}\left(z_{i}-y_{i}\right)^{2}
$$

## GetAnalysisErrorString

```
char *message = GetAnalysisErrorString (int errorNum)
```


## Purpose

Converts the error number an Analysis Library function returns into a meaningful error message.

## Parameter

Input

| Name | Type | Description |
| :---: | :--- | :--- |
| errorNum | integer | Status an Analysis Library function returns. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| message | string | Explanation of error. |

## HamWin

```
int status = HamWin (double x[], int n);
```


## Purpose

Applies a Hamming window to the $\mathbf{x}$ input signal. The following formula defines the Hamming window:

$$
w_{i}=0.54-0.46 \times \cos \left(\frac{2 \pi i}{n}\right) \quad \text { for } i=0,1, \ldots, n-1
$$

HamWin obtains the output signal using the following formula:

$$
x_{i}=x_{i} \times w_{i} \quad \text { for } i=0,1, \ldots, n-1
$$

HamWin performs the window operation in place; that is, the windowed data $\mathbf{x}$ replaces the input data $\mathbf{x}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Windowed data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## HanWin

```
int status = HanWin (double x [], int n);
```


## Purpose

Applies a Hanning window to the $\mathbf{x}$ input signal. The following formula defines the Hanning window:

$$
w_{i}=0.5-0.5 \cos \left(\frac{2 \pi i}{n}\right) \quad \text { for } i=0,1, \ldots, n-1
$$

HanWin obtains the output signal using the following formula:

$$
x_{i}=x_{i} \times w_{i} \quad \text { for } i=0,1, \ldots, n-1
$$

HanWin performs the window operation in place; that is, the windowed data $\mathbf{x}$ replaces the input data $\mathbf{x}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Windowed data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## HarmonicAnalyzer

```
int status = HarmonicAnalyzer (const double autoPowerSpectrum[],
    int autoPowerSpectrumSize, int frameSize,
    int numberOfHarmonics, int windowType,
    double samplingRate, int fundamental_Frequency,
    double harmonicAmplitude[],
    double harmonicFrequency[], int *percent_THD,
    int *percentTHDNoise);
```


## Purpose

Finds the amplitude and frequency of the fundamental and harmonic components present in autoPowerSpectrum. HarmonicAnalyzer also calculates the percent of total harmonic distortion and the total harmonic distortion plus noise.

If the sampling rate is $1,000 \mathrm{~Hz}$ and the fundamental frequency is 250 Hz , the number of harmonics is limited by samplingRate $/(2 \times$ fundamental_Frequency $)=2$. If you set numberOfHarmonics equal to 4 , HarmonicAnalyzer sets the third and the fourth element of the harmonicAmplitude and harmonicFrequency array equal to 0.0 .

Typically, you should pass the time-domain signal to ScaledWindow and then to AutoPowerSpectrum. You then pass the output of AutoPowerSpectrum to HarmonicAnalyzer.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| autoPowerSpectrum | double-precision array | Single-sided auto power spectrum <br> of the windowed signal. |
| autoPowerSpectrumSize | integer | Number of elements in <br> autoPowerSpectrum. |
| frameSize | integer | Number of samples in the <br> time-domain signal array. |
| numberOfHarmonics | integer | Number of harmonic components. |\(\left|\begin{array}{l}Window type the function applies <br>


to the time-domain signal.\end{array}\right|\)| Input sampling rate in hertz. |  |
| :--- | :--- |
| windowType | double |
| samplingRate | integer |
| fundamental_Frequency | Estimate of the fundamental <br> frequency. |

Output

| Name | Type | Description |
| :--- | :--- | :--- |
| harmonicAmplitude | double-precision array | Amplitudes of the fundamental <br> components and its harmonics. |
| harmonicFrequency | double-precision array | Frequencies of the fundamental <br> component and its harmonics. |
| percent_THD | integer | Percent total harmonic distortion <br> present in autoPowerSpectrum. |
| percentTHDNoise | integer | Percent total harmonic <br> distortion plus noise present <br> in autoPowerSpectrum. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error <br> codes. |

## Histogram

```
int status = Histogram (double inputArray[], int numberOfElements,
    double base, double top, int histogramArray[],
    double axisArray[], int intervals);
```


## Purpose

Calculates the histogram of the inputArray. If the input sequence is

$$
X=\{0,1,3,3,4,4,4,5,5,8\}
$$

the Histogram: $h(X)$ of $X$ for eight intervals is

$$
h(x)=\left\{h_{0}, h_{1}, h_{2}, h_{3}, h_{4}, h_{5}, h_{6}, h_{7}\right\}=\{1,1,0,2,3,2,0,1\}
$$

Notice that the histogram of the input sequence $X$ is a function of $X$.
The function obtains Histogram: $h(X)$ as follows: Histogram scans the input sequence $X$ to determine the range of values in it. Then the function establishes the interval width, $\Delta x$, according to the specified number of intervals,

$$
\Delta x=\frac{\max -\min }{m}
$$

where $\max$ is the maximum value found in the input sequence $X$ $\min$ is the minimum value found in the input sequence $X$ $m$ is the specified number of intervals

Let $\chi$ represent the output sequence $X$ because the histogram is a function of $X$. The function evaluates elements of $\chi$ using

$$
\chi_{i}=\min +0.5 \times \Delta x+i \times \Delta x \quad \text { for } i=0,1,2, \ldots, m-1
$$

Histogram defines the $i^{\text {th }}$ interval $\Delta_{i}$ to be the range of values from $\chi_{i}-0.5 \times \Delta x$ up to but not including $\chi_{i}+0.5 \times \Delta x$

$$
\Delta_{i}=\left[\chi_{i}-0.5 \times \Delta x: \chi_{i}+0.5 \times \Delta x\right) \quad \text { for } i=0,1,2, \ldots, m-1
$$

and defines the function $y_{i}(x)$ to be

$$
y_{i}(x)= \begin{cases}1 & \text { if } x \in \text { union of } \Delta_{i} \\ 0 & \text { elsewhere }\end{cases}
$$

Histogram has unity value if the value of $x$ falls within the specified interval. Otherwise it is zero. Notice that the interval $\Delta_{i}$ is centered about $\chi_{i}$, and its width is $\Delta_{x}$.

The last interval, $\Delta_{m-1}$, is defined as $\left[\chi_{m-i}-0.5 \times \Delta x: \chi_{m-i}+0.5 \times \Delta x\right]$. In other words, if a value equals max, it is counted as belonging to the last interval.

Finally, Histogram evaluates the histogram sequence $h$ using

$$
h_{i}=\sum_{j=0}^{n-1} y_{i}\left(x_{j}\right) \quad \text { for } i=0,1,2, \ldots, m-1
$$

where $\quad h_{i}$ represents the elements of the output sequence Histogram: $h(X)$ $n$ is the number of elements in the input sequence $X$

Histogram obtains the histogram by counting the number of times the elements in the input array fall in the $i^{t h}$ interval.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| inputArray | double-precision array | Input array. |
| numberOfElements | integer | Number of elements in inputArray. |
| base | double-precision | Lower range. |
| top | double-precision | Upper range. |
| intervals | integer | Number of intervals. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| histogramArray | integer array | Histogram of inputArray. |
| axisArray | double-precision array | Histogram axis array; contains the <br> midpoint values of the intervals. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## IIRCascadeFiltering

```
int status = IIRCascadeFiltering (const double x[], int n,
    IRFilterPtr filterInformation,
    double y[]);
```


## Purpose

Filters the input sequence using the cascade IIR filter specified by the filterInformation structure. Each of the IIR cascaded stages is second order for lowpass and highpass filters, and fourth order for bandpass and bandstop filters.
filterInformation is the pointer to the filter structure that contains the filter coefficients and the internal filter information. You must allocate this structure by calling AllocIIRFilterPtr and then call one of the cascade IIR design functions Bw_CascadeCoef, Ch_CascadeCoef, Elp_CascadeCoef, InvCh_CascadeCoef, or Bessel_CascadeCoef before you call IIRCascadeFiltering.

The filterInformation structure contains the internal filter state information for the filtering operation so you can call IIRCascadeFiltering in a loop to continually filter new input array data and produce new output filtered data.

If you finish filtering one set of input data and want to filter a completely new data set, call ResetIIRFilter before you call IIRCascadeFiltering with the new data. ResetIIRFilter causes the internal filter state information to clear before the next filtering operation.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | const double-precision | Array that contains the raw data to filter. |
| $\mathbf{n}$ | integer | Specifies the number of points in both the <br> input $\mathbf{x}$ and output $\mathbf{y}$. |
| filterInformation | IIRFilterPtr | Pointer to the filter structure that contains <br> the filter coefficients and the internal filter <br> information. <br> Refer to the AllocIIRFilterPtr <br> function description for more information <br> about the filter structure. |

## Output

| Name | Type | Description |
| :---: | :---: | :--- |
| $\mathbf{y}$ | double-precision array | Array that contains the output of the <br> IIR filtering operation. The size of this <br> array must be at least $\mathbf{n}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## IIRFiltering

```
int status = IIRFiltering (double x[], int nx, double a[], double y1[],
    int na, double b[], double x1[], int nb,
    double y[]);
```


## Purpose

Filters the input sequence using the IIR filter specified by reverse coefficients $\mathbf{a}$ and forward coefficients b by:

$$
y_{n}=\frac{1}{a_{0}}\left(\sum_{i=0}^{n b-1} b_{i} x_{n-i}-\sum_{i=1}^{n a-1} a_{i} y_{n-i}\right)
$$

The reverse and forward coefficients are obtained by respective IIR coefficient functions such as Bw_Coef.

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Raw data to filter. |
| $\mathbf{n x}$ | integer | Number of points in both the $\mathbf{x}$ coefficients <br> array and the $\mathbf{x} 1$ conditions array. |
| $\mathbf{a}$ | double-precision array | Array that contains the reverse coefficients <br> for the IIR filtering operation. |
| $\mathbf{y 1}$ | integer | y1 contains the initial conditions, or states. <br> The size of this array must be at least <br> na -1. |
| na | double-precision array | Number of coefficients in both the a <br> coefficients array and the $\mathbf{y 1}$ conditions <br> array. |
| $\mathbf{b}$ | drray that contains the forward coefficients |  |
| for the IIR filtering operation. |  |  |$|$| double-precision array |
| :--- |
| $\mathbf{x 1}$ The size of this array must be at least |
| nb -1. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y 1}$ | double-precision array | on output, $\mathbf{y 1}$ contains the final conditions <br> for the next iterations. |
| $\mathbf{x 1}$ | double-precision array | on output, $\mathbf{x 1}$ contains the final conditions <br> for the next iterations. |
| $\mathbf{y}$ | double-precision array | $\mathbf{y}$ array that contains the output of the <br> IIR filtering operation. The size of this array <br> must be at least $\mathbf{n x}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Impulse

```
int status = Impulse (int n, double amp, int index, double x[]);
```


## Purpose

Generates an array of numbers that has the pattern of an impulse waveform. Impulse obtains the $i^{\text {th }}$ element of the output array using the following formula:

$$
x_{i}= \begin{cases}a m p & \text { if } i=\text { index } \\ 0 & \text { otherwise }\end{cases}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| amp | double-precision | Amplitude. |
| index | integer | Impulse index. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Impulse array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* The following code generates the impulse pattern
x = {0.0, 0.0, 1.5, 0.0, 0.0}. */
double x[5], amp;
int n, i;
n = 5;
i = 2;
amp = 1.5;
    Impulse (n, amp, i, x);
```


## ImpulseResponse

```
int status = ImpulseResponse (double stimulus[], double response[], int n,
    double impulse[]);
```


## Purpose

Calculates the impulse response of a network based on time-domain signals stimulus and response. The impulse response is in the time domain. The impulse response is the inverse Fourier transform of the transfer function:

$$
\text { impulse }=\operatorname{ReInvFFT}\left(\frac{S x y(f)}{S x x(f)}\right)
$$

where $\operatorname{Sxy}(f)$ is the two-sided cross power spectrum of the stimulus $(x)$ with the response ( $y$ )
$S x x(f)$ is the two-sided auto power spectrum of the stimulus

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| stimulus | double-precision array | Contains the time-domain signal, usually <br> the network stimulus. |
| response | double-precision array | Contains the time-domain signal, usually <br> the network response. |
| $\mathbf{n}$ | integer | Number of elements in the input array. <br> $\mathbf{n}$ must be a power of 2. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| impulse | double-precision array | Impulse that contains the impulse response <br> of the network based on time-domain <br> signals stimulus and response. The size of <br> this array must be at least $\mathbf{n}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Integrate

```
int status = Integrate (double x[], int n, double dt, double xInit,
    double xFinal, double y[]);
```


## Purpose

Calculates the discrete integral of the input array. Integrate obtains the $i^{\text {th }}$ element of the resulting array using the following formula:

$$
y_{i}=\sum_{j=0}^{i}\left(x_{j-1}+4 x_{j}+x_{j+1}\right) \times \frac{d t}{6} \quad \text { where } x_{-1}=x \text { Init and } x_{n}=x \text { Final }
$$

Integrate can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{d t}$ | double-precision | Sampling interval. |
| $\mathbf{x I n i t}$ | double-precision | Initial condition. |
| xFinal | double-precision | Final condition. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Integrated array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* Generate an array with random numbers and integrate it. */
double x[200], y[200];
double $d t, x I n i t, x F i n a l ;$
int $n$;
$\mathrm{n}=200$;
$d t=0.001 ;$
xInit $=-0.5$;
xFinal $=-0.25$;
Uniform ( $\mathrm{n}, 17$, x );
Integrate ( $x, n, d t, x I n i t, x F i n a l, y)$;

## InvCh_BPF

```
int status = InvCh_BPF (double x[], int n, double fs, double fl, double fh,
    double atten, int order, double y[]);
```


## Purpose

Filters the input array using a digital bandpass inverse Chebyshev filter. InvCh_BPF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f l}$ | double-precision | Lower cutoff frequency. |
| $\mathbf{f h}$ | double-precision | Higher cutoff frequency. |
| atten | double-precision | Stop band attenuation in decibels. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* Generate a random signal and filter it using a fifth-order bandpass inverse Chebyshev filter. The pass band is from 200.0 to 300.0. */ double $x[256], y[256], f s, f l, f h, ~ a t t e n ;$
int $n$, order;
$\mathrm{n}=256$;
$\mathrm{fs}=1000.0 ;$
$\mathrm{fl}=200.0$;
fh = 300.0;
atten $=40.0$;
order $=5$;
Uniform ( $\mathrm{n}, 17, \mathrm{x}$ );
InvCh_BPF (x, $n, f s, f l, f h, ~ a t t e n, ~ o r d e r, ~ y) ;$

## InvCh_BSF

```
int status = InvCh_BSF (double x[], int n, double fs, double fl, double fh,
    double atten, int order, double y[]);
```


## Purpose

Filters the input array using a digital bandstop inverse Chebyshev filter. InvCh_BSF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f l}$ | double-precision | Lower cutoff frequency. |
| $\mathbf{f h}$ | double-precision | Higher cutoff frequency. |
| atten | double-precision | Stop band attenuation in decibels. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* Generate a random signal and filter it using a fifth-order bandstop inverse Chebyshev filter. The stop band is from 200.0 to 300.0. */ double $x[256], y[256], f s, f l, f h, ~ a t t e n ;$
int $n$, order;
$\mathrm{n}=256$;
$\mathrm{fs}=1000.0 ;$
$\mathrm{fl}=200.0$;
fh = 300.0;
atten $=40.0$;
order $=5$;
Uniform ( $\mathrm{n}, 17, \mathrm{x}$ );
InvCh_BSF (x, n, fs, fl, fh, atten, order, y);

## InvCh_CascadeCoef

```
int status = InvCh_CascadeCoef (double fs, double fl, double fh,
    double atten, IIRFilterPtr filterInformation);
```


## Purpose

Generates the set of cascade form filter coefficients to implement an IIR filter as specified by the inverse Chebyshev filter model.
filterInformation is the pointer to the filter structure that contains the filter coefficients and the internal filter information. You must allocate this structure by calling AllocIIRFilterPtr before you call this cascade IIR filter design function.

To redesign another filter, you should first call FreeIIRFilterPtr to free the present filter structure and then call AllocIIRFilterPtr with the new type and order parameters before you call InvCh_CascadeCoef.

If the type and order remain the same, you can call this IIR design function without calling FreeIIRFilterPtr and AllocIIRFilterPtr. In this case, you should properly reset the filtering operation for that structure by calling ResetIIRFilter before the first call to IIRCascadeFiltering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| fs | double-precision | Specifies the sampling frequency in hertz. |
| fl | double-precision | Specifies the desired lower cutoff frequency <br> of the filter in hertz. |
| fh | double-precision | Specifies the desired upper cutoff frequency <br> of the filter in hertz |
| atten | double-precision | Specifies the stop band attenuation, in <br> decibels, of the IIR filter to design. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| filterInformation | IIRFilterPtr | Pointer to the filter structure that contains <br> the filter coefficients and the internal filter <br> information. |
|  |  | Refer to the AllocIIRFilterPtr <br> function description for more information <br> about the filter structure. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Design a cascade lowpass inverse Chebyshev IIR filter. */
double fs, fl, fh, atten, x[256], y[256];
int type, order, n;
IIRFilterPtr filterInfo;
n = 256;
fs = 1000.0;
fl = 200.0;
atten = 60.0;
order = 5;
type = 0; /* lowpass */
Uniform(n,17,x);
filterInfo = AllocIIRFilterPtr(type, order);
if(filterInfo!=0) {
    InvCh_CascadeCoef(fs, fl, fh, atten, filterInfo);
    IIRCascadeFiltering(x, n, filterInfo, y);
    FreeIIRFilterPtr(filterInfo);
}
```


## InvCh_Coef

```
int status = InvCh_Coef (int type, int order, double fs, double fl,
    double fh, double atten, double a[], int na,
    double b[], int nb);
```


## Purpose

Generates the set of filter coefficients to implement an IIR filter as specified by the inverse Chebyshev filter model. type has the valid values shown in Table 2-35.

Table 2-35. Valid type Values

| Value | Description |
| :---: | :--- |
| 0 | lowpass filter; $\mathbf{f h}$ is not used |
| 1 | highpass filter; fh is not used |
| 2 | bandpass filter |
| 3 | bandstop filter |

$\mathbf{a}$ and $\mathbf{b}$ are the reverse and forward filter coefficients. Use IIRFiltering to achieve the actual filtering:

$$
y_{n}=\frac{1}{a_{0}}\left(\sum_{i=0}^{n b-1} b_{i} x_{n-i}-\sum_{i=1}^{n a-1} a_{i} y_{n-i}\right)
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| type | integer | Controls the filter type of the inverse <br> Chebyshev IIR filter coefficients. |
| order | integer | Order of the IIR filter. |
| fs | double-precision | Sampling frequency in hertz. |
| fl | double-precision | Desired lower cutoff frequency of the filter <br> in hertz. |
| fh | double-precision | Desired lower cutoff frequency of the filter <br> in hertz. |


| Name | Type | Description |
| :--- | :--- | :--- |
| atten | double-precision | Stop band attenuation, in decibels, of the <br> IIR filter to design. |
| na | integer | Number of coefficients in the a coefficient <br> array. |
| nb | integer | Number of coefficients in the $\mathbf{b}$ coefficient <br> array. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| a | double-precision array | Array that contains the reverse coefficients <br> of the designed IIR filter. |
| b | double-precision array | Array that contains the forward coefficients <br> of the designed IIR filter. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## InvCh_HPF

```
int status = InvCh_HPF (double x[], int n, double fs, double fc,
    double atten, int order, double y[]);
```


## Purpose

Filters the input array using a digital highpass inverse Chebyshev filter. InvCh_HPF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f c}$ | double-precision | Cutoff frequency. |
| atten | double-precision | Stop band attenuation in decibels. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* Generate a random signal and filter it using a fifth-order highpass inverse Chebyshev filter. */
double $x[256], y[256], f s, f c, ~ a t t e n ;$
int $n$, order;
$\mathrm{n}=256$;
$\mathrm{fs}=1000.0 ;$
fc $=200.0$;
atten $=40.0$;
order $=5$;
Uniform ( $\mathrm{n}, 17, \mathrm{x}$ );
InvCh_HPF (x, $n, f s, f c, ~ a t t e n, ~ o r d e r, ~ y) ;$

## InvCh_LPF

```
int status = InvCh_LPF (double x[], int n, double fs, double fc, double atten,
    int order, double y[]);
```


## Purpose

Filters the input array using a digital lowpass inverse Chebyshev filter. InvCh_LPF can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f c}$ | double-precision | Cutoff frequency. |
| atten | double-precision | Stop band attenuation in decibels. |
| order | integer | Filter order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Filtered data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* Generate a random signal and filter it using a fifth-order lowpass inverse Chebyshev filter. */
double $x[256], y[256], f s, f c, ~ a t t e n ;$
int $n$, order;
$\mathrm{n}=256$;
$\mathrm{fs}=1000.0 ;$
fc $=200.0$;
atten $=40.0$;
order $=5$;
Uniform ( $\mathrm{n}, 17, \mathrm{x}$ );
InvCh_LPF (x, $n, f s, f c, ~ a t t e n, ~ o r d e r, ~ y) ;$

## InvF_Dist

```
int status = InvF_Dist (double p, int n, int m, double *f);
```


## Purpose

Calculates $\mathbf{f}$, given a probability ( $0 \leq \mathbf{p}<1$ ), such that:

$$
\operatorname{prob}(F<f)=p
$$

where $F$ is a random variable from an $F$-distribution with $\mathbf{n}$ and $\mathbf{m}$ degrees of freedom

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{p}$ | double-precision | Probability $(0 \leq \mathbf{p}<1)$. |
| $\mathbf{n}$ | integer | Degrees of freedom. |
| $\mathbf{m}$ | integer | Degrees of freedom. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{f}$ | double-precision | The unique number $\mathbf{f}$ such that <br> prob $(F<f)=\mathbf{p}$, where $F$ is a random <br> variable from an F-distribution with $\mathbf{n}$ and <br> $\mathbf{m}$ degrees of freedom. |

$\sqrt{3}$ Note When $\mathbf{p}=0, \mathbf{f}=0$.

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
double p, f;
    int n, m;
    p = 0.635;
    n = 2;
    m = 4;
    InvF_Dist (p, n, m, &f);
```


## InvFFT

```
int status = InvFFT (double x[], double y[], int n);
```


## Purpose

Calculates the inverse Fast Fourier Transform of the complex data. Let $X=x+j y$ be the complex array:

$$
Y=\mathrm{FFT}^{-1}(X)
$$

InvFFT performs the operation in place and overwrites the input arrays $\mathbf{x}$ and $\mathbf{y}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Real part of complex array. |
| $\mathbf{y}$ | double-precision array | Imaginary part of complex array. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Real part of inverse FFT. |
| $\mathbf{y}$ | double-precision array | Imaginary part of inverse FFT. |

## 4. Note n must be a power of two.

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* Generate two arrays with random numbers and calculate the inverse
Fast Fourier Transform. */
double $x[256], y[256] ;$
int $n$;
$\mathrm{n}=256$;
Uniform ( $\mathrm{n}, 17, \mathrm{x}$ );
Uniform ( $\mathrm{n}, 17, \mathrm{y}$ );
$\operatorname{InvFFT}(\mathrm{x}, \mathrm{y}, \mathrm{n})$;

## InvFHT

```
int status = InvFHT (double x[], int n);
```


## Purpose

Calculates the inverse Fast Hartley Transform using the following formula:

$$
x_{i}=\frac{1}{n} \sum_{k=0}^{n-1} X_{k} \operatorname{cas}\left(\frac{2 \pi i k}{n}\right)
$$

where $x_{i}$ is the $i^{\text {th }}$ point of the inverse FHT

$$
\operatorname{cas}(x)=\cos (x)+\sin (x)
$$

InvFHT performs the operation in place and overwrites the $\mathbf{x}$ input array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Array to transform. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Inverse Fast Hartley Transform. |

## 4 Note n must be a power of two.

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* Generate an array with random numbers and calculate its inverse Fast Hartley Transform. */
double x[256];
int $n$;
$\mathrm{n}=256$;
Uniform ( $\mathrm{n}, 17, \mathrm{x}$ );
InvFHT ( $\mathrm{x}, \mathrm{n}$ );

## InvMatrix

```
int status = InvMatrix (void *x, int n, void *y);
```


## Purpose

Finds the inverse matrix of an input matrix. InvMatrix can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same matrices.

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision <br> 2D array | Input matrix. |
| $\mathbf{n}$ | integer | Dimension size of matrix. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision <br> 2D array | Inverse matrix. |

## d $\frac{\square}{3}$ Note $\quad$ The input matrix must be an n-by-n square matrix.

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## InvN_Dist

```
int status = InvN_Dist (double p, double *x);
```


## Purpose

Calculates $\mathbf{x}$, given a probability $(0<\mathbf{p}<1)$, such that:
$\operatorname{prob}(X<x)=p \quad$ where $X$ is a random variable from a standard normal distribution

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{p}$ | double-precision | Probability $(0<\mathbf{p}<1)$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision | The unique number $\mathbf{x}$ such that <br> prob $(X<\mathbf{x})=\mathbf{p}$, where $X$ is a random <br> variable from a standard normal <br> distribution. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
double p, x;
p = 0.5;
InvN_Dist (p, &x);
```


## InvT_Dist

```
int status = InvT_Dist (double p, int n, double *t);
```


## Purpose

Calculates $\mathbf{t}$, given a probability $(0<\mathbf{p}<1)$, such that:

$$
\operatorname{prob}(T<t)=p
$$

where $T$ is a random variable from a T-distribution with $\mathbf{n}$ degrees of freedom

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{p}$ | double-precision | Probability $(0<\mathbf{p}<1)$. |
| $\mathbf{n}$ | integer | Degrees of freedom. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| $\mathbf{t}$ | double-precision | The unique number $\mathbf{t}$ such that <br> $\operatorname{prob}(T<\mathbf{t})=\mathbf{p}$, where $T$ is a random <br> variable from a T-distribution with <br> n degrees of freedom. |
|  |  |  |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
double p, t;
int n;
p = 0.635;
n = 2;
InvT_Dist (p, n, &t);
```


## InvXX_Dist

```
int status = InvXX_Dist (double p, int n, double *x);
```


## Purpose

Calculates $\mathbf{x}$, given a probability $(0 \leq \mathbf{p}<1)$, such that:

$$
\operatorname{prob}(\chi<x)=p
$$

where $\chi$ is a random variable from a chi-square distribution with $\mathbf{n}$ degrees of freedom

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{p}$ | double-precision | Probability $(0 \leq \mathbf{p}<1)$. |
| $\mathbf{n}$ | integer | Degrees of freedom. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision | The unique number $\mathbf{x}$ such that <br> prob $(\chi<\mathbf{x})=\mathbf{p}$, where $\chi$ is a random <br> variable from a chi-square distribution with <br> $\mathbf{n}$ degrees of freedom. |

约 Note When $\mathbf{p}=0, \mathbf{x}=0$.

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
double p, x;
int n;
p = 0.635;
n = 2;
InvXX_Dist (p, n, &x);
```


## Ksr_BPF

```
int status = Ksr_BPF (double fs, double fl, double fh, int n, double coef[],
    double beta);
```


## Purpose

Designs a digital bandpass FIR linear phase filter using a Kaiser window. Ksr_BPF generates only the filter coefficients; it does not actually perform data filtering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f l}$ | double-precision | Lower cutoff frequency. |
| $\mathbf{f h}$ | double-precision | Higher cutoff frequency. |
| $\mathbf{n}$ | integer | Number of filter coefficients. |
| beta | double-precision | Shape parameter. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Filter coefficients. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The beta parameter controls the shape of a Kaiser window. A larger beta value results in a narrower Kaiser window. Table 2-36 lists some beta values and their equivalent windows.

Table 2-36. beta Values and Equivalent Windows

| beta | Window |
| :---: | :--- |
| 0.00 | Rectangular |
| 1.33 | Triangle |
| 3.86 | Hanning |
| 4.86 | Hamming |
| 7.04 | Blackman |

Refer to Discrete-Time Signal Processing by Oppenheim and Schafer for more information.

## Example

/* Design a 55-point bandpass FIR linear phase filter using a Kaiser window with beta $=4.5$. Filter the incoming signal with the designed filter. */
double $x[256], \operatorname{coef}[55], y[310], f s, f l, f h$, beta;
int $n, m ;$
fs $=1000.0 ; \quad / *$ sampling frequency */
fl $=200.0 ; \quad / *$ desired lower cutoff frequency */
fh $=300.0$; /* desired higher cutoff frequency */
/* pass band is from 200.0 to 300.0 */
$\mathrm{n}=55$;
/* filter length */
beta $=3$;
$\mathrm{m}=256$;
Ksr_BPF (fs, fl, fh, n, coef, beta);
Convolve (coef, n, x, m, y); /* Convolve the filter with the signal. */

## Ksr_BSF

```
int status = Ksr_BSF (double fs, double fl, double fh, int n, double coef[],
    double beta);
```


## Purpose

Designs a digital bandstop FIR linear phase filter using a Kaiser window. Ksr_BSF generates only the filter coefficients; it does not actually perform data filtering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f l}$ | double-precision | Lower cutoff frequency. |
| $\mathbf{f h}$ | double-precision | Higher cutoff frequency. |
| $\mathbf{n}$ | integer | Number of filter coefficients. |
| beta | double-precision | Shape parameter. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Filter coefficients. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The beta parameter controls the shape of a Kaiser window. A larger beta value results in a narrower Kaiser window. Table 2-37 lists some beta values and their equivalent windows.

Table 2-37. beta Values and Equivalent Windows

| beta | Window |
| :---: | :--- |
| 0.00 | Rectangular |
| 1.33 | Triangle |
| 3.86 | Hanning |
| 4.86 | Hamming |
| 7.04 | Blackman |

Refer to Discrete-Time Signal Processing by Oppenheim and Schafer for more information.

## Example

```
/* Design a 55-point bandstop FIR linear phase filter using a Kaiser
window with beta = 4.5. Filter the incoming signal with the designed
filter. */
double x[256], coef[55], y[310], fs, fl, fh, beta;
int n, m;
fs = 1000.0; /* sampling frequency */
fl = 200.0; /* desired lower cutoff frequency */
fh = 300.0; /* desired higher cutoff frequency */
    /* stop band is from 200.0 to 300.0 */
n = 55; /* filter length */
beta = 3;
m = 256;
Ksr_BSF (fs, fl, fh, n, coef, beta);
Convolve (coef, n, x, m, y); /* Convolve the filter with the signal. */
```


## Ksr_HPF

```
int status = Ksr_HPF (double fs, double fc, int n, double coef[],
    double beta);
```


## Purpose

Designs a digital highpass FIR linear phase filter using a Kaiser window. Ksr_HPF generates only the filter coefficients; it does not actually perform data filtering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f c}$ | double-precision | Cutoff frequency. |
| $\mathbf{n}$ | integer | Number of filter coefficients. |
| beta | double-precision | Shape parameter. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Filter coefficients. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The beta parameter controls the shape of a Kaiser window. A larger beta value results in a narrower Kaiser window. Table 2-38 lists some beta values and their equivalent windows.

Table 2-38. beta Values and Equivalent Windows

| beta | Window |
| :--- | :--- |
| 0.00 | Rectangular |
| 1.33 | Triangle |
| 3.86 | Hanning |
| 4.86 | Hamming |
| 7.04 | Blackman |

Refer to Discrete-Time Signal Processing by Oppenheim and Schafer for more information.

## Example

/* Design a 55-point highpass FIR linear phase filter using a Kaiser window with beta $=4.5$. Filter the incoming signal with the designed filter. */
double $x[256], \operatorname{coef}[55], y[310], f s, f c$, beta;
int $n, m ;$
fs $=1000.0 ; \quad / *$ sampling frequency */
fc $=200.0 ; \quad / *$ desired cutoff frequency */
$\mathrm{n}=55$ / /* filter length */
beta $=4.5$;
$m=256 ;$
Ksr_HPF (fs, fc, $n, ~ c o e f, ~ b e t a) ;$
Convolve (coef, $n, x, m, y) ; / *$ Convolve the filter with the signal. */

## Ksr_LPF

```
int status = Ksr_LPF (double fs, double fc, int n, double coef[],
    double beta);
```


## Purpose

Designs a digital lowpass FIR linear phase filter using a Kaiser window. Ksr_LPF generates only the filter coefficients; it does not actually perform data filtering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f c}$ | double-precision | Cutoff frequency. |
| $\mathbf{n}$ | integer | Number of filter coefficients. |
| beta | double-precision | Shape parameter. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Filter coefficients. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The beta parameter controls the shape of a Kaiser window. A larger beta value results in a narrower Kaiser window. Table 2-39 lists some beta values and their equivalent windows.

Table 2-39. beta Values and Equivalent Windows

| beta | Window |
| :--- | :--- |
| 0.00 | Rectangular |
| 1.33 | Triangle |
| 3.86 | Hanning |
| 4.86 | Hamming |
| 7.04 | Blackman |

Refer to Discrete-Time Signal Processing by Oppenheim and Schafer for more information.

## Example

```
/* Design a 55-point lowpass FIR linear phase filter using a Kaiser
window with beta = 4.5. Filter the incoming signal with the designed
filter. */
double x[256], coef[55], y[310], fs, fc, beta;
int n, m;
fs = 1000.0; /* sampling frequency */
fc = 200.0; /* desired cutoff frequency */
n = 55; /* filter length */
beta = 4.5;
m = 256;
Ksr_LPF (fs, fc, n, coef, beta);
Convolve (coef, n, x, m, y); /* Convolve the filter with the signal. */
```


## KsrWin

```
int status = KsrWin (double x[], int n, double beta);
```


## Purpose

Applies a Kaiser window to the $\mathbf{x}$ input signal. The following formula defines the Kaiser window:

$$
w_{i}=\frac{\operatorname{Io}\left(\text { beta } \times\left(1.0-a^{2}\right)^{1 / 2}\right)}{\operatorname{Io}(\text { beta })} \quad \text { for } i=0,1, \ldots, n-1
$$

where $a=\left|1-\frac{2 i}{n}\right|$
Io represents the zeroth ${ }^{\text {th }}$-order modified Bessel function of the first kind

KsrWin obtains the output signal using the formula:

$$
x_{i}=x_{i} \times w_{i} \quad \text { for } i=0,1, \ldots, n-1
$$

KsrWin performs the window operation in place; that is, the windowed data $\mathbf{x}$ replaces the input data $\mathbf{x}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| beta | double-precision | Shape parameter. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Windowed data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The beta parameter controls the shape of a Kaiser window. A larger beta value results in a narrower Kaiser window. Table 2-40 lists some beta values and their equivalent windows.

Table 2-40. beta Values and Equivalent Windows

| beta | Window |
| :--- | :--- |
| 0.00 | Rectangular |
| 1.33 | Triangle |
| 3.86 | Hanning |
| 4.86 | Hamming |
| 7.04 | Blackman |

Refer to Discrete-Time Signal Processing by Oppenheim and Schafer for more information.

## LinEqs

```
int status = LinEqs (void *A, double y[], int n, double x[]);
```


## Purpose

Solves the linear system of equations:

$$
A x=y
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | double-precision <br> 2D array | Input matrix. |
| $\mathbf{y}$ | double-precision array | Known vector. |
| $\mathbf{n}$ | integer | Dimension size of system. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Solution of vector. |

## $\rightarrow$ Note The A input matrix must be an n-by-n square matrix.

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Find the solution to the linear system of equations. */
double A[10][10], y[10], x[10];
int n;
n = 10;
    .
LinEqs (A, y, n, x);
```


## LinEv1D

```
int status = LinEv1D (double x[], int n, double a, double b, double y[]);
```


## Purpose

Performs a linear evaluation of a 1D array, $\mathbf{x}$. LinEv1D obtains the $i^{\text {th }}$ element of the output array, $\mathbf{y}$, using the formula:

$$
y_{i}=a \times x_{i}+b
$$

LinEv1D can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements. |
| $\mathbf{a}$ | double-precision | Multiplicative constant. |
| $\mathbf{b}$ | double-precision | Additive constant. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Linearly evaluated array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## LinEv2D

```
int status = LinEv2D (void *x, int n, int m, double a, double b, void *y);
```


## Purpose

Performs a linear evaluation of a 2D array, $\mathbf{x}$. LinEv2D obtains the $(i, j)^{t h}$ element of the output array, $\mathbf{y}$, using the formula:

$$
y_{i, j}=a \times x_{i, j}+b
$$

LinEv2D can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision <br> 2D array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in first dimension. |
| $\mathbf{m}$ | integer | Number of elements in second dimension. |
| $\mathbf{a}$ | double-precision | Multiplicative constant. |
| $\mathbf{b}$ | double-precision | Additive constant. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision <br> 2D array | Linearly evaluated array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## LinFit

```
int status = LinFit (double x[], double y[], int n, double z[],
    double *slope, double *intercept, double *mse);
```


## Purpose

Finds the slope and intercept values that best represent the linear fit of the data points ( $\mathbf{x}, \mathbf{y}$ ) using the least squares method. LinFit obtains the $i^{\text {th }}$ element of the output array, $\mathbf{z}$, using the following formula:

$$
z_{i}=\text { slope } \times x_{i}+\text { intercept }
$$

LinFit obtains the mean squared error (mse) using the following formula:

$$
m s e=\frac{\sum_{i=0}^{n-1}\left|z_{i}-y_{i}\right|^{2}}{n} \quad \text { where } n \text { is the number of sample points }
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | $\mathbf{x}$ values. |
| $\mathbf{y}$ | double-precision array | $\mathbf{y}$ values. |
| $\mathbf{n}$ | integer | Number of sample points. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z}$ | double-precision array | Best fit array. |
| slope | double-precision | Slope of line. |
| intercept | double-precision | y-intercept. |
| mse | double-precision | Mean squared error. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a ramp pattern and find the best linear fit. */
double x[200], y[200], z[200];
double start, end, a, b, slope, intercept, mse;
int n;
n = 200;
start = 0.0;
end = 1.99E2;
Ramp (n, start, end, x); /* x[i] = i */
a = 3.5;
b = -2.75;
LinEv1D (x, n, a, b, y); /* y[i] = a*x[i] + b */
/* Find the best linear fit in z. */
LinFit (x, y, n, z, &slope, &intercept, &mse);
```


## LU

```
int status = LU (void *A, int n, int p[], int *sign);
```


## Purpose

Performs an LU matrix decomposition:

$$
A=L U
$$

where $L$ is an $\mathbf{n}$-by-n lower triangular matrix with main diagonal elements all equal to one $U$ is an upper triangular matrix

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | double-precision <br> 2D array | Input matrix. |
| $\mathbf{n}$ | integer | Dimension size. |

Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | double-precision <br> 2D array | LU decomposition. |
| $\mathbf{p}$ | integer array | Permutation vector. |
| sign | integer | Row exchange indicator. |

## $\square$ Note $L$ and $\mathbf{U}$ output matrices overwrite the input matrix.

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

After LU executes, LU replaces the input matrix $\mathbf{A}$ with two triangular matrices. $\mathbf{L}$ occupies the lower triangular part of $\mathbf{A}$, and $\mathbf{U}$ occupies the upper triangular part of $\mathbf{A}$. The permutation vector $\mathbf{p}$ records possible row exchange information in the LU decomposition. sign $=0$ indicates that there is no such exchange or that there is an even number of such exchanges. sign $=1$ indicates that there is an odd number of such exchanges. $\mathbf{p}$ and sign are useful when solving the linear equations or computing the determinant. Use LU in conjunction with BackSub and ForwSub to solve a set of linear equations with the same matrix $\mathbf{A}$.

Refer to Numerical Recipes in C: The Art of Scientific Computing by Press, et al., for more information.

## MatrixMuI

```
int status = MatrixMul (void *X, void *Y, int n, int k, int m, void *Z);
```


## Purpose

Multiplies two 2D input matrices, $\mathbf{X}$ and $\mathbf{Y}$. MatrixMul obtains the $(i, j)^{\text {th }}$ element of the output matrix, $\mathbf{Z}$, using the formula:

$$
Z_{i, j}=\sum_{p=0}^{k-1} x_{i, p} \times y_{p, j}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{X}$ | double-precision <br> 2D array | $\mathbf{X}$ input matrix. |
| $\mathbf{Y}$ | double-precision <br> 2D array | $\mathbf{Y}$ input matrix. |
| $\mathbf{n}$ | integer | First dimension of $\mathbf{X}$. |
| $\mathbf{k}$ | integer | Second dimension of $\mathbf{X} ;$ first dimension <br> of $\mathbf{Y}$. |
| $\mathbf{m}$ | integer | Second dimension of $\mathbf{Y}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{Z}$ | double-precision <br> 2D array | Output matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

Confirm that the array sizes are correct. You must meet the following array sizes:

- $\quad \mathbf{X}$ must be $\mathbf{n}$ by $\mathbf{k}$.
- $\quad \mathbf{Y}$ must be $\mathbf{k}$ by $\mathbf{m}$.
- $\mathbf{Z}$ must be $\mathbf{n}$ by $\mathbf{m}$.


## Example

```
/* Multiply two matrices. Note: A x B - B x A, in general. */
double x[10][20], y[20][15], z[10][15];
int n, k, m;
n = 10;
k = 20;
m = 15;
MatrixMul (x, y, n, k, m, z);
```


## MatrixNorm

```
int status = MatrixNorm (void *A, int n, int m, int normType, double *norm);
```


## Purpose

Calculates the norm of a real input matrix $\mathbf{A}$. The input matrix can be square or rectangular. The norm of a matrix is a scalar that gives some measure of the size of the elements in the matrix. It is similar to the concept of magnitude or absolute value for scalar numbers.

There are different ways to calculate the norm of a matrix. The normType parameter indicates which type of norm to use to calculate the norm.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| A | double-precision <br> 2D array | Input matrix. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in A. |
| normType | integer | Type of norm to calculate. Refer to the <br> following Parameter Discussion section. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| norm | double-precision | Calculated norm of the input matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The normType parameter indicates what type of norm to use to calculate the condition number. Table 2-41 shows valid norm type values.

Table 2-41. Valid Norm Type Values

| Norm Type | Value | Meaning |
| :--- | :---: | :--- |
| 2-norm | 0 | Largest singular value of $\mathbf{A}$. |
| 1-norm | 1 | Largest column sum of $\mathbf{A .}$ |
| Frobenius-norm | 2 | Square root of the sum of the diagonal elements of $\mathbf{A}^{\mathbf{T}} \mathbf{A}$, <br> where $\mathbf{A}^{\mathbf{T}}$ is the complex conjugate transpose of $\mathbf{A}$. |
| Infinite-norm | 3 | Largest row sum of $\mathbf{A}$. |

## MatrixRank

```
int status = MatrixRank (void *A, int n, int m, double tolerance, int *rank);
```


## Purpose

Calculates the rank of the real input matrix $\mathbf{A}$. The input matrix can be square or rectangular.
The maximum number of linearly independent rows or columns of the matrix defines the rank of a matrix. The rank is always less than or equal to the minimum of the number of rows and columns of the matrix. If the rank equals this minimum value, the matrix is a full-rank matrix. Otherwise, it is a rank-deficient matrix.

The rank of a matrix can be calculated in a number of ways. MatrixRank first calculates the singular values of the input matrix and then calculates the rank as the number of singular values of the input matrix that are larger than the input tolerance.

You must specify the input tolerance as a positive number close to machine precision. If the matrix in your application is a full-rank matrix, any small value of tolerance gives the same rank. If the matrix in your application is a rank-deficient matrix, different values of tolerance can result in different values of rank.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | double-precision <br> 2D array | Input matrix. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in A. |
| tolerance | double-precision | Tolerance value. Refer to the following <br> Parameter Discussion section. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| rank | integer | Rank of the input matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

Matrix rank is the number of singular values in the input matrix that are larger than the tolerance. Set tolerance close to eps, which is the smallest possible double-precision, floating-point number.

## MaxMin1D

```
int status = MaxMin1D (double x[], int n, double *max, int *imax, double *min,
    int *imin);
```


## Purpose

Finds the maximum and minimum values in the input array and the respective indices of the first occurrence of the maximum and minimum values.

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\boldsymbol{\operatorname { m a x }}$ | double-precision | Maximum value. |
| $\boldsymbol{\operatorname { i m a x }}$ | integer | Index of $\max$ in $\mathbf{x}$ array. |
| $\boldsymbol{\operatorname { m i n }}$ | double-precision | Minimum value. |
| $\boldsymbol{\operatorname { i m i n }}$ | integer | Index of $\min$ in $\mathbf{x}$ array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate an array with random and find the maximum and minimum
values. */
double x[20], y[20];
double max, min;
int n, imax, imin;
n = 20;
Uniform (n, 17, x);
MaxMin1D (x, n, &max, &imax, &min, &imin);
```


## MaxMin2D

```
int status = MaxMin2D (void *X, int n, int m, double *max, int *imax,
    int *jmax, double *min, int *imin, int *jmin);
```


## Purpose

Finds the maximum and the minimum values in the 2D input array and the respective indices of the first occurrence of the maximum and minimum values. MaxMin2D scans the $\mathbf{X}$ array by rows.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{X}$ | double-precision <br> 2D array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in first dimension of $\mathbf{X}$. |
| $\mathbf{m}$ | integer | Number of elements in second dimension <br> of $\mathbf{X}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\boldsymbol{m a x}$ | double-precision | Maximum value. |
| imax | integer | Index of max in $\mathbf{X}$ array (first dimension). |
| jmax | integer | Index of $\max$ in $\mathbf{X}$ array (second <br> dimension). |
| $\boldsymbol{\operatorname { m i n }}$ | double-precision | Minimum value. |
| imin | integer | Index of $\mathbf{m i n}$ in $\mathbf{X}$ array (first dimension). |
| jmin | integer | Index of $\min$ in $\mathbf{X}$ array (second <br> dimension). |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* This example finds the maximum and minimum values as well as their
location within the array. */
double x[5][10], max, min;
int $n, m, i m a x, ~ j m a x, ~ i m i n, ~ j m i n ; ~$
$\mathrm{n}=5$;
m = 10;
MaxMin2D (x, $n, m, \& m a x, ~ \& i m a x, ~ \& j m a x, ~ \& m i n, ~ \& i m i n, ~ \& j m i n) ;$

## Mean

```
int status = Mean (double x[], int n, double *meanval);
```


## Purpose

Calculates the mean, or average, value of the input array. Mean calculates the mean using the following formula:

$$
\text { meanval }=\frac{\sum_{i=0}^{n-1} x_{i}}{n}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| meanval | double-precision | Mean value. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Median

```
int status = Median (double x[], int n, double *medianval);
```


## Purpose

Finds the median value of the $\mathbf{x}$ input array. To find the median value, Median first sorts the input array in ascending order. Let $S$ be the sorted array:

$$
\text { medianval }= \begin{cases}S\left(\frac{n}{2}\right) & \text { if } n \text { is odd } \\ 0.5 \times\left(S\left(\frac{n}{2}-1\right)+S\left(\frac{n}{2}\right)\right) & \text { if } n \text { is even }\end{cases}
$$

The x input array does not change.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| medianval | double-precision | Median value. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Mode

```
int status = Mode (double x[], int n, double xBase, double xTop,
                        int intervals, double *modeval);
```


## Purpose

Finds the mode of the $\mathbf{x}$ input array. The mode is defined as the value that most often occurs in a given set of samples. Mode determines the mode in terms of the histogram of the input array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| xBase | double-precision | Lower range. |
| $\mathbf{x T o p}$ | double-precision | Upper range. |
| intervals | integer | Number of intervals. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| modeval | double-precision | Mode value. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a Gaussian distributed random array and find its mode. */
double x[2000], max, min, modeval;
int n, intervals, imax, imin;
n = 2000;
intervals = 50;
GaussNoise (n, 1.0E0, 17, x);
MaxMin1D (x, n, &max, &imax, &min, &imin);
Mode (x, n, min, max, intervals, &modeval);
```


## Moment

```
int status = Moment (double x[], int n, int order, double *momentval);
```


## Purpose

Calculates the moment about the mean of the input array with the specified order. Moment uses the following formulas to find the moment:

$$
\text { momentval }=\sum_{i=0}^{n-1} \frac{\left(x_{i}-a v e\right)^{\text {order }}}{n} \quad \text { where ave }=\frac{\sum_{i=0}^{n-1} x_{i}}{n}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| order | integer | Moment order. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| momentval | double-precision | Moment about the mean. |

Note order must be greater than zero.

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate an array with random numbers and determine its skewness
(third-order moment) and its kurtosis (fourth-order moment). */
double x[200], skew, kurtosis;
int n, order;
n = 200;
Uniform (n, 17, x);
order = 3;
Moment (x, n, order, &skew);
order = 4;
Moment (x, n, order, &kurtosis);
```


## Mul1D

```
int status = Mul1D (double x[], double y[], int n, double z[]);
```


## Purpose

Multiplies two 1D arrays. Mul1D obtains the $i^{\text {th }}$ element of the output array using the following formula:

$$
z_{i}=x_{i} \times y_{i}
$$

Mul1D can perform the operation in place; that is, $\mathbf{z}$ can be the same array as either $\mathbf{x}$ or $\mathbf{y}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | $\mathbf{x}$ input array. |
| $\mathbf{y}$ | double-precision array | $\mathbf{y}$ input array. |
| $\mathbf{n}$ | integer | Number of elements to multiply. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z}$ | double-precision array | Result array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Mul2D

```
int status = Mul2D (void *X, void *Y, int n, int m, void *Z);
```


## Purpose

Multiplies two 2D arrays, $\mathbf{X}$ and $\mathbf{Y}$. Mul2D obtains the $(i, j)^{\text {th }}$ element of the output array, $\mathbf{Z}$, using the following formula:

$$
z_{i, j}=x_{i, j}+y_{i, j}
$$

Mul2D can perform the operation in place; that is, $\mathbf{Z}$ can be the same array as either $\mathbf{X}$ or $\mathbf{Y}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{X}$ | double-precision <br> 2D array | $\mathbf{X}$ input array. |
| $\mathbf{Y}$ | double-precision <br> 2D array | $\mathbf{Y}$ input array. |
| $\mathbf{n}$ | integer | Number of elements in first dimension. |
| $\mathbf{m}$ | integer | Number of elements in second dimension. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{Z}$ | double-precision <br> 2D array | Result array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## N_Dist

int status = N_Dist (double x, double *p);

## Purpose

Calculates the one-sided probability $\mathbf{p}$ :
$p=\operatorname{prob}(X \leq x) \quad$ where $X$ is a random variable from a standard normal distribution

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision | $-\infty<\mathbf{x}<\infty$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{p}$ | double-precision | Probability $(0<\mathbf{p}<1)$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

Note $\quad$ For computing the two-sided probability $p_{2}=\operatorname{prob}(-x \leq X \leq x)$, you can use the formula $p_{2}=1.0-2 \times \operatorname{prob}(X \leq-x)$.

## Example

```
double x, p;
x = -123.456;
N_Dist (x, &p);
```


## Neg1D

```
int status = Neg1D (double x[], int n, double y[]);
```


## Purpose

Negates the elements of the input array. Neg1D can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Negated values of the $\mathbf{x}$ input array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## NetworkFunctions

```
int status = NetworkFunctions (void *STIMULUS, void *RESPONSE, int n,
    int numFrames, double dt, double MAGSXY[],
    double PHASESXY[], double MAGHF[],
    double PHASEHF[], double COHERENCE[],
    double IMPULSE[], double *df);
```


## Purpose

Calculates the single-sided coherence function along with the averaged single-sided cross power spectrum, averaged single-sided frequency response, or transfer function, and impulse response from a 2D array of stimulus signals and a 2D array of response signals.

NetworkFunctions calculates the network functions as follows:

$$
\begin{gathered}
\text { avg cross power }=\operatorname{average}(S x y(f)) \\
\text { avg transfer function }=\frac{\operatorname{average}(S x y(f))}{\operatorname{average}(S x x(f))}
\end{gathered}
$$

average impulse response $=\operatorname{ReInvFFT}$ (avg two-sided transfer function)

$$
\text { coherence }=\frac{\mid \text { average }\left.S x y(f)\right|^{2}}{\text { average } S x x(f) \times \text { average } S y y(f)}
$$

where $\operatorname{Sxy}(f)$ is the two-sided cross power spectrum of $x$ and $y$
$\operatorname{Sxx}(f)$ is the two-sided auto power spectrum of $x$
$\operatorname{Syy}(f)$ is the two-sided auto power spectrum of $y$
$x$ is the stimulus signal
$y$ is the response signal
STIMULUS is a 2D array that contains a time-domain signal, usually the network stimulus. RESPONSE is a 2D array that contains a time-domain signal, usually the network response.

Each row in the stimulus array represents one frame of the network stimulus and is associated with one row of the response array, which represents one frame of the network response.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| STIMULUS | double-precision <br> 2D array | Contains the time-domain signal, usually <br> the network stimulus. The number of rows <br> should equal numFrames, and the number <br> of columns should equal $\mathbf{n}$. The size of this <br> array must be at least numFrames $\times \mathbf{n}$. |
| RESPONSE | double-precision <br> 2D array | Contains the time-domain signal, usually <br> the network stimulus. The number of rows <br> should equal numFrames, and the number <br> of columns should equal $\mathbf{n}$. The size of this <br> array must be at least numFrames $\times \mathbf{n}$. |
| $\mathbf{n}$ | integer | Number of elements in one frame of the <br> input stimulus and response arrays. |
| numFrames | integer | Number of frames, or rows, the input <br> stimulus and response arrays contain. |
| dt | double-precision | Sampling period of the time-domain signal, <br> usually in seconds. dt $=1 / f s$, where $f s$ <br> is |
| the sampling frequency of the time-domain |  |  |
| signal. |  |  |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| MAGSXY | double-precision array | Averaged single-sided cross power <br> spectrum between the stimulus and <br> response, in volts rms square if the input <br> signals are in volts. If the input signals are <br> not in volts, the results are in input signal <br> units rms square. This array must be at least <br> $\mathbf{n} / 2$ elements long. |
| PHASESXY | double-precision array | Averaged single-sided phase spectrum in <br> radians showing the difference between the <br> phases of the response signal and the <br> stimulus signal. This array must be at least <br> $\mathbf{n} / 2$ elements long. |


| Name | Type | Description |
| :--- | :--- | :--- |
| MAGHF | double-precision array | $\begin{array}{l}\text { Magnitude of the averaged single-sided } \\ \text { transfer function between the stimulus and } \\ \text { response signals. This array must be at least } \\ \mathbf{n} / 2 \text { elements long. }\end{array}$ |
| PHASEHF | double-precision array | $\begin{array}{l}\text { Phase, in radians of the averaged } \\ \text { single-sided transfer function between the } \\ \text { stimulus and response signals. }\end{array}$ |
| COHERENCE | double-precision array | $\begin{array}{l}\text { Averaged single-sided coherence function } \\ \text { spectrum. The coherence function shows } \\ \text { the frequency content of the response as a } \\ \text { result of the stimulus and measures the } \\ \text { validity of the network frequency response } \\ \text { measurement. This array must be at least }\end{array}$ |
| $\mathbf{n / 2 ~ e l e m e n t s ~ l o n g . ~}$ |  |  |$\}$

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## NonLinearFit

```
int status = NonLinearFit (double x[], double y[], double z[], int n,
    ModelFun *modelFunction, double a[], int ncoef,
    double *mse);
```


## Purpose

Uses the Levenberg-Marquardt algorithm to determine the least squares set of coefficients that best fit the set of input data points $(x, y)$ as expressed by a nonlinear function $\mathbf{y}=\mathrm{f}(\mathbf{x}, \mathbf{a})$ where $\mathbf{a}$ is the set of coefficients. NonLinearFit also gives the best fit curve $\mathbf{y}=f(\mathbf{x}, \mathbf{a})$.

You must pass a pointer to the nonlinear function $f(\mathbf{x}, \mathbf{a})$ along with a set of initial guess coefficients $\mathbf{a}$. NonLinearFit does not always give the correct answer. The correct output sometimes depends on the initial choice of a. It is very important to verify the final result.

NonLinearFit calculates the output mse (mean squared error) using the following formula:

$$
m s e=\frac{\sum_{i=0}^{n-1}\left(y_{i}-\mathrm{f}(x, a)\right)^{2}}{n}
$$

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Array of x-coordinates of the $(x, y)$ data sets <br> to fit. |
| $\mathbf{y}$ | double-precision array | Array of y-coordinates of the $(x, y)$ data <br> sets to fit. |
| $\mathbf{n}$ | integer | Number of elements in both the $\mathbf{x}$ and $\mathbf{y}$ <br> arrays. |
| modelFunction | ModelFun | Pointer to the model function, $\mathrm{f}\left(\mathbf{x}_{\mathbf{i}}, \mathbf{a}\right)$, used <br> in the nonlinear fitting algorithm. The <br> model function must be defined as follows: <br> double ModelFunct <br> (double x, double a [ ], <br> int ncoef); <br> where $\mathbf{a ~ c o n t a i n s ~ t h e ~ f u n c t i o n ~ c o e f f i c i e n t s . ~}$ |


| Name | Type | Description |
| :--- | :--- | :--- |
| a | double-precision array | On input, a gives a set of initial guess <br> coefficients. |
| ncoef | integer | Number of coefficients (size of a). |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z}$ | double-precision array | Best fit array, $\mathbf{y}=\mathrm{f}(\mathbf{x}, \mathbf{a})$. |
| $\mathbf{a}$ | double-precision array | Best fit coefficients. |
| mse | double-precision | Mean squared error between $\mathbf{y}$ and $\mathbf{z}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## NonLinearFitWithMaxIters

```
int status = NonLinearFitWithMaxIters (double x[], double y[], double z[],
    int n, int maximumIterations,
    ModelFun *modelFunction, double a[], int ncoef,
    double *mse);
```


## Purpose

Uses the Levenberg-Marquardt algorithm to determine the least squares set of coefficients that best fit the set of input data points $(x, y)$ as expressed by a nonlinear function $\mathbf{y}=\mathrm{f}(\mathbf{x}, \mathbf{a})$ where $\mathbf{a}$ is the set of coefficients. NonLinearFitWithMaxIters also gives the best fit curve $\mathbf{y}=\mathrm{f}(\mathbf{x}, \mathbf{a})$.

You must pass a pointer to the nonlinear function $f(\mathbf{x}, \mathbf{a})$ along with a set of initial guess coefficients a. NonLinearFit does not always give the correct answer. The correct output sometimes depends on the initial choice of $\mathbf{a}$. It is very important to verify the final result.

NonLinearFitWithMaxIters calculates the output mse (mean squared error) using the following formula:

$$
m s e=\frac{\sum_{i=0}^{n-1}\left(y_{i}-\mathrm{f}(x, a)\right)^{2}}{n}
$$

W Note If NonLinearFitWithMaxIters reaches the maximum number of iterations without reaching a solution, it returns an error. The outputs $\mathbf{z}$, a, and mse contain the best filtered array, best fit coefficients, and the mean square error at the end of maximum iterations.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Array of x-coordinates of the $(x, y)$ data <br> sets to fit. |
| $\mathbf{y}$ | double-precision array | Array of y-coordinates of the $(x, y)$ data <br> sets to fit. |
| $\mathbf{n}$ | integer | Number of elements in both the $\mathbf{x}$ and <br> $\mathbf{y}$ arrays. |


| Name | Type | Description |
| :--- | :--- | :--- |
| maximumIterations | integer | Maximum number of iterations <br> allowed. |
| modelFunction | ModelFun pointer | Pointer to the model function, $\mathrm{f}\left(\mathbf{x}_{\mathbf{i}}, \mathbf{a}\right)$, <br> used in the nonlinear fitting algorithm. <br> The model function must be defined as <br> follows: <br> double ModelFunct <br> (double x, double a [ ], <br> int ncoef); <br> where a contains the function <br> coefficients. |
| a | double-precision array | On input, a gives a set of initial guess <br> coefficients. |
| ncoef | integer | Number of coefficients (size of a). |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z}$ | double-precision array | Best fit array, $\mathbf{y}=\mathrm{f}(\mathbf{x}, \mathbf{a})$. |
| $\mathbf{a}$ | double-precision array | Best fit coefficients. |
| mse | double-precision | Mean squared error between $\mathbf{y}$ and $\mathbf{z}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Normal1D

```
int status = Normal1D (double x[], int n, double y[], double *ave,
    double *sDev);
```


## Purpose

Normalizes a 1D input vector. The output vector has the following form:

$$
y_{i}=\frac{x_{i}-a v e}{s D e v}
$$

where ave and $s D e v$ are the mean and the standard deviation of the input vector
Refer to the StdDev function description for the formulas Normal1D uses to find the mean and the standard deviation.

Normal1D can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input vector. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Normalized vector. |
| ave | double-precision | Mean value of $\mathbf{x}$. |
| sDev | double-precision | Standard deviation of $\mathbf{x}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* Generate a vector (1D array) with random samples and normalize it. */
double x[200], y[200], ave, sDev;
int $n$;
$\mathrm{n}=200$;
Uniform ( $\mathrm{n}, 17, \mathrm{x}$ );
Normal1D (x, $n, y, \& a v e, ~ \& s D e v) ;$

## Normal2D

```
int status = Normal2D (void *X, int n, int m, void *Y, double *ave,
    double *sDev);
```


## Purpose

Normalizes a 2D input matrix. The output matrix has the following form:

$$
y_{i, j}=\frac{x_{i, j}-a v e}{s D e v}
$$

where ave and $s D e v$ are the mean and the standard deviation of the input matrix
Refer to the StdDev function description for the formulas Normal2D uses to find the mean and the standard deviation.

Normal2D can perform the operation in place; that is, $\mathbf{X}$ and $\mathbf{Y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{X}$ | double-precision <br> 2D array | Input matrix. |
| $\mathbf{n}$ | integer | Size of first dimension. |
| $\mathbf{m}$ | integer | Size of second dimension. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{Y}$ | double-precision <br> 2D array | Normalized matrix. |
| ave | double-precision | Mean value of $\mathbf{X}$. |
| sDev | double-precision | Standard deviation of $\mathbf{X}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
    /* Normalize a matrix (2D array). */
    double x[10][20], y[10][20], ave, sDev;
    int n, m;
    n = 10;
    m = 20;
    •
    .
    Normal2D (x, n, m, y, &ave, &sDev);
```


## NumericIntegration

```
int status = NumericIntegration (double x[], int n, double dt, int method,
    double *ir);
```


## Purpose

Performs numeric integration on the data the input array $\mathbf{x}$ contains using one of the following four numeric integration methods: Trapezoidal Rule, Simpson's Rule, Simpson's $3 / 8$ Rule, or Bode Rule. You normally obtain the data to integrate by sampling some function $\mathrm{f}(t)$ at multiples of $\mathbf{d t}$. Your samples are $\mathrm{f}(0), \mathrm{f}(d t), \mathrm{f}(2 d t)$, and so on. $\mathbf{d t}$ is the sampling step size.

## Applying Multiple Methods when Number of Points Is Insufficient

If you do not provide a sufficient number of points for the integration method you choose, NumericIntegration applies the method you choose to all points it can. For the points that remain, NumericIntegration uses the next possible lower-order method.

For example, if you choose Bode Rule as the integration method, Table 2-42 shows how NumericIntegration evaluates the integral for different numbers of data points. If you provide 224 points and choose the Bode Rule method, NumericIntegration arrives at the result by performing 55 Bode Rule method partial evaluations and one Simpson's $3 / 8$ Rule method evaluation.

Table 2-42. Bode Rule Example

| Number of Points | Partial Evaluations Performed |
| :---: | :--- |
| 224 | 55 Bode, 1 Simpson's 3/8 |
| 225 | 56 Bode |
| 226 | 56 Bode, Trapezoidal |
| 227 | 56 Bode, 1 Simpson's |
| 228 | 57 Bode, 1 Simpson's 3/8 |

## Formulas for Integration Methods

For $i=0,1,2, \ldots, \operatorname{int}((\mathbf{n}-1) / k)$, where $\mathbf{n}$ is the number of data points, $k$ is an integer dependent on the method, and $\mathbf{x}$ is the input array, Table 2-43 shows the basic formulas for each of the four integration methods.

Table 2-43. Formulas for Integration Methods

| Integration Method | Formula |
| :--- | :--- |
| Trapezoidal Rule | $(d t / 2) \times\left(x_{i}+x_{i+1}\right) \quad$ for $k=1$ |
| Simpson's Rule | $(d t / 3) \times\left(x_{2 i}+4 x_{2 i+1}+x_{2 i+2}\right) \quad$ for $k=2$ |
| Simpson's 3/8 Rule | $(d t / 8) \times\left(3 x_{3 i}+9 x_{3 i+1}+9 x_{3 i+2}+3 x_{3 i+3}\right) \quad$ for $k=3$ |
| Bode Rule | $(d t / 45) \times\left(14 x_{4 i}+64 x_{4 i+1}+24 x_{4 i+2}+64 x_{4 i+3}+14 x_{4 i+4}\right)$ <br> for $k=4$ |

Each method depends on the sampling interval, $\mathbf{d t}$, and calculates the integral by using successive applications of the basic formula to perform partial evaluations. The number of points each partial evaluation uses represents the order of the method. The result is the sum of these successive partial evaluations.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Array that contains data to integrate. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{d t}$ | integer | Interval size, which represents the sampling <br> step size to use to obtain the data. |
| method | integer | Integration method. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| ir | double | Result of the numeric integration. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

Table 2-44 shows valid method parameter values.
Table 2-44. Valid Integration Method Values

| Integration Method | Value |
| :--- | :---: |
| Trapezoidal Rule | 0 |
| Simpson's Rule | 1 |
| Simpson's 3/8 Rule | 2 |
| Bode Rule | 3 |

## OuterProduct

```
int status = OuterProduct (double x[], int nx, double y[], int ny,
    void *outerProduct);
```


## Purpose

Calculates the outer product of the real input vectors $\mathbf{x}$ and $\mathbf{y}$.
Let $x_{i}$ represent the elements of the $\mathbf{n x}$-element vector $\mathbf{x}$ for $i=0,1,2, \ldots, \mathbf{n x}-1$.
Let $y_{j}$ represent the elements of the ny-element vector $\mathbf{y}$ for $j=0,1,2, \ldots, \mathbf{n y}-1$.
The outer product of these two vectors is a matrix $\mathbf{O}$ of dimensions $\mathbf{n}$-by- $\mathbf{m}$, where the $(i, j)^{\text {th }}$ element of $\mathbf{O}$ is given by:

$$
o_{i, j}=x_{i} \times y_{j}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input real vector $\mathbf{x}$. |
| $\mathbf{n x}$ | integer | Number of elements in $\mathbf{x .}$ |
| $\mathbf{y}$ | double-precision array | Input real vector $\mathbf{y}$. |
| $\mathbf{n y}$ | integer | Number of elements in $\mathbf{y .}$ |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| outerProduct | double-precision <br> 2D array | Calculated outer product matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## PeakDetector

```
int status = PeakDetector (double x[], int n, double threshold,
    int width, int polarity, int initialize,
    int endOfData, int *count, double **locations,
    double **amplitudes, double **secondDerivatives);
```


## Purpose

Finds the location, amplitude, and second derivatives of peaks or valleys in the input array $\mathbf{x}$.
The input data might be a single array or consecutive blocks of data, which are useful when the application involves large data arrays or real-time processing. The initialize and endOfData parameters help you work with consecutive blocks of data. For example, if you have three blocks of data, you can perform peak detection on them according to the following pseudocode:

```
for i = 1 to 3
    Acquire data
    if (i == 1)
        Initialize = True
    else
        Initialize = False
    if (i == 3)
        EndOfData = True
    else
        EndOfData = False
    Set width, threshold, choice
    Call PeakDectector function
    Copy the calculated locations, amplitudes and second derivatives
    to different variables so they won't be overwritten during the next
    iteration of the loop.
continue
```

PeakDetector is based on an algorithm that fits a quadratic polynomial to sequential groups of data points. The width value specifies the number of data points to use. The best choice for the value of width is 3 . Larger widths can reduce the apparent amplitude of peaks and shift the apparent locations.

For each peak or valley, PeakDetect or tests the quadratic fit against the threshold level. PeakDetector ignores peaks with heights lower than the threshold or valleys with troughs higher than the threshold.

You must use the initialize and endOfData parameters to notify PeakDetector when you pass the first and last blocks as the $\mathbf{x}$ parameter so that PeakDetector can initialize and release data internal to the peak detection algorithm.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in input array $\mathbf{x}$. |
| threshold | double | Threshold value to use to reject peaks <br> and valleys that are too small. |
| width | integer | Span, which specifies the number of <br> consecutive data points to use in the <br> quadratic least squares fit. |
| polarity | integer | Pass 0 to detect peaks; pass 1 to <br> detect valleys. |
| initialize | integer | Pass a nonzero value if the current input <br> array is the first data block (or the only <br> data block) to process; otherwise, pass 0. |
| endOfData | integer | Pass a nonzero value if the current input <br> array is the last data block (or the only <br> data block) to process; otherwise, pass 0. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| count | integer | Contains the number of peaks or valleys <br> found in the current block of data. <br> This is the size of the three output arrays: <br> locations, amplitudes, and <br> secondDerivatives. |
| locations | double-precision <br> pointer | Dynamically allocated array that <br> contains the locations of the peaks or <br> valleys PeakDetector finds in the <br> current block of data. |


| Name | Type | Description |
| :--- | :--- | :--- |
| amplitudes | double-precision <br> pointer | Dynamically allocated array that <br> contains the amplitudes of the peaks <br> or valleys PeakDetector finds in the <br> current block of data. |
| secondDerivatives | double-precision <br> pointer | Dynamically allocated array that <br> contains the second derivatives of the <br> peaks or valleys PeakDetector finds in <br> the current block of data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The threshold parameter eliminates the effect of noise in the input data. PeakDetector ignores any peak with a fitted amplitude that is less than threshold and any valley with a fitted amplitude that is greater than threshold.

The width parameter value should not exceed approximately half of the half-width of the peaks or valleys. It can be much smaller for noise-free data.

The elements of the generated locations array represent indices from the beginning of processing, the most recent call to PeakDetector with a nonzero initialize value.

When you no longer need the locations, amplitudes, or secondDerivatives arrays, free them using FreeAnalysisMem.

## PolyEv1D

```
int status = PolyEv1D (double x[], int n, double coef[], int k, double y[]);
```


## Purpose

Performs a polynomial evaluation on the input array. PolyEv1D obtains the $i^{\text {th }}$ element of the output array using the following formula:

$$
y_{i}=\sum_{j=0}^{k-1} \operatorname{coef}_{j} \times x_{i}^{j}
$$

PolyEv1D can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements. |
| coef | double-precision array | Coefficients array. |
| $\mathbf{k}$ | integer | Number of coefficients. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Polynomially evaluated array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The order of the polynomial equals the number of elements in the coefficients array minus one; that is, if there are $\mathbf{k}$ elements in the coef array, then order $=\mathbf{k}-1$.

## Example

```
/* Generate an array with random numbers, let the Ramp function
generate the coefficients { 1, 2, 3, 4, 5} and find the polynomial
evaluation of the array. */
double x[20], y[20], a[5];
double first, last;
int n, k;
n = 20;
k = 5;
first = 1.0;
last = 5.0;
Uniform (n, 17, x);
Ramp (k, first, last, a);
PolyEv1D (x, n, a, k, y);
```


## PolyEv2D

```
int status = PolyEv2D (void *X, int n, int m, double coef[], int k, void *Y);
```


## Purpose

Performs a polynomial evaluation on a 2D input array. PolyEv2D obtains the $(i, j)^{\text {th }}$ element of the output array using the following formula:

$$
y_{i, j}=\sum_{p=0}^{k-1} \operatorname{coef}_{p} \times x_{i, j}^{p}
$$

PolyEv2D can perform the operation in place; that is, $\mathbf{X}$ and $\mathbf{Y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{X}$ | double-precision <br> 2D array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in first dimension. |
| $\mathbf{m}$ | integer | Number of elements in second dimension. |
| $\mathbf{c o e f}$ | double-precision array | Coefficients array. |
| $\mathbf{k}$ | integer | Number of coefficients. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{Y}$ | double-precision <br> 2D array | Polynomially evaluated array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The order of the polynomial equals the number of elements in the coefficients array minus one; that is, if there are $\mathbf{k}$ elements in the coef array, then $\operatorname{order}=\mathbf{k}-1$.

## Example

```
/* Perform a polynomial evaluation of a 2D array, let the Ramp function
generate the coefficients {1, 2, 3, 4, 5} and find the polynomial
evaluation of the array. */
double x[5][10], y[5][10], a[5];
double first, last;
int n, m, k;
n = 5;
k = 5;
m = 10;
first = 1.0;
last = 5.0;
Ramp (k, first, last, a);
PolyEv2D (x, n, m, a, k, y);
```


## PolyFit

```
int status = PolyFit (double x[], double y[], int n, int order, double z[],
    double coef[], double *mse);
```


## Purpose

Finds the coefficients that best represent the polynomial fit of the data points $(x, y)$ using the least squares method. PolyFit obtains the $i^{\text {th }}$ element of the output array using the following formula:

$$
z_{i}=\sum_{n=0}^{o r d e r} \operatorname{coef}_{n} x_{i}^{n}
$$

PolyFit obtains the mean squared error (mse) using the following formula:

$$
m s e=\frac{\sum_{i=0}^{n-1}\left|z_{i}-y_{i}\right|^{2}}{n}
$$

where order is the polynomial order, and $n$ is the number of sample points
If the elements in $\mathbf{x}$ are large and order is also large, you might see unstable results. One solution is to scale the input data elements to the range $[-1: 1]$. To do this, perform the following steps:

1. Find the number, for example, $k$, in $\mathbf{x}$ that has the largest magnitude, or absolute value.
2. Divide all elements in the array by the absolute value of $k$.
3. Apply Polyfit and rescale the results in the output array by multiplying all elements in the output array by the absolute value of $k$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | $x$ values. |
| $\mathbf{y}$ | double-precision array | $y$ values. |
| $\mathbf{n}$ | integer | Number of sample points. |
| order | integer | Polynomial order. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z}$ | double-precision array | Best fit. |
| coef | double-precision array | Polynomial coefficients. |
| mse | double-precision | Mean squared error. |

## $\square$ Note The size of the coefficients array must be order +1.

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a 10th-order polynomial pattern with random coefficients
and find the polynomial fit. */
double x[200], y[200], z[200], a[11], coef[11];
double first, last, mse;
int n, k, order;
n = 200;
first = 0.0;
last = 1.99E2;
Ramp (n, first, last, x) /* x[i] = i */
k = 11;
Uniform (k, 17, a);
PolyEv1D (x, n, a, k, y);/* polynomial pattern */
/* Find the best polynomial fit. */
order = 10;
PolyFit (x, y, n, order, z, coef, &mse);
```


## PolyInterp

```
int status = PolyInterp (double x[], double y[], int n, double x_val,
    double *Interp_Val, double *Error);
```


## Purpose

Calculates the value of the unique polynomial $P$ of degree $\mathbf{n}-1$ passing through the $\mathbf{n}$ points $\left(x_{i}, \mathrm{f}\left(x_{i}\right)\right)$ at $\mathbf{x}_{-}$val, along with an estimate of the error in the interpolation, given a set of $\mathbf{n}$ points $\left(x_{i}, \mathrm{f}\left(x_{i}\right)\right.$ ) in the plane where f is some function and given a value $\mathbf{x}_{-}$val at which f is to be interpolated or extrapolated.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Values at which the function to be <br> interpolated is known. |
| $\mathbf{y}$ | double-precision array | Function values $\mathrm{f}(x)$ at the known $\mathbf{x}$ values. |
| $\mathbf{n}$ | integer | Number of points in $\mathbf{x}$ and in $\mathbf{y}$. |
| $\mathbf{x}$ _val | double-precision | Value at which f is to be interpolated or <br> extrapolated. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| Interp_Val | double-precision | Interpolated or extrapolated value at $\mathbf{x}$ _val. |
| Error | double-precision | Estimate of the error in the interpolation. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Using This Function

All input arrays should be the same size. If the value of $\mathbf{x}$ _val is in the range of $\mathbf{x}$, PolyInterp performs interpolation; otherwise, it performs extrapolation. If $\mathbf{x}_{\mathbf{\prime}} \mathbf{v a l}$ is too far from the range of $\mathbf{x}$, Error might be large, and PolyInterp would not produce a satisfactory extrapolation.

## Example

```
/* Pick points randomly, pick an x in the range of X-values, run a
polynomial through the points, and interpolate at x_val. */
double X[10], Y[10], Interp_Val, Error, x_val, high, low;
int n, i;
n = 10;
WhiteNoise (n, 5.0, 17, X);
WhiteNoise (n, 5.0, 17, Y);
high = X[0];
low = X[0];
for(i=0; i<n; i++) {
    if (X[i] > high) high = X[i];
    if (X[i] < low) low = X[i];
}
x_val = (high + low)/2.0;
PolyInterp (x, y, n, x_val, &Interp_Val, &Error);
```


## PowerFrequencyEstimate

```
int status = PowerFrequencyEstimate (double autoSpectrum[], int n,
    double searchFreq, WindowStruct windowConstants,
    double df, int span, double *freqPeak,
    double *powerPeak);
```


## Purpose

Calculates the estimated power and frequency around a peak in the power spectrum of a time-domain signal. With PowerFrequencyEstimate, you can achieve good frequency estimates for measured peaks that lie between frequency lines on the spectrum. PowerFrequencyEstimate also makes corrections for the window function you use.

PowerFrequencyEstimate calculates the estimated frequency peak using the following formula:

PowerFrequencyEstimate calculates the estimated power peak as follows:

$$
\text { powerPeak }=\frac{\sum_{j=\frac{\sum_{i-\text { span }}^{2}}{2} \text { autoSpectrum }_{j}}^{e n b w}}{\text { ensan }}
$$

where $i=$ index of the searchFreq
$d f$ is the frequency interval, usually in hertz, as output by the AutoPowerSpect rum function
enbw is the equivalent noise bandwidth member of the structure windowConstants as output by the ScaledWindow function

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| autoSpectrum | double-precision array | $\begin{array}{l}\text { Single-sided power spectrum as output } \\ \text { by AutoPowerSpectrum. }\end{array}$ |
| n | integer | $\begin{array}{l}\text { Number of elements in the input } \\ \text { autoSpectrum array. }\end{array}$ |
| searchFreq | double-precision | $\begin{array}{l}\text { Frequency, usually in hertz, of the } \\ \text { frequency around which you want to } \\ \text { estimate the frequency and power. } \\ \text { If searchFreq is less than zero } \\ \text { or is not a valid frequency, } \\ \text { PowerFrequencyEstimate } \\ \text { automatically searches for the } \\ \text { maximum peak in the autoSpectrum } \\ \text { array and estimates the frequency and } \\ \text { power around the maximum peak. }\end{array}$ |
| windowConstants | WindowStruct | $\begin{array}{l}\text { Structure that contains the following } \\ \text { useful constants for the selected window: } \\ \text { enbw is the equivalent noise bandwidth } \\ \text { of the selected window. You can use } \\ \text { this value to calculate the power in a } \\ \text { given frequency span. }\end{array}$ |
| coherentgain is the peak gain of |  |  |
| the window, relative to the peak |  |  |
| gain of the Rectangular window. |  |  |
| PowerFrequencyEstimate uses |  |  |
| this value to normalize peak signal |  |  |
| gains to that of the Rectangular |  |  |
| window. ScaledWindow creates the |  |  |
| windowConstants structure. |  |  |$\}$


| Name | Type | Description |
| :--- | :--- | :--- |
| df | double-precision | Frequency interval, in hertz, as output by <br> AmpPhaseSpectrum, <br> AutoPowerSpectrum, <br> CrossPowerSpectrum, <br> NetworkFunctions, or <br> TransferFunction. |
| span | integer | Number of frequency lines, or bins, <br> around the peak to include in the <br> peak frequency and power estimation. <br> The estimation includes the power in <br> span/2 frequency lines before the peak <br> frequency line, the peak frequency line <br> itself, and span/2 frequency lines after <br> the peak. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| freqPeak | double-precision | Points to the estimated frequency of the <br> estimated peak power in autospectrum. |
| powerPeak | double-precision | Points to the estimated peak power in <br> autospectrum. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Prod1D

```
int status = Prod1D (double x[], int n, double *prod);
```


## Purpose

Finds the product of the $\mathbf{n}$ elements of the input array. Prod1D obtains the product of the elements using the following formula:

$$
\text { prod }=\prod_{i=0}^{n-1} x_{i}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| prod | double-precision | Product of elements. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## PseudoInverse

```
int status = PseudoInverse (void *A, int n, int m, double tolerance, void *B);
```


## Purpose

Calculates the generalized inverse of the real input matrix $\mathbf{A}$. The input matrix can be square or rectangular. The dimensions of the input matrix $\mathbf{A}$ are $\mathbf{n}-\mathrm{by}-\mathbf{m}$. The dimensions of the output matrix (inverse) B are m-by-n.

Note $\quad$ In the case of rectangular matrices with $\mathbf{n}<\mathbf{m}$ (number of rows less than number of columns), take the transpose of the input matrix before you pass it to PseudoInverse. The actual pseudoinverse is then the transpose of the result matrix PseudoInverse calculates.

PseudoInverse uses the Singular Value Decomposition (SVD) technique. Define the pseudoinverse of a scalar $s$ to be $1 / s$ if $s$ does not equal zero, and zero otherwise. Similarly, define the pseudoinverse of a diagonal matrix by transposing the matrix and then taking the scalar pseudoinverse of each entry. If $A^{\dagger}$ denotes the pseudoinverse of a matrix $\mathbf{A}$ whose singular value decomposition is given by:

$$
A=U S V^{T}
$$

then:

$$
A^{\dagger}=U S^{\dagger} V^{T}
$$

where $S^{\dagger}$ is the pseudoinverse of the diagonal matrix $S$ that contains the singular values of $A$
The pseudoinverse exists for both square and rectangular matrices. If the input matrix is square and nonsingular, the pseudoinverse is the same as the general matrix inverse.
$\sqrt{3}$ Note Do not use Pseudo Inverse to calculate the inverse of a square matrix because it takes more time. Use GenInvMatrix instead.

The tolerance parameter must be a small positive number close to machine precision. PseudoInverse sets all singular values of the input matrix smaller than tolerance equal to zero.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | double-precision <br> 2D array | Input real matrix. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in $\mathbf{A}$. |
| tolerance | double-precision | Tolerance value. Refer to the following <br> Parameter Discussion section. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| B | double-precision <br> 2D array | Calculated pseudoinverse matrix. <br> It is $\mathbf{m}$-by- $\mathbf{n}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The value of tolerance determines the level of accuracy in your final solution. Set tolerance close to eps, which is the smallest possible double-precision, floating-point number.

## Pulse

```
int status = Pulse (int n, double amp[], int delay, int width,
    double pulsePattern[]);
```


## Purpose

Generates an array of numbers that represents the pattern of a pulse waveform. Pulse obtains the $i^{\text {th }}$ element of the output array using the formula:

$$
\text { pulsePattern }_{i}= \begin{cases}a m p & \text { if delay } \leq i<(\text { delay }+ \text { width }) \\ 0 & \text { otherwise }\end{cases}
$$

for $i=0,1,2, \ldots, n-1$

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples. |
| amp | double-precision | Pulse amplitude. |
| delay | integer | Pulse delay. |
| width | integer | Pulse width. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| pulsePattern | double-precision array | Pulse pattern array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* The following code generates the following pulse pattern
pulsePattern = {0.0, 0.0, 0.0, 2.0, 2.0, 2.0, 2.0, 2.0, 0.0, 0.0}. */
double pulsePattern[10], amp;
int n, delay, width;
n = 10;
delay = 3;
width = 5;
amp = 2.0;
Pulse (n, amp, delay, width, pulsePattern);
```


## PulseParam

```
int status = PulseParam (double pulsePattern[], int n, double *amp,
    double *amp90, double *amp50, double *amp10,
    double *top, double *base, double *topOvershoot,
    double *baseOvershoot, int *delay, int *width,
    int *riseTime, int *fallTime, double *slewRate);
```


## Purpose

Analyzes the input array values for a pulse pattern and determines the pulse parameters that best describe the pulse pattern. PulseParam assumes that the input array has a bimodal distribution, a distribution that contains two distinct peak values.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| pulsePattern | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| amp | double-precision | Amplitude. |
| amp90 | double-precision | $90 \%$ amplitude. |
| amp50 | double-precision | $50 \%$ amplitude. |
| amp10 | double-precision | $10 \%$ amplitude. |
| top | double-precision | Top value. |
| base | double-precision | Base value. |
| topOvershoot | double-precision | Top overshoot. |
| baseOvershoot | double-precision | Base overshoot. |
| delay | integer | Pulse delay. |
| width | integer | Width delay. |
| riseTime | integer | Rise time. |


| Name | Type | Description |
| :--- | :--- | :--- |
| fallTime | integer | Fall time. |
| slewRate | double-precision | Slew rate. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The returned parameters are as follows:

```
top = upper mode
base = lower mode
amp = top - base
amp90 = 90% amplitude
amp50}= 50% amplitud
amp10 = 10% amplitude
topOvershoot = maximum value - top
baseOvershoot = base - minimum value
delay = rising edge index (50% amplitude)
width = falling edge index (50% amplitude) - delay
riseTime = 90% amplitude index - 10% amplitude index of rising edge
fallTime = 10% amplitude index - 90% amplitude index on falling edge
slewRate = (90% amplitude - 10% amplitude)/riseTime
```

The parameters delay, width, riseTime, and fallTime are integers because the input is a discrete representation of a signal.

## Example

```
/* Generate a noisy pulse pattern and determine its pulse
parameters. */
double x[200], y[200], amp, amp90, amp50, amp10, top, base;
double topOvershoot, baseOvershoot, slewRate, noiseLevel;
int n, delay, width, riseTime, fallTime;
n = 200;
amp = 5.0;
delay = 50;
width = 100;
noiseLevel = 0.5;
Pulse (n, amp, delay, width, x); /* Generate a pulse. */
WhiteNoise (n, noiseLevel, 17, y); /* Generate noise signal. */
Add1D (x, y, n, x); /* Noisy Pulse. */
PulseParam (x, n, &amp, &amp90, &amp50, &amp10, &top, &base,
                &topOvershoot, &baseOvershoot, &delay, &width,
                    &riseTime, &fallTime, &slewRate);
```


## QR

```
int status = QR (void *A, int n, int m, int algorithm, void *Q, void *R);
```


## Purpose

Calculates the QR factorization of the real input matrix $\mathbf{A}$. The input matrix can be square or rectangular.

The following formula defines the QR factorization of a $\mathbf{n}-$ by $-\mathbf{m}$ matrix $\mathbf{A}$ :

$$
A=Q R
$$

where $\quad Q$ is an orthogonal matrix of dimensions $\mathbf{n}$-by-n
$R$ is an upper triangular matrix of dimensions $\mathbf{n}$-by-m
$Q R$ can calculate factorization in many ways. $Q R$ provides three methods for the factorization: Householder, Givens, and Fast Givens. You can use QR factorization to solve linear systems with more equations than unknowns.

## [. 3 Note In the case of rectangular matrices with $\mathbf{n}>\mathbf{m}$ (number of rows greater than

 number of columns), you cannot use the Fast Givens algorithm.
## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | double-precision <br> 2D array | Input real matrix. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in A. |
| algorithm | integer | Algorithm to use. Refer to the following <br> Parameter Discussion section. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{Q}$ | double-precision <br> 2D array | Calculated orthogonal matrix of the <br> QR factorization. |
| $\mathbf{R}$ | double-precision <br> 2D array | Calculated upper triangular matrix of the <br> QR factorization. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

Table 2-45 shows valid algorithm values for the factorization methods.
Table 2-45. Valid Algorithm Values

| Algorithm | Value |
| :--- | :---: |
| Householder | 0 |
| Givens | 1 |
| Fast Givens | 2 |

## QScale1D

```
int status = QScale1D (double x[], int n, double y[], double *scale);
```


## Purpose

Scales the input array by its maximum absolute value. QScale1D can obtain the $i^{t^{\text {th }} \text { element }}$ of the scaled array using the following formula:

$$
y_{i}=\frac{x_{i}}{\text { scale }} \quad \text { where scale is the maximum absolute value in the input array }
$$

QScale1D determines the constant scale.
QScale1D can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Scaled array. |
| scale | double-precision | Scaling constant. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## QScale2D

```
int status = QScale2D (void *X, int n, int m, void *Y, double *scale);
```


## Purpose

Scales a 2D input array by its maximum absolute value. QScale2D can obtain the $(i, j)^{\text {th }}$ element of the scaled array using the following formula:

$$
y_{i, j}=\frac{x_{i, j}}{\text { scale }} \quad \text { where scale is the maximum absolute value of the input array }
$$

QScale2D determines the constant scale.
QScale2D can perform the operation in place; that is, $\mathbf{X}$ and $\mathbf{Y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{X}$ | double-precision <br> 2D array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in first dimension. |
| $\mathbf{m}$ | integer | Number of elements in second dimension. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{Y}$ | double-precision <br> 2D array | Scaled array. |
| scale | double-precision | Scaling constant. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Ramp

```
int status = Ramp (int n, double first, double last, double rampvals[]);
```


## Purpose

Generates an output array that represents a ramp pattern. Ramp obtains the $i^{\text {th }}$ element of the output array using the formula:

$$
\text { rampvals }_{i}=\text { first }+i \Delta x \quad \text { where } \Delta x=\frac{\text { last }- \text { first }}{n-1}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples. |
| first | double-precision | Initial ramp value. |
| last | double-precision | Final ramp value. |

## Output

| Name | Type | Description |
| ---: | :--- | :--- |
| rampvals | double-precision array | Ramp array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

The value of last does not have to be greater than the value of first. If the condition last $<$ first is met, Ramp generates a negatively sloped ramp pattern.

## Example

/* The following code generates the pattern $\{-5.0,-4.0,-3.0,-2.0$, -1.0, 0.0, 1.0, 2.0, 3.0, 4.0, 5.0\}. */
double rampvals[11], first, last;
int n ;
n = 11;
first = -5.0;
last $=5.0$;
Ramp (n, first, last, rampvals);

## RatInterp

```
int status = RatInterp (double x[], double y[], int n, double x_val,
    double *Interp_Val, double *Error);
```


## Purpose

Returns the value of a particular rational function $\mathrm{P}(x) / \mathrm{Q}(x)$ passing through the $\mathbf{n}$ points $\left(x_{i}, \mathrm{f}\left(x_{i}\right)\right)$ at $\mathbf{x} \_$val, given a set of $\mathbf{n}$ points $\left(x_{i}, \mathrm{f}\left(x_{i}\right)\right)$ in the plane where f is some function, and a value $\mathbf{x}_{\mathbf{\prime}}$ val at which f is to be interpolated. $P$ and $Q$ are polynomials, and $\mathbf{n}$ is the number of elements in $\mathbf{x}$.

The function $\mathrm{P}(x) / \mathrm{Q}(x)$ is the unique rational function that passes through the given points and satisfies the following conditions:

$$
\left.\operatorname{deg}(P)=\operatorname{deg}(Q)=\frac{n-1}{2}\right\} \quad \text { if } n \text { is odd }
$$

$$
\left.\begin{array}{c}
\operatorname{deg}(Q)=\frac{n}{2} \\
\operatorname{deg}(P)=\frac{n}{2}-1
\end{array}\right\} \quad \text { if } n \text { is even } \quad \text { where } \operatorname{deg}() \text { is the order of the polynomial function }
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Values at which the function to be <br> interpolated is known. |
| $\mathbf{y}$ | double-precision array | Function values at the known $\mathbf{x}$ values. |
| $\mathbf{n}$ | integer | Number of points in $\mathbf{x}$ and in $\mathbf{y}$. |
| $\mathbf{x}$ _val | double-precision | Value at which the rational function is to be <br> interpolated or extrapolated. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| Interp_Val | double-precision | Interpolated value at $\mathbf{x} \_$val. |
| Error | double-precision | Estimate of the error in the interpolation. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Using This Function

All input arrays should be the same size. If the value of $\mathbf{x}$ _val is in the range of $\mathbf{x}$, Rat Interp performs interpolation; otherwise, it performs extrapolation. If $\mathbf{x}$ _val is too far from the range of $\mathbf{x}$, Error might be large, and Rat Interp would not produce a satisfactory extrapolation.

## Example

```
/* Pick points randomly, pick an x in the range of x-values, run a
rational function through the points and interpolate at x_val. */
double x[10], y[10], Interp_Val, Error, x_val, high, low;
int n, i;
n = 10;
WhiteNoise (n, 5.0, 17, x);
WhiteNoise (n, 5.0, 17, y);
high = x[0];
low = x[0];
for(i=0; i<n; i++) {
    if (x[i] > high) high = x[i];
    if (x[i] < low) low = x[i];
}
x_val = (high + low)/2.0;
RatInterp (x, y, n, x_val, &Interp_Val, &Error);
```


## ReFFT

```
int status = ReFFT (double x[], double y[], int n);
```


## Purpose

Calculates the Fourier Transform of a real input array.

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Array to transform. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Real part of Fast Fourier Transform. |
| $\mathbf{y}$ | double-precision array | Imaginary part of Fast Fourier Transform. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

$\mathbf{n}$ must be a power of two. ReFFT performs the operation in place and overwrites the input array $\mathbf{x}$. The output array $\mathbf{y}$ must be at least the same size as the input array $\mathbf{x}$ because performing an FFT on a real array results in a complex sequence.

## Example

```
/* Generate an array with random numbers and calculate the Fast Fourier
Transform. */
double x[256], y[256];
int n;
n = 256;
Uniform (n, x);
ReFFT (x, y, n);
```


## ReInvFFT

```
int status = ReInvFFT (double x[], double y[], int n);
```


## Purpose

Calculates the inverse Fast Fourier Transform of a complex sequence that results in a real output array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Real part to transform. |
| $\mathbf{y}$ | double-precision array | Imaginary part to transform. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Real inverse Fast Fourier Transform. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

$\mathbf{n}$ must be a power of two. ReInvFFT performs the operation in place and overwrites the input array $\mathbf{x}$. The $\mathbf{y}$ array remains unchanged.

## Example

```
/* Generate an array with random numbers and calculate its real inverse
Fast Fourier Transform. */
double x[256], y[256];
int n;
n = 256;
Uniform (n, 17, x);
Uniform (n, 17, y);
ReInvFFT (x, y, n);
```


## ResetIIRFilter

```
int status = ResetIIRFilter (IIRFilterPtr filterInformation);
```


## Purpose

Sets the reset flag in the filterInfo filter structure so that the internal filter state information is reset to zero before the next cascade IIR filtering operation.

## Parameters

Input

| Name | Type | Description |
| :---: | :--- | :--- |
| filterInformation | IIRFilterPtr | Pointer to the filter structure that contains <br> the filter coefficients and the internal filter <br> information. |
|  |  | Refer to the AllocIIRFilterPtr <br> function description for more information <br> about the filter structure. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* How to use function ResetIIRFilter. */
double fs, fl, fh, x[256], y[256];
int type, order, n;
IIRFilterPtr filterInfo;
n = 256;
fs = 1000.0;
fl = 200.0;
order = 5;
type = 0; /* lowpass */
filterInfo = AllocIIRFilterPtr(type, order);
if(filterInfo!=0) {
    Bw_CascadeCoef(fs, fl, fh, filterInfo);
    Uniform(n, 17, x);
    IIRCascadeFiltering(x, n, filterInfo, y);
    Uniform(n,20,x);
    ResetIIRFilter(filterInfo); /* Reset the filter for a new data
                                    set. */
    IIRCascadeFiltering(x, n, filterInfo, y);
    FreeIIRFilterPtr(filterInfo);
    }
```


## Reverse

```
int status = Reverse (double x[], int n, double y[]);
```


## Purpose

Reverses the order of the elements of the input array using the following formula:

$$
y_{i}=x_{n-i-1} \quad \text { for } i=0,1, \ldots, n-1
$$

Reverse can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Reversed array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## RMS

```
int status = RMS (double x[], int n, double *rmsval);
```


## Purpose

Calculates the root-mean-square (rms) value of the input array. RMS uses the following formula to find the rms value:

$$
r m s=\sqrt{\frac{\sum_{i=0}^{n-1} x_{i}^{2}}{n}}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| rmsval | double-precision | Root-mean-square value. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## SawtoothWave

```
int status = SawtoothWave (int n, double amp, double f, double *phase,
    double x[]);
```


## Purpose

Generates an array that contains a sawtooth wave. SawtoothWave generates the output array $\mathbf{x}$ according to the following formula:

$$
\begin{aligned}
& \qquad x_{i}=a m p \times \operatorname{sawtooth}(p h a s e+f \times 360.0 \times i) \\
& \text { where sawtooth }(p)= \begin{cases}\frac{p \text { modulo } 360.0}{180.0} & 0 \leq p \text { modulo } 360.0<180.0 \\
\frac{p \text { modulo } 360.0}{180.0}-2 & 180.0 \leq p \text { modulo } 360.0<360.0\end{cases}
\end{aligned}
$$

You can use SawtoothWave to simulate a continuous acquisition from a sawtooth wave function generator. The unit of the input phase is in degrees, and Sawt oothWave sets phase to $(\mathbf{p h a s e}+\mathbf{f} \times 360.0 \times \mathbf{n})$ modulo 360.0 before it returns.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples to generate. |
| $\mathbf{a m p}$ | double-precision | Amplitude of the resulting signal. |
| $\mathbf{f}$ | double-precision | Frequency of the resulting signal in <br> normalized units of cycles/sample. |
| phase | double-precision | Pointer to the initial phase, in degrees, of <br> the generated signal. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| phase | double-precision | Upon completion of Sawt oot hWave, phase <br> points to the phase of the next portion of the <br> signal. Use this parameter in the next call to <br> SawtoothWave to simulate a continuous <br> function generator. |
| $\mathbf{x}$ | double-precision array | Contains the generated sawtooth <br> wave signal. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Scale1D

```
int status = Scale1D (double x[], int n, double y[], double *offset,
    double *scale);
```


## Purpose

Scales the input array. The scaled output array is in the range [ $-1: 1$ ]. Scale1D can obtain the $i^{\text {th }}$ element of the scaled array using the following formulas:

$$
\begin{gathered}
y_{i}=\frac{x_{i}-\text { offset }}{\text { scale }} \\
\text { scale }=\frac{\text { max }- \text { min }}{2} \\
\text { offset }=\text { min }+ \text { scale }
\end{gathered}
$$

where $\max$ and $\min$ are the maximum and minimum values in the input array, respectively
Scale1D determines the values of the constants scale and offset. Scale1D can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Scaled array. |
| offset | double-precision | Offsetting constant. |
| scale | double-precision | Scaling constant. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Scale2D

```
int status = Scale2D (void *X, int n, int m, void *Y, double *offset,
    double *scale);
```


## Purpose

Scales the input array. The scaled output array is in the range [ $-1: 1$ ]. Scale2D can obtain the $(i, j)^{t h}$ element of the scaled array using the following formulas:

$$
\begin{gathered}
y_{i, j}=\frac{x_{i, j}-\text { offset }}{\text { scale }} \\
\text { scale }=\frac{\text { max }- \text { min }}{2} \\
\text { offset }=\text { min }+ \text { scale }
\end{gathered}
$$

where $\max$ and $\min$ are the maximum and minimum values in the input array, respectively Scale2D determines the values of the constants scale and offset.

Scale2D can perform the operation in place; that is, $\mathbf{X}$ and $\mathbf{Y}$ can be the same array.

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{X}$ | double-precision <br> 2D array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in first dimension. |
| $\mathbf{m}$ | integer | Number of elements in second dimension. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{Y}$ | double-precision <br> 2D array | Scaled array. |
| offset | double-precision | Offsetting constant. |
| scale | double-precision | Scaling constant. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## ScaledWindow

```
int status = ScaledWindow (double x[], int n, int windowType,
    WindowStruct *windowConstants);
```


## Purpose

Applies a scaled window to the time-domain signal and outputs window constants for further analysis.

The windowed time-domain signal is scaled so that when ScaledWindow calculates the power or amplitude spectrum of the windowed waveform, all windows provide the same level within the accuracy constraints of the window. ScaledWindow also returns important window constants for the window you select. These constants are useful when you use functions that perform computations on the power spectrum, such as PowerFrequencyEstimate.
windowType has the values shown in Table 2-46.
Table 2-46. windowType Values

| Value | Description |
| :---: | :--- |
| 0 | Uniform |
| 1 | Hanning |
| 2 | Hamming |
| 3 | Blackman-Harris |
| 4 | Exact Blackman |
| 5 | Blackman |
| 6 | Flattop |
| 7 | Four Term Blackman-Harris |
| 8 | Seven Term Blackman-Harris |

$\mathbf{x}$ is the time-domain signal multiplied by the scaled window.
windowConstants is a structure that contains the following important constants for the selected window. WindowStruct is defined by the following C typedef statement.

```
typedef struct {
    double enbw;
    double coherentgain;
    } WindowStruct;
```

enbw is the equivalent noise bandwidth of the window you select. You can use this value to calculate the power in a given frequency span.
coherentgain is the peak gain of the window, relative to the peak gain of the Rectangular window. You can use this value to normalize peak signal gains to that of the Rectangular window.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array that contains time-domain <br> signal to window. |
| $\mathbf{n}$ | integer | Number of elements in the input array. |
| windowType | integer | Type of the window function to apply to <br> the input signal. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Windowed version of $\mathbf{x}$. |
| windowConstants | WindowStruct | $\begin{array}{l}\text { Pointer to a structure that contains the } \\ \text { following useful constants for the selected } \\ \text { window: } \\ \text { enbw is the equivalent noise bandwidth } \\ \text { of the selected window. You can use this } \\ \text { value to calculate the power in a given } \\ \text { frequency span. } \\ \text { coherentgain is the peak gain of the } \\ \text { window, relative to the peak gain of the }\end{array}$ |
| Uniform window. ScaledWindow uses |  |  |
| this value to normalize peak signal gains |  |  |
| to that of the Uniform window. |  |  |$\}$

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Set1D

```
int status = Set1D (double x[], int n, double a);
```


## Purpose

Sets the elements of the $\mathbf{x}$ array to a constant value.

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| $\mathbf{a}$ | double-precision | Constant value. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Result array; set to the value of $\mathbf{a}$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Shift

```
int status = Shift (double x[], int n, int shifts, double y[]);
```


## Purpose

Shifts the elements of the input array using the following formula:

$$
y_{i}=x_{i-s h i f t s}
$$

You can specify the number of shifts to be in the positive (right) or negative (left) direction.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| shifts | integer | Number of shifts. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Shifted array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

This is not a circular shift. Shift does not retain the shifted values and replaces the trailing portion of the shift with zero. Shift cannot perform the operation in place; that is, the input and output arrays cannot be the same.

## Example

```
/* Generate an array with random numbers and shift it by 20 samples. */
double x[200], y[200];
int n;
int shifts;
n = 200;
shifts = 20;
Uniform (n, 17, x);
Shift (x, n, shifts, y);
```


## Sinc

```
int status = Sinc (int n, double amp, double delay, double dt, double x[]);
```


## Purpose

Generates an array that contains a sinc pattern. Sinc generates the output array $\mathbf{x}$ according to the following formula:

$$
x_{i}=a m p \times \operatorname{sinc}(i \times d t-\text { delay }) \quad \text { where } \operatorname{sinc}(x)=\frac{\sin (\pi x)}{\pi x}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples to generate. |
| amp | double-precision | Amplitude of the resulting signal. |
| delay | double-precision | Shifts the peak value of the sinc pattern to <br> the index. |
| dt | double-precision | Sampling interval; inversely proportional to <br> the width of the main lobe of the sinc pattern <br> Sinc generates. |

## Output

| Name | Type | Description |
| :---: | :---: | :--- |
| $\mathbf{x}$ | double-precision array | Contains the sinc pattern Sinc generates. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## SinePattern

```
int status = SinePattern (int n, double amp, double phase, double cycles,
    double sine[]);
```


## Purpose

Generates an output array with a sinusoidal pattern. SinePattern obtains the $i^{t h}$ element of the double-precision output array using the following formula:

$$
\operatorname{sine}_{i}=a m p \times \sin \left(\frac{2 \pi i \times \text { cycles }}{n}+\frac{\pi \times \text { phase }}{180.0}\right)
$$

SinePattern assumes the phase value is in degrees and not in radians.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples. |
| amp | double-precision | Amplitude. |
| phase | double-precision | Phase, in degrees. |
| cycles | double-precision | Number of cycles. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| sine | double-precision array | Sinusoidal pattern. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* The following code generates a cosinusoidal pattern. */
double x[8], amp, phase, cycles;
int $n$;
$\mathrm{n}=8$;
amp $=1.0$;
phase $=90.0$;
cycles $=1.5$;
SinePattern (n, amp, phase, cycles, x);

## SineWave

```
int status = SineWave (int n, double amp, double f, double *phase,
    double x[]);
```


## Purpose

Generates an array that contains a sine wave. SineWave generates the $i^{\text {th }}$ element of the output array $\mathbf{x}$ according to the following formula:

$$
x_{i}=\operatorname{amp} \times \sin \left(p h_{i}\right) \quad \text { where } p h_{i}=\frac{\pi}{180.0}(\text { phase }+f \times 360.0 \times i)
$$

where the normalized frequency is the ratio of actual frequency to the sampling frequency
For example, if the actual frequency desired is 100 Hz and the sampling frequency is 500 Hz , the normalized frequency is 0.2 . You can use SineWave to simulate a continuous acquisition from a sine wave function generator. The unit of the input phase is in degrees, and SineWave sets phase to ( $\mathbf{p h a s e}+\mathbf{f} \times 360.0 \times \mathbf{n}$ ) modulo 360.0 before it returns.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples to generate. |
| amp | double-precision | Amplitude of the resulting signal. |
| $\mathbf{f}$ | double-precision | Frequency of the resulting signal in <br> normalized units of cycles/sample. |
| phase | double-precision pointer | Pointer to the initial phase, in degrees, of <br> the sine wave signal SineWave generates. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| phase | double-precision | Upon completion of SineWave, phase <br> points to the phase of the next portion of the <br> signal. Use this parameter in the next call to <br> SineWave to simulate a continuous <br> function generator. |
| $\mathbf{x}$ | double-precision array | Contains the sine wave signal SineWave <br> generates. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Sort

```
int status = Sort (double x[], int n, int direction, double y[]);
```


## Purpose

Sorts the $\mathbf{x}$ input array in ascending or descending order. Sort can perform the operation in place; that is, $\mathbf{x}$ and $\mathbf{y}$ can be the same array.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements to sort. |
| direction | integer | $0=$ ascending <br> nonzero $=$ descending |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Sorted array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random array of numbers and sort them in ascending
order. */
double x[200], y[200];
int n;
int dir;
n = 200;
dir = 0;
Uniform (n, 17, x);
Sort (x, n, dir, y);
```


## SpecialMatrix

```
int status = SpecialMatrix (int matrixType, int m, double x[], int nx,
double y[], int ny, void *Z);
```


## Purpose

Generates a special type of real matrix depending on the value of matrixType. There are five possible matrix types: Identity, Diagonal, Toeplitz, Vandermonde, and Companion. Table 2-47 shows each matrix type and its behavior.

Table 2-47. Matrix Type and Behaviors

| Matrix Type | Behavior |
| :--- | :--- |
| Identity | SpecialMatrix generates an $\mathbf{m}$-by-m identity matrix. |
| Diagonal | SpecialMatrix generates an $\mathbf{n x}$-by-nx diagonal matrix with <br> diagonal elements that are the elements of $\mathbf{x}$. |
| Toeplitz | SpecialMatrix generates an $\mathbf{n x}$-by-ny Toeplitz matrix, which has $\mathbf{x}$ <br> as its first column and $\mathbf{y}$ as its first row. If the first element of $\mathbf{x}$ and $\mathbf{y}$ <br> are different, SpecialMatrix uses the first element of $\mathbf{x}$. |
| Vandermonde | SpecialMatrix generates an $\mathbf{n x}$-by-nx Vandermonde matrix in <br> which the $k^{\text {th }}$ column, for $k=0,1,2, \ldots, \mathbf{n x}-1$, equals the <br> $(\mathbf{n x}-k-1)^{\text {th }}$ power of the elements of $\mathbf{x}$. The elements of a <br> Vandermonde matrix are as follows: <br> $b_{i, j}=x_{i}^{\mathbf{n x}-j-1} \quad$ where $i, j=0,1, \ldots, n x-1$ |
| Companion | SpecialMatrix generates an $\mathbf{n x}-1-b y-\mathbf{n x}-1$ companion matrix. <br> Assuming that the vector $\mathbf{x}$ consists of polynomial coefficients where <br> the first element of $\mathbf{x}$ is the coefficient of the highest order and the last <br> element of $\mathbf{x}$ is the constant term in the polynomial, Specialmatrix <br> constructs the corresponding companion matrix as follows: <br> The first row of the matrix is <br> $b_{0, j-1}=\frac{-x_{j}}{x_{0}} \quad$ for $j=1,2, \ldots, n x-1$ <br> and the remaining rows of the generated matrix form an identity <br> matrix. The eigenvalues of a companion matrix contain the roots of the <br> corresponding polynomial. |

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| matrixType | integer | Type of matrix to generate. Refer to the <br> following Parameter Discussion section. |
| $\mathbf{m}$ | integer | Number of rows and columns to generate <br> when matrixType is Identity matrix. |
| $\mathbf{x}$ | double-precision array | Complex vector used to generate a Diagonal <br> matrix, Toeplitz matrix, Vandermonde <br> matrix, or Companion matrix. |
| $\mathbf{n x}$ | integer | Number of elements in vector $\mathbf{x}$. |
| $\mathbf{y}$ | double-precision array | Second vector to use to generate the <br> Toeplitz matrix. |
| $\mathbf{n y}$ | integer | Number of elements in vector $\mathbf{y}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{Z}$ | integer | Generated matrix. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

Table 2-48 shows valid matrix type values.
Table 2-48. Valid Matrix Type Values

| Matrix Type | Value |
| :--- | :---: |
| Identity matrix | 0 |
| Diagonal matrix | 1 |
| Toeplitz matrix | 2 |
| Vandermonde matrix | 3 |
| Companion matrix | 4 |

## Spectrum

```
int status = Spectrum (double x[], int n);
```


## Purpose

Calculates the power spectrum of the input real data. Spect rum performs the operation in place and overwrites the input array $\mathbf{x}$. Spect rum uses the following formula to obtain the power spectrum $p s$ :

$$
p s=\frac{|\mathrm{FFT}(x)|^{2}}{n^{2}}
$$

n must be a power of two.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Power spectrum. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate an array with random numbers and calculate its power
spectrum. */
double x[256];
int n;
n = 256;
Uniform (n, 17, x);
Spectrum (x, n);
```


## SpectrumUnitConversion

```
int status = SpectrumUnitConversion (double spectrum[], int n, int type,
    int scalingMode, int displayUnits, double df,
    WindowStruct windowConstants,
    double convertedSpectrum[], char unitString[]);
```


## Purpose

Converts the input spectrum, which is the power, amplitude, or gain, to alternate formats, including log, decibels or $d B m$, and spectral density.
spectrum is the input array that contains a spectrum of the type the type selector specifies. type has the values shown in Table 2-49.

Table 2-49. Valid type Values

| Value | Description |
| :---: | :--- |
| 0 | Power (volts rms square); AutoSpect rum calculates |
| 1 | Amplitude (volts, root-mean-square); AmpPhaseSpect rum calculates |
| 2 | Gain (amplitude ratio); TransferFunction calculates |

unitString is a character array that specifies the base unit of the time domain waveform from which SpectrumUnitConversion calculates the input spectrum. The signal unit is often set to " V " (volts). The size of unitString must be at least $(12+$ size of unitString).
scalingMode has three selections for the output unit type, as shown in Table 2-50.
Table 2-50. Valid scalingMode Values

| Value | Description |
| :---: | :--- |
| 0 | Linear |
| 1 | Decibels |
| 2 | dBm |

displayUnit has the selections for the display unit, assuming volts for the base unit, as shown in Table 2-51.

Table 2-51. Valid displayUnit Values

| Value | Description |
| :---: | :--- |
| 0 | $\mathrm{~V}_{\mathrm{rms}}$ (volts, root-mean-square) |
| 1 | $\mathrm{~V}_{\mathrm{pk}}$ (volts peak) |
| 2 | $\mathrm{~V}_{\mathrm{rms}}{ }^{2}$ (volts rms square) |
| 3 | $\mathrm{~V}_{\mathrm{pk}}{ }^{2}$ (volts peak square) |
| 4 | $\mathrm{~V}_{\mathrm{rms}} / \sqrt{\mathrm{Hz}}$ (volts, root-mean-square, per root hertz) |
| 5 | $\mathrm{~V}_{\mathrm{pk}} / \sqrt{\mathrm{Hz}}$ (volts peak per root hertz) |
| 6 | $\mathrm{~V}_{\mathrm{rms}}{ }^{2} / \mathrm{Hz}$ (volts rms square per hertz) |
| 7 | $\mathrm{~V}_{\mathrm{pk}}{ }^{2} / \mathrm{Hz}$ (volts peak square per hertz) |

The last four selections are amplitude spectral density $(4,5)$ and power spectral density $(6,7)$. The structure windowConstants contains constants for the window you select in ScaledWindow. You need this input only when you use the spectral density output formats, or the last four display unit selections.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| spectrum | double-precision array | Input array that contains a spectrum <br> of the type the spectrum selector <br> specifies. It should be a power, <br> amplitude, or gain spectrum. |
| $\mathbf{n}$ | integer | Number of elements in the input <br> spectrum. |
| type | integer | Type of the input spectrum. |
| scalingMode | integer | Type of the scaling of the output <br> spectrum. |
| displayUnits | integer | Unit of the output spectrum, assuming <br> "V" for the input unitString. |


| Name | Type | Description |
| :--- | :--- | :--- |
| df | double-precision | The frequency interval, in hertz, <br> as output by AmpPhaseSpectrum, <br> AutoPowerSpectrum, <br> CrossPowerSpectrum, <br> NetworkFunctions, or <br> TransferFunction. |
| windowConstants | WindowStruct | Structure that contains the following <br> useful constants for the selected <br> window: <br> enbw is the equivalent noise bandwidth <br> of the selected window. You can use <br> this value to calculate the power in a <br> given frequency span. <br> coherentgain is the peak gain of <br> the window, relative to the peak <br> gain of the Rectangular window. <br> SpectrumUnitConversion uses <br> this value normalize peak signal gains <br> to that of the Rectangular window. <br> ScaledWindow outputs this structure. |
| unitString |  | string |
|  |  | String that contains, on input, the base <br> unit of the analyzed signal; "V" for a <br> voltage signal. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| convertedSpectrum | double-precision array | Input spectrum, power, amplitude, or <br> gain, to convert to alternate formats, <br> including log, decibels or dBm, and <br> spectral density. The size of this array <br> must be at least n. |
| unitString | string | Contains, upon completion of <br> Spect rumUnitConversion, the unit <br> of the output convertedSpectrum. <br> The size of this string must be at least <br> $(12+$ size of unitString). |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## SpInterp

```
int status = SpInterp (double x[], double y[], double y2[], int n,
    double x_val, double *Interp_Val);
```


## Purpose

Performs a cubic spline interpolation of the function $f$ at a value $\mathbf{x}_{-}$val, where $\mathbf{x}_{-}$val is in the same range as $x_{i}$, given a tabulated function of the form $y_{i}=\mathrm{f}\left(x_{i}\right)$ for $i=0,1, \ldots, n-1$, with $x<x_{i}+1$, and given the second derivatives that specify the interpolant at the $\mathbf{n}$ nodes of $\mathbf{x}$. The Spline procedure supplies the second derivatives. If $\mathbf{x}$ _val falls in the interval [ $\left.x_{i}, x_{i}+1\right]$, the interpolated value is as follows:

$$
\text { Interp_Val }=A y_{i}+B y_{i+1}+C y^{\prime \prime}{ }_{i}+D y^{\prime \prime}{ }_{i+1}
$$

where $A=\frac{x_{i+1}-x_{-} v a l}{x_{i+1}-x_{i}}$

$$
\begin{aligned}
& B=1-A \\
& C=\frac{\left(A^{3}-A\right)\left(x_{i+1}-x_{i}\right)^{2}}{6} \\
& D=\frac{\left(B^{3}-B\right)\left(x_{i+1}-x_{i}\right)^{2}}{6}
\end{aligned}
$$

$y^{\prime \prime}$ denotes the second derivative of $\mathbf{y}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | x values at which $f$ is known; these values <br> must be in ascending order. |
| $\mathbf{y}$ | double-precision array | Function values $y_{i}=\mathrm{f}\left(x_{i}\right)$. |
| $\mathbf{y 2}$ | double-precision array | Array of second derivatives that specify the <br> interpolant. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}, \mathbf{y}$, and $\mathbf{y} \mathbf{2}$. |
| $\mathbf{x}$ _val | double-precision | x value at which $f$ is to be interpolated. |

## Output

| Name | Type | Description |
| :---: | :--- | :--- |
| Interp_Val | double-precision | Interpolated value. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

/* Choose ascending X-values. Pick corresponding Y-values randomly. Set boundary conditions and specify the cubic spline interpolant to run through the points. Pick an $x$ in the same range as $X$ and interpolate. Pick another $x$ and interpolate again. */ double X[100], Y[100], Y2[100], b1, b2, x_val; int $n$, i;
$\mathrm{n}=100$;
for (i=0; i<n; i++)
$X[i]=i * 0.1 ;$
WhiteNoise (n, 5.0, 17, Y);
b1 $=0.0$;
b2 $=0.0$;
Spline (X, Y, $n, b 1, b 2, ~ Y 2) ;$
x_val $=0.331$;
SpInterp (X, Y, Y2, n, x_val, \&Interp_Val);
x_val = 0.7698;
SpInterp (X, Y, Y2, n, x_val, \&Interp_Val);

## Spline

```
int status = Spline (double x[], double y[], int n, double b1, double b2,
    double y2[]);
```


## Purpose

Calculates the second derivatives used by the cubic spline interpolant, given a tabulated function of the form $y_{i}=\mathrm{f}\left(x_{i}\right)$ for $i=0,1, \ldots, n-1$, with $x_{i}<x_{i}+1$, and given the boundary conditions $\mathbf{b} \mathbf{1}$ and $\mathbf{b} \mathbf{2}$ such that the interpolant's second derivative matches the specified values at $x_{0}$ and $x_{n-1}$.

You can use this array with SpInterp to calculate an interpolation value.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | x values at which $f$ is known; these values <br> must be in ascending order. |
| $\mathbf{y}$ | double-precision array | Function values $y_{i}=\mathrm{f}\left(x_{i}\right)$. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}, \mathbf{y}$, and $\mathbf{y} \mathbf{2}$. |
| $\mathbf{b 1}$ | double-precision | First boundary condition $x^{\prime \prime}{ }_{0}$. |
| $\mathbf{b 2}$ | double-precision | Second boundary condition $\left(x^{\prime \prime}{ }_{n-1}\right)$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y 2}$ | double-precision array | Array of second derivatives that specify the <br> interpolant. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

These second derivatives represent the continuously differentiable curve to run though the n points $\left(x_{i}, y_{i}\right)$.

## Example

```
/* Choose ascending X-values. Pick corresponding Y-values randomly.
Set boundary conditions and specify the cubic spline interpolant to
run through the points. */
double X[100], Y[100], Y2[100], b1, b2;
int n, i;
n = 100;
for(i=0; i<n; i++)
    X[i] = i * 0.1;
WhiteNoise (n, 5.0, 17, Y);
b1=0.0;
b2=0.0;
Spline (X, Y, n, b1, b2, Y2);
```


## SquareWave

```
int status = SquareWave (int n, double amp, double f, double *phase,
    double dutyCycle, double x[]);
```


## Purpose

Generates an array that contains a square wave. SquareWave generates the output array $\mathbf{x}$ according to the following formula:

$$
\begin{aligned}
& x_{i}=a m p \times \operatorname{square}(p h a s e+f \times 360.0 \times i) \quad \text { where } f \text { is normalized frequency } \\
& \text { square }(p)=\left\{\begin{array}{cc}
1.0 & 0 \leq p \text { modulo } 360.0<\frac{\text { duty }}{100.0} \times 360.0 \\
-1.0 & \frac{\text { duty }}{100.0} \times 360.0 \leq p \text { modulo } 360.0<360.0
\end{array}\right.
\end{aligned}
$$

You can use SquareWave to simulate a continuous acquisition from a square wave function generator. The unit of the input phase is in degrees, and SquareWave sets phase to $(\mathbf{p h a s e}+\mathbf{f} \times 360.0 \times \mathbf{n})$ modulo 360.0 before it returns.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples to generate. |
| amp | double-precision | Amplitude of the resulting signal. |
| $\mathbf{f}$ | double-precision | Frequency of the resulting signal in <br> normalized units of cycles/sample. |
| dutyCycle | double-precision | Contains the duty cycle, in percent, of the <br> square wave signal SquareWave generates. |
| phase | double-precision | Points to the initial phase, in degrees, of the <br> square wave signal SquareWave generates. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| phase | double-precision | Upon completion of SquareWave, phase <br> points to the phase of the next portion of the <br> signal. Use this parameter in the next call to <br> SquareWave to simulate a continuous <br> function generator. |
| $\mathbf{x}$ | double-precision array | Contains the square wave signal <br> SquareWave generates. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## StdDev

```
int status = StdDev (double x[], int n, double *meanval, double *sDev);
```


## Purpose

Calculates the standard deviation and the mean, or average, values of the input array. StdDev uses the following formulas to find the mean and the standard deviation:

$$
\begin{gathered}
\text { meanval }=\frac{\sum_{i=0}^{n-1} x_{i}}{n} \\
\text { sDev }=\sqrt{\frac{\sum_{i=0}^{n-1}\left(x_{i}-\text { meanval }\right)^{2}}{n}}
\end{gathered}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| meanval | double-precision | Mean value. |
| sDev | double-precision | Standard deviation. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Sub1D

```
int status = Sub1D (double x[], double y[], int n, double z[]);
```


## Purpose

Subtracts two 1D arrays. Sub1D can obtain the $i^{t h}$ element of the output array using the following formula:

$$
z_{i}=x_{i}-y_{i}
$$

Sub1D can perform the operation in place; that is, $\mathbf{z}$ can be either $\mathbf{x}$ or $\mathbf{y}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | $\mathbf{x}$ input array. |
| $\mathbf{y}$ | double-precision array | $\mathbf{y}$ input array. |
| $\mathbf{n}$ | integer | Number of elements to subtract. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{z}$ | double-precision array | Result array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Sub2D

```
int status = Sub2D (void *X, void *Y, int n, int m, void *Z);
```


## Purpose

Subtracts two 2D arrays. Sub2D obtains the $(i, j)^{t h}$ element of the output array using the formula:

$$
z_{i, j}=x_{i, j}-y_{i, j}
$$

Sub2D can perform the operation in place; that is, $\mathbf{Z}$ can be either $\mathbf{X}$ or $\mathbf{Y}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{X}$ | double-precision <br> 2D array | $\mathbf{X}$ input array. |
| $\mathbf{Y}$ | double-precision <br> 2D array | $\mathbf{Y}$ input array. |
| $\mathbf{n}$ | integer | Number of elements in first dimension. |
| $\mathbf{m}$ | integer | Number of elements in second dimension. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{Z}$ | double-precision <br> 2D array | Result array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Subset1D

```
int status = Subset1D (double x[], int n, int index, int length, double y[]);
```


## Purpose

Extracts a subset of the input array. The output array contains the number of elements you specify by the length. Subset1D starts copying from $\mathbf{x}$ to $\mathbf{y}$ at the index element of $\mathbf{x}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |
| index | integer | Initial index for the subset. |
| length | integer | Number of elements to copy to the subset. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision array | Subset array. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* The following example generates y = {0.0, 1.0, 2.0, 3.0}. */
double x[11], y[4], first, last;
int n, index, length;
n = 11;
index = 5;
length = 4;
first = -5.0;
last = 5.0;
Ramp (n, first, last, x);
Subset1D (x, n, index, length, y);
```


## Sum1D

```
int status = Sum1D (double x[], int n, double *sum);
```


## Purpose

Finds the sum of the elements of the input array. Sum1D obtains the sum of the elements using the following formula:

$$
\text { sum }=\sum_{i=0}^{n-1} x_{i}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| sum | double-precision | Sum of elements. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Generate a random array and sum the elements. */
double x[20], sum;
int n;
n = 20;
Uniform (n, 17, x);
Sum1D (x, n, &sum);
```


## Sum2D

```
int status = Sum2D (void *X, int n, int m, double *sum);
```


## Purpose

Finds the sum of the elements in the input 2D array. Sum2D obtains the sum using the following formula:

$$
s u m=\sum_{i=0}^{n-1} \sum_{j=0}^{m-1} x_{i, j}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{X}$ | double-precision <br> 2D array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in first dimension. |
| $\mathbf{m}$ | integer | Number of elements in second dimension. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| sum | double-precision | Sum of the elements. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## SVD

```
int status = SVD (void *A, int n, int m, void *U, double s[], void *V);
```


## Purpose

Calculates the Singular Value Decomposition (SVD) factorization of the real input matrix $\mathbf{A}$. The input matrix can be square or rectangular.

The following formula defines the SVD factorization of an $\mathbf{n}$-by-m matrix $\mathbf{A}$ :

$$
A=U S V^{T}
$$

where $U$ is an orthogonal matrix of dimensions $\mathbf{n}$-by-m
$V$ is an orthogonal matrix of dimensions $\mathbf{m}$-by-m $S$ is a diagonal matrix of dimensions $\mathbf{m}$-by-m

The diagonal elements of $\mathbf{S}$ are called the singular values of $\mathbf{A}$ and are arranged in descending order. SVD stores them in the output array $\mathbf{s}$. The columns of the output matrices $\mathbf{U}$ and $\mathbf{V}$ are the corresponding singular vectors.

The Singular Value Decomposition is an eigenvalue-like decomposition for rectangular matrices. You can use it to calculate the condition number of a matrix or to solve linear, least square problems. SVD is useful for ill-conditioned or rank-deficient problems because it can detect small singular values.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | double-precision <br> 2D array | Input real matrix. |
| $\mathbf{n}$ | integer | Number of rows in $\mathbf{A}$. |
| $\mathbf{m}$ | integer | Number of columns in $\mathbf{A}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{U}$ | double-precision <br> 2D array | The $\mathbf{n}$-by-m orthogonal matrix <br> SVD factorization generates. |
| $\mathbf{s}$ | double-precision array | Array that contains the singular values <br> of $\mathbf{A}$, in descending order. |
| $\mathbf{V}$ | double-precision <br> 2D array | The $\mathbf{m}$-by-m orthogonal matrix <br> SVD factorization generates. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## SVDS

```
int status = SVDS(void *A, int n, int m, double s[]);
```


## Purpose

SVDS is similar to SVD, but it calculates only the singular values that result from the Singular Value Decomposition factorization of the real input matrix. The input matrix can be square or rectangular.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | double-precision <br> 2D array | Input real matrix. |
| $\mathbf{n}$ | integer | Number of rows in A. |
| $\mathbf{m}$ | integer | Number of columns in $\mathbf{A}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{s}$ | double-precision array | Array that contains the singular values of $\mathbf{A}$, <br> in descending order. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## SymEigenValueVector

```
int status = SymEigenValueVector (void *A, int n, int outputChoice,
    double eigenValues[], void *eigenVectors);
```


## Purpose

Calculates the eigenvalues $\lambda$ and the corresponding eigenvectors $\mathbf{x}$ of a real, symmetric square input matrix $\mathbf{A}$. The following formula defines the eigenvalues and the corresponding eigenvectors:

$$
A x=\lambda x
$$

The eigenvalues and the eigenvectors are all real-valued.
The outputChoice parameter determines what to calculate. Depending on your application, you can choose to calculate just the eigenvalues or to calculate both the eigenvalues and the eigenvectors.

The eigenValues output parameter is a 1D, real array of $\mathbf{n}$ elements. The eigenVectors output parameter is an $\mathbf{n}$-by-n real matrix (2D array). Each $i^{\text {th }}$ column of this matrix is the eigenvector that corresponds to the $i^{\text {th }}$ component of the eigenValues. Each eigenvector is normalized so that its largest component equals one.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{A}$ | double-precision <br> 2D array | Input symmetric square matrix. |
| $\mathbf{n}$ | integer | Number of elements in one dimension of <br> the matrix. |
| outputChoice | integer | Pass 0 for eigenvalues only; 1 for both <br> eigenvalues and eigenvectors. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| eigenValues | double-precision array | Resulting eigenvalues of the input matrix. |
| eigenVectors | double-precision <br> 2D array | Resulting eigenvectors of the input matrix. <br> You can pass NULL if outputChoice is 0. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## T_Dist

```
int status = T_Dist (double t, int n, double *p);
```


## Purpose

Calculates the one-sided probability $\mathbf{p}$ :

$$
p=\operatorname{prob}(T \leq t)
$$

where $T$ is a random variable from the T-distribution with $\mathbf{n}$ degrees of freedom

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{t}$ | double-precision | $-\infty<t<\infty$. |
| $\mathbf{n}$ | integer | Degrees of freedom. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{p}$ | double-precision | Probability $(0 \leq \mathbf{p}<1)$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
double t, p;
int n;
t = -123.456;
n = 6;
T_Dist (t, n, &p);
```


## ToPolar

```
int status = ToPolar (double x, double y, double *mag, double *phase);
```


## Purpose

Converts the rectangular coordinates ( $\mathbf{x}, \mathbf{y}$ ) to polar coordinates (mag, phase). ToPolar obtains the polar coordinates using the following formulas:

$$
\begin{aligned}
\text { mag } & =\sqrt{x^{2}+y^{2}} \\
\text { phase } & =\arctan \left(\frac{y}{x}\right)
\end{aligned}
$$

The phase value is in the range $[-\pi: \pi]$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision | x-coordinate. |
| $\mathbf{y}$ | double-precision | $\mathbf{y}$-coordinate. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| mag | double-precision | Magnitude. |
| phase | double-precision | Phase, in radians. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* Convert the rectangular coordinates to polar coordinates. */
double x, y, mag, phase;
x = 1.5;
y = -2.5;
ToPolar (x, y, &mag, &phase);
```


## ToPolar1D

```
int status = ToPolar1D (double x[], double y[], int n, double mag[],
    double phase[]);
```


## Purpose

Converts the set of rectangular coordinate points $(\mathbf{x}, \mathbf{y})$ to a set of polar coordinate points (mag, phase). ToPolar1D obtains the $i^{\text {th }}$ element of the polar coordinate set using the following formulas:

$$
\begin{aligned}
\operatorname{mag}_{i} & =\sqrt{x_{i}^{2}+y_{i}^{2}} \\
\text { phase }_{i} & =\arctan \left(\frac{y_{i}}{x_{i}}\right)
\end{aligned}
$$

The phase value is in the range $[-\pi: \pi]$.
ToPolar1D can perform the operations in place; that is, $\mathbf{x}$ and $\mathbf{m a g}$, and $\mathbf{y}$ and $\mathbf{p h a s e}$, can be the same arrays, respectively.

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | x-coordinate. |
| $\mathbf{y}$ | double-precision array | $\mathbf{y}$-coordinate. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| mag | double-precision array | Magnitude. |
| phase | double-precision array | Phase, in radians. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## ToRect

```
int status = ToRect (double mag, double phase, double *x, double *y);
```


## Purpose

Converts the polar coordinates (mag, phase) to rectangular coordinates ( $\mathbf{x}, \mathbf{y}$ ). ToRect obtains the rectangular coordinates using the following formulas:

$$
\begin{aligned}
& x=m a g \times \cos (\text { phase }) \\
& y=m a g \times \sin (\text { phase })
\end{aligned}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| mag | double-precision | Magnitude. |
| phase | double-precision | Phase, in radians. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision | x-coordinate. |
| $\mathbf{y}$ | double-precision | $\mathbf{y}$-coordinate. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## ToRect1D

```
int status = ToRect1D (double mag[], double phase[], int n, double x[],
    double y[]);
```


## Purpose

Converts the set of polar coordinate points (mag, phase) to a set of rectangular coordinate points ( $\mathbf{x}, \mathbf{y}$ ). ToRect1D obtains the $i^{i^{h}}$ element of the rectangular set using the following formulas:

$$
\begin{aligned}
& x_{i}=\operatorname{mag}_{i} \times \cos \left(\text { phase }_{i}\right) \\
& y_{i}=\operatorname{mag}_{i} \times \sin \left(\text { phase }_{i}\right)
\end{aligned}
$$

ToRect1D can perform the operations in place; that is, $\mathbf{x}$ and mag, and $\mathbf{y}$ and phase, can be the same arrays, respectively.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| mag | double-precision array | Magnitude. |
| phase | double-precision array | Phase, in radians. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | x-coordinate. |
| $\mathbf{y}$ | double-precision array | $\mathbf{y}$-coordinate. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Trace

```
int status = Trace (void *X, int n, double *traceval);
```


## Purpose

Finds the trace of the 2D input matrix $\mathbf{X}$. The trace is the sum of the matrix elements along the main diagonal. Trace obtains the trace using the following formula:

$$
\text { trace }=\sum_{i=0}^{n-1} x_{i, i}
$$

The input matrix must be an $\mathbf{n}$-by-n square matrix.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{X}$ | double-precision <br> 2D array | Input matrix. |
| $\mathbf{n}$ | integer | Size of matrix. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| traceval | double-precision | Trace. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## TransferFunction

```
int status = TransferFunction (double stimulus[], double response[], int n,
    double dt, double magHf[], double phaseHf[],
    double *df);
```


## Purpose

Calculates the single-sided transfer function, also known as the frequency response, from the time-domain stimulus signal and the time-domain response signal of a network under test.

TransferFunction calculates the transfer function $\mathbf{H f}$ as follows:

$$
H f=\frac{\mathrm{FFT}(\text { response })}{\mathrm{FFT}(\text { stimulus })}
$$

and transforms this result to single-sided magnitude and phase.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| stimulus | double-precision array | Contains the time-domain signal, usually <br> the network stimulus. |
| response | double-precision array | Contains the time-domain signal, usually <br> the network response. |
| $\mathbf{n}$ | integer | Number of elements in the input stimulus <br> and response arrays. $\mathbf{n}$ must be a power of 2. |
| $\mathbf{d t}$ | double-precision | Sampling period of the time-domain <br> signals, usually in seconds. dt $=1 / f s$, <br> where $f s$ is the sampling frequency of the <br> time-domain signals. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| magHf | double-precision array | Magnitude of the averaged single-sided <br> transfer function between the stimulus and <br> response signals. This array must be at least <br> $\mathbf{n} / 2$ elements long. |
| phaseHf | double-precision array | Phase, in radians, of the averaged <br> single-sided transfer function between the <br> stimulus and response signals. This array <br> must be at least $\mathbf{n} / 2$ elements long. |
| df | double-precision | Points to the frequency interval, in hertz, if <br> $\mathbf{d t}$ is in seconds. $\mathbf{d f}=1 /(\mathbf{n} \times \mathbf{d t})$ |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Transpose

```
int status = Transpose (void *x, int n, int m, void *y);
```


## Purpose

Finds the transpose of a 2D input matrix. Transpose obtains the $(i, j)^{\text {th }}$ element of the resulting matrix using the following formula:

$$
y_{i, j}=x_{j, i}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision <br> 2D array | Input matrix. |
| $\mathbf{n}$ | integer | Size of first dimension. |
| $\mathbf{m}$ | integer | Size of second dimension. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{y}$ | double-precision <br> 2D array | Transpose matrix. |

Note If the input matrix has n-by-m dimensions, the output matrix must have m-by-n dimensions.

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Triangle

```
int status = Triangle (int n, double amp, double tri[]);
```


## Purpose

Generates an output array that has a triangular pattern. Triangle obtains the $i^{\text {th }}$ element of the double-precision output array using the following formulas:

$$
\operatorname{tri}_{i}= \begin{cases}\operatorname{amp}\left(\frac{1-|2 i-n|}{n}\right) & \text { if } n \text { is even } \\ \operatorname{amp}\left(\frac{1-|2 i-n+1|}{n-1}\right) & \text { if } n \text { is odd }\end{cases}
$$

## Parameters

## Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples. |
| amp | double-precision | Amplitude. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| tri | double-precision array | Triangular pattern. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
/* The following code generates the pattern tri = {0.0, 1.0, 2.0, 3.0,
4.0, 3.0, 2.0, 1.0}. */
double tri[8], amp;
int n;
n = 8;
amp = 4.0;
Triangle (n, amp, tri);
```


## TriangleWave

```
int status = TriangleWave (int n, double amp, double f, double *phase,
    double x[]);
```


## Purpose

Generates an array that contains a triangle wave. TriangleWave generates the output array $\mathbf{x}$ according to the following formula:

$$
\begin{aligned}
& x_{i}=a m p \times \operatorname{tri}(p h a s e+f \times 360.0 \times i) \\
& \operatorname{tri}(p)= \begin{cases}2 \times \frac{p \text { modulo } 360.0}{180.0} & 0 \leq p \text { modulo } 360.0<90.0 \\
2 \times \frac{1-(p \text { modulo } 360.0)}{180.0} & 90.0 \leq p \text { modulo } 360.0<270.0 \\
2 \times \frac{p \text { modulo } 360.0}{180.0-2.0} & 270.0 \leq p \text { modulo } 360.0<360.0\end{cases}
\end{aligned}
$$

You can use TriangleWave to simulate a continuous acquisition from a triangle wave function generator. The unit of the input phase is in degrees, and TriangleWave sets phase to $(\mathbf{p h a s e}+\mathbf{f} \times 360.0 \times \mathbf{n})$ modulo 360.0 before it returns.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples to generate. |
| amp | double-precision | Amplitude of the resulting signal. |
| $\mathbf{f}$ | double-precision | Frequency of the resulting signal in <br> normalized units of cycles/sample. |
| phase | double-precision | Points to the initial phase, in degrees, of the <br> triangle wave signal TriangleWave <br> generates. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| phase | double-precision | Upon completion of TriangleWave, phase <br> points to the phase of the next portion of the <br> signal. Use this parameter in the next call to <br> TriangleWave to simulate a continuous <br> function generator. |
| $\mathbf{x}$ | double-precision array | Contains the triangle wave signal <br> TriangleWave generates. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## TriWin

```
int status = TriWin (double x[], int n);
```


## Purpose

Applies a triangular window to the $\mathbf{x}$ input signal. The following formula defines the triangular window:

$$
w_{i}=\frac{1-|2 \times i-n|}{n} \quad \text { for } i=0,1, \ldots, n-1
$$

TriWin obtains the output signal using the following formula:

$$
x_{i}=x_{i} \times w_{i} \quad \text { for } i=0,1, \ldots, n-1
$$

TriWin performs the window operation in place; that is, the windowed data $\mathbf{x}$ replaces the input data $\mathbf{x}$.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input data. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Windowed data. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Uniform

```
int status = Uniform (int n, int seed, double x[]);
```


## Purpose

Generates an array of random numbers that are uniformly distributed between zero and one.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples. |
| seed | integer | Seed value. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Random pattern between 0 and 1. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

When seed $\geq 0$, Uniform generates a new random sequence using the seed value.
When seed $<0$, the previously generated random sequence continues.

## Example

```
/* The following code generates an array of random numbers between
0 and 1. */
double x[20];
int n;
n = 20;
Uniform (n, 17, x);
```


## UnWrap1D

```
int status = UnWrap1D (double phase[], int n);
```


## Purpose

Unwraps the discontinuous phase values that are in the range $[-\pi: \pi]$ to create continuous values. UnWrap1D overwrites the input array phase.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| phase | double-precision array | Array of discontinuous phase values. |
| $\mathbf{n}$ | integer | Number of elements. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| phase | double-precision array | Array of continuous phase values. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Variance

```
int status = Variance (double x[], int n, double *meanval, double *var);
```


## Purpose

Calculates the variance and the mean, or average, values of the input array. Variance uses the following formulas to find the mean and the variance:

$$
\begin{aligned}
& \text { meanval }=\frac{\sum_{i=0}^{n-1} x_{i}}{n} \\
& \text { var }=\frac{\sum_{i=0}^{n-1}\left(x_{i}-{\text { meanval })^{2}}_{n}^{n}\right.}{n}
\end{aligned}
$$

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision array | Input array. |
| $\mathbf{n}$ | integer | Number of elements in $\mathbf{x}$. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| meanval | double-precision | Mean value. |
| var | double-precision | Variance. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## WhiteNoise

```
int status = WhiteNoise (int n, double amp, int seed, double *noise[]);
```


## Purpose

Generates an array of random numbers that are uniformly distributed between -amp and amp.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{n}$ | integer | Number of samples. |
| amp | double-precision | Amplitude. |
| seed | integer | Seed value. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| noise | double-precision array | Noise pattern. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

When seed $\geq 0$, WhiteNoise generates a new random sequence using the seed value.
When seed $<0$, the previously generated random sequence continues.

## Example

```
/* The following code generates an array of random numbers between
-5 and 5. */
double x[20], amp;
int n;
n = 20;
amp = 5.0;
WhiteNoise (n, amp, 17, x);
```


## Wind_BPF

```
int status = Wind_BPF (double fs, double fl, double fh, int n, double coef[],
    int windType);
```


## Purpose

Designs a digital bandpass FIR linear phase filter using a windowing technique. Five windows are available. Wind_BPF generates only the filter coefficients; it does not actually perform data filtering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f l}$ | double-precision | Lower cutoff frequency. |
| $\mathbf{f h}$ | double-precision | Higher cutoff frequency. |
| $\mathbf{n}$ | integer | Number of filter coefficients. |
| windType | integer | Window type. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Filter coefficients. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

windType selects one of the five windows as shown in Table 2-52.
Table 2-52. Valid windType Values

| windType | Window | Attenuation (dB) | Transition <br> Bandwidth (fs/n) |
| :---: | :--- | :---: | :---: |
| 1 | Rectangular | 21 | 0.9 |
| 2 | Triangular | 25 | 1.18 |
| 3 | Hanning | 44 | 2.5 |
| 4 | Hamming | 53 | 3.13 |
| 5 | Blackman | 74 | 4.6 |

## Using This Function

The attenuation value determines the approximate peak value of the sidelobes. Transition bandwidth determines a frequency range over which the filter response changes from the pass band to the stop band or from the stop band to the pass band. For more information, refer to Discrete-Time Signal Processing by Oppenheim and Schafer.

## Example

```
/* Design a 55-point bandpass FIR linear phase filter that can achieve
at least a 44 dB attenuation and filter the incoming signal with the
designed filter. */
double x[256], coef[55], y[310], fs, fl, fh;
int n, m, windType;
fs = 1000.0; /* sampling frequency */
fl = 200.0; /* desired lower cutoff frequency */
fh = 300.0; /* desired higher cutoff frequency */
/* pass band is from 200.0 to 300.0 */
n = 55; /* filter length */
windType = 3; /* using Hanning window */
m = 256;
Wind_BPF (fs, fl, fh, n, coef, windType);
Convolve (coef, n, x, m, y); /* Convolve the filter with the signal. */
```


## Wind_BSF

```
int status = Wind_BSF (double fs, double fl, double fh, int n, double coef[],
    int windType);
```


## Purpose

Designs a digital bandstop FIR linear phase filter using a windowing technique. Five windows are available. Wind_BSF generates only the filter coefficients; it does not actually perform data filtering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| $\mathbf{f l}$ | double-precision | Lower cutoff frequency. |
| $\mathbf{f h}$ | double-precision | Higher cutoff frequency. |
| $\mathbf{n}$ | integer | Number of filter coefficients. |
| windType | integer | Window type. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Filter coefficients. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

windType selects one of the five windows as shown in Table 2-53.
Table 2-53. Valid windType Values

| windType | Window | Attenuation (dB) | Transition <br> Bandwidth (fs/n) |
| :---: | :--- | :---: | :---: |
| 1 | Rectangular | 21 | 0.9 |
| 2 | Triangular | 25 | 1.18 |
| 3 | Hanning | 44 | 2.5 |
| 4 | Hamming | 53 | 3.13 |
| 5 | Blackman | 74 | 4.6 |

## Using This Function

The attenuation value determines the approximate peak value of the sidelobes. Transition bandwidth determines a frequency range over which the filter response changes from the pass band to the stop band or from the stop band to the pass band. For more information, refer to Discrete-Time Signal Processing by Oppenheim and Schafer.

## Example

```
/* Design a 55-point bandstop FIR linear phase filter that can achieve
at least a 44 dB attenuation and filter the incoming signal with the
designed filter. */
double x[256], coef[55], y[310], fs, fl, fh;
int n, m, windType;
fs = 1000.0; /* sampling frequency */
fl = 200.0; /* desired lower cutoff frequency */
fh = 300.0; /* desired higher cutoff frequency */
    /* stop band is from 200.0 to 300.0 */
    n = 55; /* filter length */
    windType = 3; /* using Hanning window */
    m = 256;
    Wind_BSF (fs, fl, fh, n, coef, windType);
    Convolve (coef, n, x, m, y); /* Convolve the filter with the signal. */
```


## Wind_HPF

```
int status = Wind_HPF (double fs, double fc, int n, double coef[],
    int windType);
```


## Purpose

Designs a digital highpass FIR linear phase filter using a windowing technique. Five windows are available. Wind_HPF generates only the filter coefficients; it does not actually perform data filtering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| fs | double-precision | Sampling frequency. |
| fc | double-precision | Cutoff frequency. |
| $\mathbf{n}$ | integer | Number of filter coefficients. |
| windType | integer | Window type. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Filter coefficients. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

windType selects one of the five windows as shown in Table 2-54.
Table 2-54. Valid windType Values

| windType | Window | Attenuation (dB) | Transition <br> Bandwidth (fs/n) |
| :---: | :--- | :---: | :---: |
| 1 | Rectangular | 21 | 0.9 |
| 2 | Triangular | 25 | 1.18 |
| 3 | Hanning | 44 | 2.5 |
| 4 | Hamming | 53 | 3.13 |
| 5 | Blackman | 74 | 4.6 |

## Using This Function

The attenuation value determines the approximate peak value of the sidelobes. Transition bandwidth determines a frequency range over which the filter response changes from the pass band to the stop band or from the stop band to the pass band. For more information, refer to Discrete-Time Signal Processing by Oppenheim and Schafer.

## Example

```
/* Design a 55-point highpass FIR linear phase filter that can achieve
at least a 44 dB attenuation and filter the incoming signal with the
designed filter. */
double x[256], coef[55], y[310], fs, fc;
int n, m, windType;
fs = 1000.0; /* sampling frequency */
fc = 200.0; /* desired cutoff frequency */
n = 55; /* filter length */
windType = 3; /* using Hanning window */
m = 256;
Wind_HPF (fs, fc, n, coef, windType);
Convolve (coef, n, x, m, y); /* Convolve the filter with the signal. */
```


## Wind_LPF

```
int status = Wind_LPF (double fs, double fc, int n, double coef[],
    int windType);
```


## Purpose

Designs a digital lowpass FIR linear phase filter using a windowing technique. Five windows are available. Wind_LPF generates only the filter coefficients; it does not actually perform data filtering.

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{f s}$ | double-precision | Sampling frequency. |
| fc | double-precision | Cutoff frequency. |
| $\mathbf{n}$ | integer | Number of filter coefficients. |
| windType | integer | Window type. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| coef | double-precision array | Filter coefficients. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Parameter Discussion

windType selects one of the five windows as shown in Table 2-55.
Table 2-55. Valid windType Values

| windType | Window | Attenuation (dB) | Transition <br> Bandwidth (fs/n) |
| :---: | :--- | :---: | :---: |
| 1 | Rectangular | 21 | 0.9 |
| 2 | Triangular | 25 | 1.18 |
| 3 | Hanning | 44 | 2.5 |
| 4 | Hamming | 53 | 3.13 |
| 5 | Blackman | 74 | 4.6 |

## Using This Function

The attenuation value determines the approximate peak value of the sidelobes. Transition bandwidth determines a frequency range over which the filter response changes from the pass band to the stop band or from the stop band to the pass band. For more information, refer to Discrete-Time Signal Processing by Oppenheim and Schafer.

## Example

```
/* Design a 55-point lowpass FIR linear phase filter that can achieve
at least a 44 dB attenuation and filter the incoming signal with the
designed filter. */
double x[256], coef[55], y[310], fs, fc;
int n, m, windType;
fs = 1000.0; /* sampling frequency */
fc = 200.0; /* desired cutoff frequency */
n = 55; /* filter length */
windType = 3; /* using Hanning window */
m = 256;
Wind_LPF (fs, fc, n, coef, windType);
Convolve (coef, n, x, m, y); /* Convolve the filter with the signal. */
```


## XX_Dist

```
int status = XX_Dist (double x, int n, double *p);
```


## Purpose

Approximates the one-sided probability $\mathbf{p}$ :

$$
p=\operatorname{prob}(X \leq x)
$$

where $X$ is a random variable from the $\chi^{2}$-distribution with $\mathbf{n}$ degrees of freedom

## Parameters

Input

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{x}$ | double-precision | $-\infty<x<\infty$. |
| $\mathbf{n}$ | integer | Degrees of freedom. |

## Output

| Name | Type | Description |
| :--- | :--- | :--- |
| $\mathbf{p}$ | double-precision | Probability $(0 \leq \mathbf{p}<1)$. |

## Return Value

| Name | Type | Description |
| :--- | :--- | :--- |
| status | integer | Refer to Appendix A for error codes. |

## Example

```
double x, p;
int n;
x = -123.456;
n = 6;
XX_Dist (x, n, &p);
/* Now p = O because chi-square distributed variables are
non-negative. */
```


## Error Codes

This appendix contains error codes the Advanced Analysis Library functions return. If an error condition occurs during a call to any of the functions in the LabWindows Analysis Library, the return value status contains the returned error code. This code is a value that specifies the type of error that occurred. Table A-2 lists the error codes in numeric order. For your convenience, Table A-1 lists the error codes alphabetically by symbolic name.

Table A-1. Advanced Analysis Library Error Codes, Sorted Alphabetically

| Symbolic Name | Code | Error Message |
| :--- | :--- | :--- |
| ArraySizeAnlysErr | -20008 | Specified conditions on the input arrays have <br> not been met. |
| AttenGTRippleAnlysErr | -20028 | Attenuation must be greater than the ripple <br> amplitude. |
| AttenGTZeroAnlysErr | -20025 | Attenuation must be greater than zero. |
| BalanceAnlysErr | -20047 | Data is unbalanced. |
| BandSpecAnlysErr | -20023 | Invalid band specification. |
| BaseGETopAnlysErr | -20101 | Base must be less than top. |
| BetaFuncAnlysErr | -20057 | Parameter to the beta function must meet the <br> condition $0<p<1$. |
| CategoryAnlysErr | -20055 | Invalid number of categories or samples. |
| ColumnAnlysErr | -20051 | First column in the X matrix must be all ones. |
| CyclesAnlysErr | -20012 | Number of cycles must meet the condition <br> $0<$ cycles $\leq$ samples. |
| DataAnlysErr | -20045 | Total number of data points <br> must be equal to product of <br> $($ levels/each factor) $\times$ (observations/cell). |
| DecFactAnlysErr | -20022 | Decimating factor must meet the condition <br> $0<$ decimating factor $\leq$ samples. |

Table A-1. Advanced Analysis Library Error Codes, Sorted Alphabetically (Continued)

| Symbolic Name | Code | Error Message |
| :--- | :--- | :--- |
| DelayWidthAnlysErr | -20014 | Delay and width must meet the condition <br> $0 \leq($ delay + width) < samples. |
| DimensionAnlysErr | -20058 | Invalid number of dimensions or dependent <br> variables. |
| DistinctAnlysErr | -20049 | x-values must be distinct. |
| DivByZeroAnlysErr | -20060 | Divide by zero. |
| DtGTZeroAnlysErr | -20016 | dt or dx must be greater than zero. |
| EqRplDesignAnlysErr | -20031 | Filter cannot be designed with the specified <br> input parameters. |
| EqSamplesAnlysErr | -20002 | Input sequences must be the same size. |
| EvenSizeAnlysErr | -20033 | Number of coefficients must be odd for <br> this filter. |
| FactorAnlysErr | -20043 | Level of factor is outside the allowable range. |
| FreedomAnlysErr | -20052 | Invalid degrees of freedom. |
| IndexLengthAnlysErr | -20018 | Index and length must meet the condition <br> $0 \leq$ (index + length) < samples. |
| MoAnlysErr | -20017 | Index must meet the condition <br> $0 \leq$ index < samples. |
| IndexLTSamplesAnlysErr | -20062 | Naximum iteration exceeded. |
| ModelAnlysErr | InvSelectionAnlysErr | -20061 |
| Invalid selection. |  |  |
| MaxfilterInfoAnlysErr | -20066 | Information in the IIR filter structure <br> is invalid. |
| MixedSignAnlysErr | -20036 | Second array must be all positive or negative <br> and nonzero. |
| allowable range. |  |  |

Table A-1. Advanced Analysis Library Error Codes, Sorted Alphabetically (Continued)

| Symbolic Name | Code | Error Message |
| :--- | :--- | :--- |
| NyquistAnlysErr | -20020 | Cut-off frequency, fc, must meet the condition <br> $0 \leq f c \leq(f s / 2)$. |
| ObservationsAnlysErr | -20044 | There must be at least one observation. |
| OddSizeAnlysErr | -20034 | Number of coefficients must be even for <br> this filter. |
| OrderGEZeroAnlysErr | -20103 | Order must be greater than or equal to zero. |
| OrderGTZeroAnlysErr | -20021 | Order must be greater than zero. |
| OutOfMemAnlysErr | -20001 | There is not enough memory left to perform <br> the specified routine. |
| PoleAnlysErr | -20050 | Interpolating function has a pole at the <br> requested value. |
| PolyAnlysErr | -20063 | Invalid polynomial. |
| PowerOfTwoAnlysErr | -20009 | Size of the input array must be a valid power <br> of two: size $=2^{m}$. |
| ProbabilityAnlysErr | -20053 | Probability must meet the condition 0<p<1. |
| RippleGTZeroAnlysErr | -20024 | Ripple must be greater than zero. |
| SamplesGEThreeAnlysErr | -20007 | Number of samples must be greater than or <br> equal to three. |
| SquareMatrixAnlysErr | -20040 | Input matrix must be a square matrix. |
| SamplesGETwoAnlysErr | -20006 | Number of samples must be greater than or <br> equal to two. |
| SamplesGEZeroAnlysErr | -20004 | Number of samples must be greater than or <br> equal to zero. |
| ShiftRangeAnlysErr | -20102 | Shifts must meet the condition |
| shifts\| < samples. |  |  |

Table A-1. Advanced Analysis Library Error Codes, Sorted Alphabetically (Continued)

| Symbolic Name | Code | Error Message |
| :--- | :---: | :--- |
| TableAnlysErr | -20056 | Contingency table has a negative number. |
| UpperGELowerAnlysErr | -20019 | Upper value must be greater than or equal to <br> the lower value. |
| ZeroVectorAnlysErr | -20065 | Elements of the vector cannot be all zero. |

Table A-2. Advanced Analysis Library Error Codes, Sorted Numerically

| Code | Symbolic Name | Error Message |
| :---: | :--- | :--- |
| 0 | NoAnlysErr | No error; the call was successful. |
| -20001 | OutOfMemAnlysErr | There is not enough memory left to perform <br> the specified routine. |
| -20002 | EqSamplesAnlysErr | Input sequences must be the same size. |
| -20003 | SamplesGTZeroAnlysErr | Number of samples must be greater than zero. |
| -20004 | SamplesGEZeroAnlysErr | Number of samples must be greater than or <br> equal to zero. |
| -20006 | SamplesGETwoAnlysErr | Number of samples must be greater than or <br> equal to two. |
| -20007 | SamplesGEThreeAnlysErr | Number of samples must be greater than or <br> equal to three. |
| -20008 | ArraySizeAnlysErr | Specified conditions on the input arrays have <br> not been met. |
| -20009 | PowerOfTwoAnlysErr | Size of the input array must be a valid power <br> of two: size $=2^{m}$. |
| -20012 | CyclesAnlysErr | Number of cycles must meet the condition <br> $0<$ cycles $\leq$ samples. |
| -20014 | DelayWidthAnlysErr | Delay and width must meet the condition <br> $0 \leq($ delay + width $<$ samples. |
| -20016 | DtGTZeroAnlysErr | dt or dx must be greater than zero. |
| -20017 | IndexLTSamplesAnlysErr | Index must meet the condition: <br> $0 \leq$ index < samples. |

Table A-2. Advanced Analysis Library Error Codes, Sorted Numerically (Continued)

| Code | Symbolic Name | Error Message |
| :---: | :--- | :--- |
| -20018 | IndexLengthAnlysErr | Index and length must meet the condition <br> $0 \leq($ index + length < samples. |
| -20019 | UpperGELowerAnlysErr | Upper value must be greater than or equal to <br> the lower value. |
| -20020 | NyquistAnlysErr | Cut-off frequency, $f c$, must meet the <br> condition: $0 \leq f c \leq(f s / 2)$. |
| -20021 | OrderGTZeroAnlysErr | Order must be greater than zero. |
| -20022 | DecFactAnlysErr | Decimating factor must meet the condition <br> $0<$ decimating factor $\leq$ samples. |
| -20023 | BandSpecAnlysErr | Invalid band specification. |
| -20024 | RippleGTZeroAnlysErr | Ripple must be greater than zero. |
| -20025 | AttenGTZeroAnlysErr | Attenuation must be greater than zero. |
| -20028 | AttenGTRippleAnlysErr | Attenuation must be greater than the ripple <br> amplitude. |
| -20031 | EqRplDesignAnlysErr | Filter cannot be designed with the specified <br> input parameters. |
| -20033 | EvenSizeAnlysErr | Number of coefficients must be odd for <br> this filter. |
| -20044 | ObservationsAnlysErr | There must be at least one observation. |
| -20036 | OddSizeAnlysErr | Number of coefficients must be even for <br> this filter. |
| -20041 | SingularMatrixAnlysErr | Input matrix is singular. The system of <br> equations cannot be solved. |
| -20042 | LevelsAnlysErr | Sumber of levels is outside the <br> allowable range. |
| SquareMatrixAnlysErr |  |  |
| and nonzero. |  |  |

Table A-2. Advanced Analysis Library Error Codes, Sorted Numerically (Continued)

| Code | Symbolic Name | Error Message |
| :---: | :---: | :---: |
| -20045 | DataAnlysErr | Total number of data points must be equal to product of (levels/each factor) $\times$ (observations/cell). |
| -20047 | BalanceAnlysErr | Data is unbalanced. |
| -20048 | ModelAnlysErr | Random Effect model was requested when the Fixed Effect model is required. |
| -20049 | DistinctAnlysErr | x -values must be distinct. |
| -20050 | PoleAnlysErr | Interpolating function has a pole at the requested value. |
| -20051 | ColumnAnlysErr | First column in the X matrix must be all ones. |
| -20052 | FreedomAnlysErr | Invalid degrees of freedom. |
| -20053 | ProbabilityAnlysErr | Probability must meet the condition $0<p<1$. |
| -20055 | CategoryAnlysErr | Invalid number of categories or samples. |
| -20056 | TableAnlysErr | Contingency table has a negative number. |
| -20057 | BetaFuncAnlysErr | Parameter to the beta function must meet the condition $0<p<1$. |
| -20058 | DimensionAnlysErr | Invalid number of dimensions or dependent variables. |
| -20060 | DivByZeroAnlysErr | Divide by zero. |
| -20061 | InvSelectionAnlysErr | Invalid selection. |
| -20062 | MaxIterAnlysErr | Maximum iteration exceeded. |
| -20063 | PolyAnlysErr | Invalid polynomial. |
| -20065 | ZeroVectorAnlysErr | Elements of the vector cannot be all zero. |
| -20066 | IIRFilterInfoAnlysErr | Information in the IIR filter structure is invalid. |
| -20101 | BaseGETopAnlysErr | Base must be less than top. |

Table A-2. Advanced Analysis Library Error Codes, Sorted Numerically (Continued)

| Code | Symbolic Name | Error Message |
| :---: | :--- | :--- |
| -20102 | ShiftRangeAnlysErr | Shifts must meet the condition <br> $\mid$ shifts $\mid<$ samples. |
| -20103 | OrderGEZeroAnlysErr | Order must be greater than or equal to zero. |

## Customer Communication

For your convenience, this appendix contains forms to help you gather the information necessary to help us solve your technical problems and a form you can use to comment on the product documentation. When you contact us, we need the information on the Technical Support Form and the configuration form, if your manual contains one, about your system configuration to answer your questions as quickly as possible.

National Instruments has technical assistance through electronic, fax, and telephone systems to quickly provide the information you need. Our electronic services include a bulletin board service, an FTP site, a fax-on-demand system, and e-mail support. If you have a hardware or software problem, first try the electronic support systems. If the information available on these systems does not answer your questions, we offer fax and telephone support through our technical support centers, which are staffed by applications engineers.

## Electronic Services

## Bulletin Board Support

National Instruments has BBS and FTP sites dedicated for 24-hour support with a collection of files and documents to answer most common customer questions. From these sites, you can also download the latest instrument drivers, updates, and example programs. For recorded instructions on how to use the bulletin board and FTP services and for BBS automated information, call 512795 6990. You can access these services at:

United States: 5127945422
Up to 14,400 baud, 8 data bits, 1 stop bit, no parity
United Kingdom: 01635551422
Up to 9,600 baud, 8 data bits, 1 stop bit, no parity
France: 0148651559
Up to 9,600 baud, 8 data bits, 1 stop bit, no parity

## FTP Support

To access our FTP site, log on to our Internet host, ftp. natinst.com, as anonymous and use your Internet address, such as joesmith@anywhere.com, as your password. The support files and documents are located in the / support directories.

## Fax-on-Demand Support

Fax-on-Demand is a 24 -hour information retrieval system containing a library of documents on a wide range of technical information. You can access Fax-on-Demand from a touch-tone telephone at 5124181111 .

## E-Mail Support (Currently USA Only)

You can submit technical support questions to the applications engineering team through e-mail at the Internet address listed below. Remember to include your name, address, and phone number so we can contact you with solutions and suggestions.

```
support@natinst.com
```


## Telephone and Fax Support

National Instruments has branch offices all over the world. Use the list below to find the technical support number for your country. If there is no National Instruments office in your country, contact the source from which you purchased your software to obtain support.

| Country | Telephone | Fax |
| :--- | :--- | :--- |
| Australia | 0398795166 | 0398796277 |
| Austria | 06624579900 | 066245799019 |
| Belgium | 027570020 | 027570311 |
| Brazil | 012883336 | 0112888528 |
| Canada (Ontario) | 9057850085 | 9057850086 |
| Canada (Quebec) | 5146948521 | 5146944399 |
| Denmark | 45762600 | 45762602 |
| Finland | 0972572511 | 0972572555 |
| France | 0148142424 | 0148142414 |
| Germany | 0897413130 | 0897146035 |
| Hong Kong | 26453186 | 26868505 |
| Israel | 036120092 | 036120095 |
| Italy | 02413091 | 0241309215 |
| Japan | 0354722970 | 0354722977 |
| Korea | 025967456 | 025967455 |
| Mexico | 55202635 | 55203282 |
| Netherlands | 0348433466 | 0348430673 |
| Norway | 32848400 | 32848600 |
| Singapore | 2265886 | 2265887 |
| Spain | 916400085 | 916400533 |
| Sweden | 087304970 | 087304370 |
| Switzerland | 0562005151 | 0562005155 |
| Taiwan | 023771200 | 027374644 |
| United Kingdom | 01635523545 | 01635523154 |
| United States | 5127958248 | 5127945678 |
|  |  |  |

## Technical Support Form

Photocopy this form and update it each time you make changes to your software or hardware, and use the completed copy of this form as a reference for your current configuration. Completing this form accurately before contacting National Instruments for technical support helps our applications engineers answer your questions more efficiently.

If you are using any National Instruments hardware or software products related to this problem, include the configuration forms from their user manuals. Include additional pages if necessary.

Name $\qquad$
Company $\qquad$
Address $\qquad$

Fax (__ ) ___ Phone (
$\qquad$
Operating system (include version number)
Clock speed ___ MHz RAM ___ MB Display adapter $\qquad$
Mouse ___yes ___no Other adapters installed $\qquad$
Hard disk capacity $\qquad$ MB Brand

Instruments used $\qquad$

| National Instruments hardware product model___ Revision ___ |  |
| :--- | :--- |
| Configuration | Version |
| National Instruments software product ___ |  |
| Configuration |  |
| The problem is: |  |

The problem is: $\qquad$
$\qquad$
$\qquad$
$\qquad$

List any error messages: $\qquad$
$\qquad$

The following steps reproduce the problem: $\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$

## LabWindows/CVI Hardware and Software Configuration Form

Record the settings and revisions of your hardware and software on the line to the right of each item. Complete a new copy of this form each time you revise your software or hardware configuration, and use this form as a reference for your current configuration. Completing this form accurately before contacting National Instruments for technical support helps our applications engineers answer your questions more efficiently.

## National Instruments Products

Hardware revision
Interrupt level of hardware $\qquad$
DMA channels of hardware $\qquad$
Base I/O address of hardware $\qquad$
Programming choice $\qquad$
National Instruments software $\qquad$
Other boards in system $\qquad$
Base I/O address of other boards $\qquad$
DMA channels of other boards $\qquad$
Interrupt level of other boards $\qquad$

## Other Products

Computer make and model $\qquad$
Microprocessor $\qquad$
Clock frequency or speed $\qquad$
Type of video board installed $\qquad$
Operating system version $\qquad$
Operating system mode $\qquad$
Programming language $\qquad$
Programming language version $\qquad$
Other boards in system $\qquad$
Base I/O address of other boards $\qquad$
DMA channels of other boards $\qquad$
Interrupt level of other boards $\qquad$

## Documentation Comment Form

National Instruments encourages you to comment on the documentation supplied with our products. This information helps us provide quality products to meet your needs.
Title: LabWindows/CVI Advanced Analysis Library Reference Manual
Edition Date: February 1998
Part Number: 320686D-01

Please comment on the completeness, clarity, and organization of the manual.
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$

If you find errors in the manual, please record the page numbers and describe the errors.
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$
$\qquad$

Thank you for your help.
Name $\qquad$
Title $\qquad$
Company $\qquad$
Address $\qquad$

E-Mail Address $\qquad$
Phone (___ ) $\qquad$ Fax $\qquad$ ) $\qquad$
Mail to: Technical Publications
National Instruments Corporation 6504 Bridge Point Parkway
Austin, Texas 78730-5039

Fax to: Technical Publications National Instruments Corporation 5127945678

## Glossary

| Prefix | Meaning | Value |
| :---: | :---: | :---: |
| $\mathrm{p}-$ | pico- | $10^{-12}$ |
| $\mathrm{n}-$ | nano- | $10^{-9}$ |
| $\mu-$ | micro- | $10^{-6}$ |
| $\mathrm{~m}-$ | milli- | $10^{-3}$ |
| k- | kilo- | $10^{3}$ |
| M- | mega- | $10^{6}$ |

## Numbers/Symbols

1D
2D
A

ANOVA

## C

cm
centimeters.

## D

DFT
Discrete Fourier Transform.
DSP
digital signal processing.

## F

| FFT | Fast Fourier Transform. |
| :--- | :--- |
| FHT | Fast Hartley Transform. <br> FIR |
| function panel | A user interface to the LabWindows/ <br> interactively execute library function <br> a program. |
| function tree | The hierarchical structure in which the <br> instrument driver are grouped. The f <br> library or instrument driver by present <br> to the operation they perform, as opp <br> available functions. |
| H | hertz. <br> Hz |
| IDFT | inverse Discrete Fourier Transform. |
| Infinite impulse response. |  |

## S

sidelobes
The lobes that have lower peak amplitude than the mainlobe and that lie on either side of the mainlobe in the frequency spectrum. The mainlobe is the lobe that has the highest peak amplitude and that is usually centered around zero frequency in the frequency spectrum.
volts.

## Index

## Numbers

1D array functions. See one-dimensional array operation functions.
1D complex operation functions. See one-dimensional complex operation functions.
2D array functions. See two-dimensional array operation functions.

## A

Abs1D function, 2-1
ACDCEstimator function, 2-3
Add1D function, 2-4
Add2D function, 2-5
Advanced Analysis Library functions
class and subclass descriptions, 1-8 to 1-10
error codes
alphabetical list, A-1 to A-4 numeric list, A-4 to A-7
function panels
array operation functions, 1-2 to 1-3
complex operation functions, 1-3 to 1-4
curve fitting functions, 1-7
function tree (table), 1-1 to 1-8
hints for using, 1-10
interpolation functions, 1-7
measurement functions, 1-6
signal generation functions, 1-2
signal processing functions, 1-4 to 1-6
statistics functions, 1-6
vector and matrix algebra functions, 1-7 to 1-8
function reference
Abs1D, 2-1
ACDCEstimator, 2-3
Add1D, 2-4
Add2D, 2-5

AllocIIRFilterPtr, 2-6 to 2-7
AmpPhaseSpectrum, 2-8 to 2-9
ANOVA1Way, 2-10 to 2-15
ANOVA2Way, 2-16 to 2-26
ANOVA3Way, 2-27 to 2-41
ArbitraryWave, 2-42 to 2-43
AutoPowerSpectrum, 2-44 to 2-45
BackSub, 2-46 to 2-47
Bessel_CascadeCoef, 2-48 to 2-49
Bessel_Coef, 2-50 to 2-51
BkmanWin, 2-52
BlkHarrisWin, 2-53
Bw_BPF, 2-54 to 2-55
Bw_BSF, 2-56 to 2-57
Bw_CascadeCoef, 2-58 to 2-59
Bw_Coef, 2-60 to 2-61
Bw_HPF, 2-62 to 2-63
Bw_LPF, 2-64 to 2-65
CascadeToDirectCoef, 2-66 to 2-67
Ch_BPF, 2-68 to 2-69
Ch_BSF, 2-70 to 2-71
Ch_CascadeCoef, 2-72 to 2-73
Ch_Coef, 2-74 to 2-75
Ch_HPF, 2-76 to 2-77
Ch_LPF, 2-78 to 2-79
CheckPosDef, 2-80
Chirp, 2-81
Cholesky, 2-82 to 2-83
Clear1D, 2-84
Clip, 2-85
ConditionNumber, 2-86 to 2-87
Contingency_Table, 2-88 to 2-91
Convolve, 2-92 to 2-93
Copy1D, 2-94
Correlate, 2-95 to 2-96
CosTaperedWin, 2-97
CrossPowerSpectrum, 2-98 to 2-99

CrossSpectrum, 2-100 to 2-101
CxAdd, 2-102
CxAdd1D, 2-103
CxCheckPosDef, 2-104
CxCholesky, 2-105 to 2-106
CxConditionNumber, 2-107 to 2-108
CxDeterminant, 2-109 to 2-110
CxDiv, 2-111
CxDiv1D, 2-112
CxDotProduct, 2-113
CxEigenValueVector, 2-114 to 2-115
CxExp, 2-116
CxGenInvMatrix, 2-117 to 2-118
CxGenLinEqs, 2-119 to 2-120
CxLinEv1D, 2-121 to 2-122
CxLn, 2-123
CxLog, 2-124
CxLU, 2-125 to 2-126
CxMatrixMul, 2-127 to 2-128
CxMatrixNorm, 2-129 to 2-130
CxMatrixRank, 2-131 to 2-132
CxMul, 2-133
CxMul1D, 2-134
CxOuterProduct, 2-135 to 2-136
CxPolyRoots, 2-137 to 2-138
CxPow, 2-139
CxPseudoInverse, 2-140 to 2-141
CxQR, 2-142 to 2-143
CxRecip, 2-144
CxSpecialMatrix, 2-145 to 2-147
CxSqrt, 2-148
CxSub, 2-149
CxSub1D, 2-150
CxSVD, 2-151 to 2-152
CxSVDS, 2-153
CxTrace, 2-154
CxTranspose, 2-155
Decimate, 2-156
Deconvolve, 2-157
Determinant, 2-158

Difference, 2-159 to 2-160
Div1D, 2-161
Div2D, 2-162
DotProduct, 2-163
Elp_BPF, 2-164 to 2-165
Elp_BSF, 2-166 to 2-167
Elp_CascadeCoef, 2-168 to 2-169
Elp_Coef, 2-170 to 2-171
Elp_HPF, 2-172 to 2-173
Elp_LPF, 2-174 to 2-175
Equi_Ripple, 2-176 to 2-179
EquiRpl_BPF, 2-180 to 2-181
EquiRpl_BSF, 2-182 to 2-183
EquiRpl_HPF, 2-184 to 2-185
EquiRpl_LPF, 2-186 to 2-187
ExBkmanWin, 2-188
ExpFit, 2-189 to 2-190
ExpWin, 2-191
F_Dist, 2-192
FFT, 2-193 to 2-194
FHT, 2-195 to 2-196
FIR_Coef, 2-197 to 2-198
FlatTopWin, 2-199
ForceWin, 2-200
ForwSub, 2-201 to 2-202
FreeAnalysisMem, 2-203
FreeIIRFilterPtr, 2-204
GaussNoise, 2-205
GenCosWin, 2-206
GenDeterminant, 2-207 to 2-208
GenEigenValueVector, 2-209 to 2-210
GenInvMatrix, 2-211 to 2-212
GenLinEqs, 2-213 to 2-214
GenLSFit, 2-215 to 2-223
GenLSFitCoef, 2-224 to 2-226
GetAnalysisErrorString, 2-227
HamWin, 2-228
HanWin, 2-229
HarmonicAnalyzer, 2-230 to 2-231
Histogram, 2-232 to 2-233

IIRCascadeFiltering, 2-234 to 2-235
IIRFiltering, 2-236 to 2-237
Impulse, 2-238
ImpulseResponse, 2-239 to 2-240
Integrate, 2-241 to 2-242
InvCh_BPF, 2-243 to 2-244
InvCh_BSF, 2-245 to 2-246
InvCh_CascadeCoef, 2-247 to 2-248
InvCh_Coef, 2-249 to 2-250
InvCh_HFP, 2-251 to 2-252
InvCh_LPF, 2-253 to 2-254
InvF_Dist, 2-255 to 2-256
InvFFT, 2-257 to 2-258
InvFHT, 2-259 to 2-260
InvMatrix, 2-261
InvN_Dist, 2-262
InvT_Dist, 2-263
InvXX_Dist, 2-264
Ksr_BPF, 2-265 to 2-266
Ksr_BSF, 2-267 to 2-268
Ksr_HPF, 2-269 to 2-270
Ksr_LPF, 2-271 to 2-272
KsrWin, 2-273 to 2-274
LinEqs, 2-275
LinEv1D, 2-276
LinEv2D, 2-277
LinFit, 2-278 to 2-279
LU, 2-280 to 2-281
MatrixMul, 2-282 to 2-283
MatrixNorm, 2-284 to 2-285
MatrixRank, 2-286 to 2-287
MaxMin1D, 2-288
MaxMin2D, 2-289 to 2-290
Mean, 2-291
Median, 2-292
Mode, 2-293
Moment, 2-294 to 2-295
Mul1D, 2-296
Mul2D, 2-297
N-Dist, 2-298

Neg1D, 2-299
NetworkFunctions, 2-300 to 2-302
NonLinearFit, 2-303 to 2-304
NonLinearFitWith MaxIters, 2-305
Normal1D, 2-307 to 2-308
Normal2D, 2-309 to 2-310
NumericIntegration, 2-311 to 2-313
OuterProduct, 2-314
PeakDetector, 2-315 to 2-317
PolyEv1D, 2-318 to 2-319
PolyEv2D, 2-320 to 2-321
PolyFit, 2-322 to 2-323
PolyInterp, 2-324 to 2-325
PowerFrequencyEstimate, 2-326 to 2-328
Prod1D, 2-329
PseudoInverse, 2-330 to 2-331
Pulse, 2-332 to 2-333
PulseParam, 2-234 to 2-336
QR, 2-337 to 2-338
QScale1D, 2-339
QScale2D, 2-340
Ramp, 2-341 to 2-342
RatInterp, 2-343 to 2-344
ReFFT, 2-345
ReInvFFT, 2-346
ResetIIRFilter, 2-347 to 2-348
Reverse, 2-349
RMS, 2-350
SawtoothWave, 2-351 to 2-352
Scale1D, 2-353 to 2-354
Scale2D, 2-355 to 2-356
ScaledWindow, 2-357 to 2-358
Set1D, 2-359
Shift, 2-360 to 2-361
Sinc, 2-362
SinePattern, 2-363 to 2-364
SineWave, 2-365 to 2-366
Sort, 2-367
SpecialMatrix, 2-368 to 2-370
Spectrum, 2-371

SpectrumUnitConversion, 2-372 to 2-375
SpInterp, 2-376 to 2-377
Spline, 2-378 to 2-379
SquareWave, 2-380 to 2-381
StdDev, 2-382
Sub1D, 2-383
Sub2D, 2-384
Subset1D, 2-385
Sum1D, 2-386
Sum2D, 2-387
SVD, 2-388 to 2-389
SVDS, 2-390
SymEigenValueVector, 2-391 to 2-392
T_Dist, 2-393
ToPolar, 2-394
ToPolar1D, 2-395
ToRect, 2-396
ToRect1D, 2-397
Trace, 2-398
TransferFunction, 2-399
Transpose, 2-401
Triangle, 2-402
TriangleWave, 2-403 to 2-404
TriWin, 2-405
Uniform, 2-406
UnWrap1D, 2-407
Variance, 2-408
WhiteNoise, 2-409
Wind_BPF, 2-410 to 2-411
Wind_BSF, 2-412 to 2-413
Wind_HPF, 2-414 to 2-415
Wind_LPF, 2-416 to 2-417
XX_Dist, 2-418
AllocIIRFilterPtr function, 2-6 to 2-7
AmpPhaseSpectrum function, 2-8 to 2-9
analysis of variance functions
ANOVA1Way, 2-10 to 2-15
assumptions, 2-12
examples, 2-14 to 2-15
factors and levels, 2-11
formulas, 2-13
general method of using, 2-11
hypothesis, 2-12
parameters, 2-10
purpose, 2-10
return value, 2-11
statistical method, 2-12
ANOVA2Way, 2-16 to 2-26
assumptions, 2-20
examples, 2-24 to 2-26
factors, levels, and cells, 2-18
formulas, 2-21 to 2-23
general method of using, 2-19
hypotheses, 2-20 to 2-21
parameters, 2-16 to 2-17
purpose, 2-16
random and fixed effects, 2-18 to 2-19
return value, 2-17
statistical model, 2-19 to 2-20
ANOVA3Way, 2-27 to 2-41
assumptions, 2-31
examples, 2-37 to 2-41
factors, levels, and cells, 2-29 to 2-30
formulas, 2-33 to 2-37
general method of using, 2-30
hypotheses, 2-32 to 2-33
parameters, 2-27 to 2-29
purpose, 2-27
random and fixed effects, 2-30
return value, 2-29
statistical model, 2-31
definition, 1-9
function tree, 1-6
ArbitraryWave function, 2-42 to 2-43 array analysis, performing in place, 1-10 array operation functions

Abs1D, 2-1
Add1D, 2-4
Add2D, 2-5

Clear1D, 2-84
Copy1D, 2-94
definition, 1-8
Div1D, 2-161
Div2D, 2-162
function tree, 1-2 to 1-3
LinEv1D, 2-276
LinEv2D, 2-277
MaxMin1D, 2-288
MaxMin2D, 2-289 to 2-290
Mul1D, 2-296
Mul2D, 2-297
Neg1D, 2-299
PolyEv1D, 2-318 to 2-319
PolyEv2D, 2-320 to 2-321
Prod1D, 2-329
QScale1D, 2-339
QScale2D, 2-340
Scale1D, 2-353 to 2-354
Scale2D, 2-355 to 2-356
Set1D, 2-359
Sub1D, 2-383
Sub2D, 2-384
Subset1D, 2-385
Sum1D, 2-386
Sum2D, 2-387
UnWrap1D, 2-407
AutoPowerSpectrum function, 2-44 to 2-45

## B

BackSub function, 2-46 to 2-47
basic statistics functions
definition, 1-9
function tree, 1-6
Histogram, 2-232 to 2-233
Mean, 2-291
Median, 2-292
Mode, 2-293
Moment, 2-294 to 2-295
RMS, 2-350

StdDev, 2-382
Variance, 2-408
Bessel_CascadeCoef function, 2-48 to 2-49
Bessel_Coef function, 2-50 to 2-51
BkmanWin function, 2-52
BlkHarrisWin function, 2-53
Bw_BPF function, 2-54 to 2-55
Bw_BSF function, 2-56 to 2-57
Bw_CascadeCoef function, 2-58 to 2-59
Bw_Coef function, 2-60 to 2-61
Bw_HPF function, 2-62 to 2-63
Bw_LPF function, 2-64 to 2-65

## C

CascadeToDirectCoef function, 2-66 to 2-67
Ch_BPF function, 2-68 to 2-69
Ch_BSF function, 2-70 to 2-71
Ch_CascadeCoef function, 2-72 to 2-73
Ch_Coef function, 2-74 to 2-75
Ch_HPF function, 2-76 to 2-77
Ch_LPF function, 2-78 to 2-79
CheckPosDef function, 2-80
Chirp function, 2-81
chi-square tests, 2-89 to 2-90
Cholesky function, 2-82 to 2-83
Clear1D function, 2-84
Clip function, 2-85
complex matrix functions. See vector and matrix algebra functions.
complex operation functions
CxAdd, 2-102
CxAdd1D, 2-103
CxDiv, 2-111
CxDiv1D, 2-112
CxExp, 2-116
CxLinEv1D, 2-121 to 2-122
CxLn, 2-123
CxLog, 2-124
CxMul, 2-133
CxMul1D, 2-134

CxPow, 2-139
CxRecip, 2-144
CxSqrt, 2-148
CxSub, 2-149
CxSub1D, 2-150
definition, 1-9
function tree, 1-3 to 1-4
ToPolar, 2-394
ToPolar1D, 2-395
ToRect, 2-396
ToRect1D, 2-397
ConditionNumber function, 2-86 to 2-87
Contingency_Table function, 2-88 to 2-91
chi-square test of homogeneity, 2-89
chi-square test of independence, 2-90
example, 2-91
formulas, 2-91
hypothesis testing, 2-90
parameters, 2-88
purpose, 2-88
return value, 2-89
Convolve function, 2-92 to 2-93
Copy1D function, 2-94
Correlate function, 2-95 to 2-96
CosTaperedWin function, 2-97
CrossPowerSpectrum function, 2-98 to 2-99
CrossSpectrum function, 2-100 to 2-101
curve fitting, 1-21
curve fitting functions
definition, 1-10
ExpFit, 2-189 to 2-190
function tree, 1-7
GenLSFit, 2-215 to 2-223
GenLSFitCoef, 2-224 to 2-226
LinFit, 2-278 to 2-279
NonLinearFit, 2-303 to 2-304
PolyFit, 2-322 to 2-323
customer communication, xvi, B-1 to B-2
CxAdd function, 2-102
CxAdd1D function, 2-103

CxCheckPosDef function, 2-104
CxCholesky function, 2-105 to 2-106
CxConditionNumber function, 2-107 to 2-108
CxDeterminant function, 2-109 to 2-110
CxDiv function, 2-111
CxDiv1D function, 2-112
CxDotProduct function, 2-113
CxEigenValueVector function, 2-114 to 2-115
CxExp function, 2-116
CxGenInvMatrix function, 2-117 to 2-118
CxGenLinEqs function, 2-119 to 2-120
CxLinEv1D function, 2-121 to 2-122
CxLn function, 2-123
CxLog function, 2-124
CxLU function, 2-125 to 2-126
CxMatrixMul function, 2-127 to 2-128
CxMatrixNorm function, 2-129 to 2-130
CxMatrixRank function, 2-131 to 2-132
CxMul function, 2-133
CxMul1D function, 2-134
CxOuterProduct function, 2-135 to 2-136
CxPolyRoots function, 2-137 to 2-138
CxPow function, 2-139
CxPseudoInverse function, 2-140 to 2-141
CxQR function, 2-142 to 2-143
CxRecip function, 2-144
CxSpecialMatrix function, 2-145 to 2-147
CxSqrt function, 2-148
CxSub function, 2-149
CxSub1D function, 2-150
CxSVD function, 2-151 to 2-152
CxSVDS function, 2-153
CxTrace function, 2-154
CxTranspose function, 2-155

## D

Decimate function, 2-156
Deconvolve function, 2-157
Determinant function, 2-158
Difference function, 2-159 to 2-160
digital filters. See FIR filters; IIR filters. Discrete Fourier Transform (DFT), 1-11
Div1D function, 2-161
Div2D function, 2-162
documentation
conventions used in manual, xiii-xiv
organization of manual, xiii
related documentation, xiv-xvi
DotProduct function, 2-163

## E

electronic support services, B-1 to B-2
Elp_BPF function, 2-164 to 2-165
Elp_BSF function, 2-166 to 2-167
Elp_CascadeCoef function, 2-168 to 2-169
Elp_Coef function, 2-170 to 2-171
Elp_HPF function, 2-172 to 2-173
Elp_LPF function, 2-174 to 2-175
e-mail support, B-2
Equi_Ripple function
description, 2-176 to 2-177
designing FIR filters, 1-16
examples, 2-177 to 2-179
problems with convergence (caution), 1-17
EquiRpl_BPF function, 2-180 to 2-181
EquiRpl_BSF function, 2-182 to 2-183
EquiRpl_HPF function, 2-184 to 2-185
EquiRpl_LPF function, 2-186 to 2-187
error codes
alphabetical list, A-1 to A-4
numeric list, A-4 to A-7
errors
converting error number with GetAnalysisErrorString function, 2-227
reporting analysis errors, 1-11
ExBkmanWin function, 2-188
ExpFit function, 2-189 to 2-190
ExpWin function, 2-191

## F

Fast Fourier Transform (FFT), 1-11 to 1-12.
See also frequency domain functions.
fax and telephone support numbers, B-2
Fax-on-Demand support, B-2
F_Dist function, 2-192
FFT function, 2-193 to 2-194
FHT function, 2-195 to 2-196
finite impulse response functions. See
FIR digital filter functions; FIR filters.
FIR digital filter functions definition, 1-9
FIR_Coef, 2-197 to 2-198
function tree, 1-5
Ksr_BPF, 2-265 to 2-266
Ksr_BSF, 2-267 to 2-268
Ksr_HPF, 2-269 to 2-270
Ksr_LPF, 2-271 to 2-272
Wind_BPF, 2-410 to 2-411
Wind_BSF, 2-412 to 2-413
Wind_HPF, 2-414 to 2-415
Wind_LPF, 2-416 to 2-417
FIR filters
compared with IIR filters, 1-15
definition, 1-15
designing, 1-16 to 1-17
FIR_Coef function, 2-197 to 2-198
FlatTopWin function, 2-199
ForceWin function, 2-200
ForwSub function, 2-201 to 2-202
Fourier Transform integral, 1-11
FreeAnalysisMem function, 2-203
FreeIIRFilterPtr function, 2-204
frequency domain functions
conventions and restrictions related to
Fast Fourier Transform, 1-12
CrossSpectrum, 2-100 to 2-101
definition, 1-9
FFT, 2-193 to 2-194
FHT, 2-195 to 2-196
function tree, 1-4
InvFFT, 2-257 to 2-258
InvFHT, 2-259 to 2-260
notation for describing Fast Fourier
Transform operations, 1-12
ReFFT, 2-345
ReInvFFT, 2-346
Spectrum, 2-371
FTP support, B-1
function panels. See under Advanced Analysis
Library functions.

## G

GaussNoise function, 2-205
GenCosWin function, 2-206
GenDeterminant function, 2-207 to 2-208
GenEigenValueVector function, 2-209 to 2-210
generated code stored in Interactive window, 1-10
GenInvMatrix function, 2-211 to 2-212
GenLinEqs function, 2-213 to 2-214
GenLSFit function, 2-215 to 2-223
example, 2-221 to 2-223
parameters, 2-215 to 2-216
purpose, 2-215
return value, 2-216
using the function, 2-217 to 2-220
GenLSFitCoef function, 2-224 to 2-226
GetAnalysisErrorString function, 2-227

## H

HamWin function, 2-228
HanWin function, 2-229
HarmonicAnalyzer function, 2-230 to 2-231
Histogram function, 2-232 to 2-233

IEW. See Interactive Execution window.
IIR digital filter functions
AllocIIRFilterPtr, 2-6 to 2-7
Bessel_CascadeCoef, 2-48 to 2-49
Bessel_Coef, 2-50 to 2-51
Bw_BPF, 2-54 to 2-55
Bw_BSF, 2-56 to 2-57
Bw_CascadeCoef, 2-58 to 2-59
Bw_Coef, 2-60 to 2-61
Bw_HPF, 2-62 to 2-63
Bw_LPF, 2-64 to 2-65
CascadeToDirectCoef, 2-66 to 2-67
Ch_BPF, 2-68 to 2-69
Ch_BSF, 2-70 to 2-71
Ch_CascadeCoef, 2-72 to 2-73
Ch_Coef, 2-74 to 2-75
Ch_HPF, 2-76 to 2-77
Ch_LPF, 2-78 to 2-79
definition, 1-9
Elp_BPF, 2-164 to 2-165
Elp_BSF, 2-166 to 2-167
Elp_CascadeCoef, 2-168 to 2-169
Elp_Coef, 2-170 to 2-171
Elp_HPF, 2-172 to 2-173
Elp_LPF, 2-174 to 2-175
Equi_Ripple, 2-176 to 2-179
EquiRpl_BPF, 2-180 to 2-181
EquiRpl_BSF, 2-182 to 2-183
EquiRpl_HPF, 2-184 to 2-185
EquiRpl_LPF, 2-186 to 2-187
FreeIIRFilterPtr, 2-204
function tree, 1-4 to 1-5
IIRCascadeFiltering, 2-234 to 2-235
IIRFiltering, 2-236 to 2-237
InvCh_BPF, 2-243 to 2-244
InvCh_BSF, 2-245 to 2-246
InvCh_CascadeCoef, 2-247 to 2-248
InvCh_Coef, 2-249 to 2-250
InvCh_HFP, 2-251 to 2-252

InvCh_LPF, 2-253 to 2-254
ResetIIRFilter, 2-347 to 2-348
IIR filters, 1-17 to 1-19
cascaded filter stages, 1-18
compared with FIR filters, 1-15
direct form, 1-17
fourth order, 1-18 to 1-19
mathematical form, 1-17
second order, 1-18
types, 1-19
IIRCascadeFiltering function, 2-234 to 2-235
IIRFiltering function, 2-236 to 2-237
Impulse function, 2-238
ImpulseResponse function, 2-239 to 2-240
infinite impulse response functions. See
IIR digital filter functions; IIR filters.
Integrate function, 2-241 to 2-242
Interactive Execution window, 1-10
interpolation functions
definition, 1-10
function tree, 1-7
PolyInterp, 2-324 to 2-325
RatInterp, 2-343 to 2-344
SpInterp, 2-376 to 2-377
Spline, 2-378 to 2-379
InvCh_BPF function, 2-243 to 2-244
InvCh_BSF function, 2-245 to 2-246
InvCh_CascadeCoef function, 2-247 to 2-248
InvCh_Coef function, 2-249 to 2-250
InvCh_HFP function, 2-251 to 2-252
InvCh_LPF function, 2-253 to 2-254
InvF_Dist function, 2-255 to 2-256
InvFFT function, 2-257 to 2-258
InvFHT function, 2-259 to 2-260
InvMatrix function, 2-261
InvN_Dist function, 2-262
InvT_Dist function, 2-263
InvXX_Dist function, 2-264

## K

Ksr_BPF function, 2-265 to 2-266
Ksr_BSF function, 2-267 to 2-268
Ksr_HPF function, 2-269 to 2-270
Ksr_LPF function, 2-271 to 2-272
KsrWin function, 2-273 to 2-274

## L

LinEqs function, 2-275
LinEv1D function, 2-276
LinEv2D function, 2-277
LinFit function, 2-278 to 2-279
LU function, 2-280 to 2-281

## M

manual. See documentation.
matrix algebra functions. See vector and matrix algebra functions.
MatrixMul function, 2-282 to 2-283
MatrixNorm function, 2-284 to 2-285
MatrixRank function, 2-286 to 2-287
MaxMin1D function, 2-288
MaxMin2D function, 2-289 to 2-290
Mean function, 2-291
measurement functions
ACDCEstimator, 2-3
AmpPhaseSpectrum, 2-8 to 2-9
AutoPowerSpectrum, 2-44 to 2-45
characteristics, 1-20
CrossPowerSpectrum, 2-98 to 2-99
definition, 1-9
function tree, 1-6
HarmonicAnalyzer, 2-230 to 2-231
ImpulseResponse, 2-239 to 2-240
NetworkFunctions, 2-300 to 2-302
PowerFrequencyEstimate, 2-326 to 2-328
purpose and use, 1-19 to 1-20
ScaledWindow, 2-357 to 2-358

SpectrumUnitConversion, 2-372 to 2-375
TransferFunction, 2-399
Median function, 2-292
Mode function, 2-293
Moment function, 2-294 to 2-295
Mul1D function, 2-296
Mul2D function, 2-297

## N

N-Dist function, 2-298
Neg1D function, 2-299
NetworkFunctions function, 2-300 to 2-302
NonLinearFit function, 2-303 to 2-304
NonLinearFitWith MaxIters function, 2-305 nonparametric statistics function

Contingency_Table, 2-88 to 2-91
definition, 1-9
function tree, 1-6
Normal1D function, 2-307 to 2-308
Normal2D function, 2-309 to 2-310
NumericIntegration function, 2-311 to 2-313

## 0

one-dimensional array operation functions
Abs1D, 2-1
Add1D, 2-4
definition, 1-8
Div1D, 2-161
function tree, 1-2 to 1-3
LinEv1D, 2-276
MaxMin1D, 2-288
Mul1D, 2-296
Neg1D, 2-299
PolyEv1D, 2-318 to 2-319
Prod1D, 2-329
QScale1D, 2-339
Scale1D, 2-353 to 2-354
Sub1D, 2-383
Subset1D, 2-385

Sum1D, 2-386
Sum2D, 2-387
one-dimensional complex operation functions
CxAdd1D, 2-103
CxDiv1D, 2-112
CxLinEv1D, 2-121 to 2-122
CxMul1D, 2-134
CxSub1D, 2-150
definition, 1-9
function tree, 1-4
ToPolar1D, 2-395
ToRect1D, 2-397
OuterProduct function, 2-314

## P

PeakDetector function, 2-315 to 2-317
performance considerations, analysis functions, 1-10
PolyEv1D function, 2-318 to 2-319
PolyEv2D function, 2-320 to 2-321
PolyFit function, 2-322 to 2-323
PolyInterp function, 2-324 to 2-325
PowerFrequencyEstimate function, 2-326 to 2-328
probability distribution functions
definition, 1-9
F_Dist, 2-192
function tree, 1-6
InvF_Dist, 2-255 to 2-256
InvN_Dist, 2-262
InvT_Dist, 2-263
InvXX_Dist, 2-264
N-Dist, 2-298
T_Dist, 2-393
XX_Dist, 2-418
Prod1D function, 2-329
PseudoInverse function, 2-330 to 2-331
Pulse function, 2-332 to 2-333
PulseParam function, 2-234 to 2-336

## Q

QR function, 2-337 to 2-338
QScale1D function, 2-339
QScale2D function, 2-340

## R

Radix-4 and Radix-8 algorithms, 1-12
Ramp function, 2-341 to 2-342
RatInterp function, 2-343 to 2-344
real matrix functions. See vector and matrix algebra functions.
ReFFT function, 2-345
ReInvFFT function, 2-346
ResetIIRFilter function, 2-347 to 2-348
Reverse function, 2-349
RMS function, 2-350

## S

SawtoothWave function, 2-351 to 2-352
Scale1D function, 2-353 to 2-354
Scale2D function, 2-355 to 2-356
ScaledWindow function, 2-357 to 2-358
Set1D function, 2-359
Shift function, 2-360 to 2-361
signal generation functions
ArbitraryWave, 2-42 to 2-43
Chirp, 2-81
definition, 1-9
function tree, 1-2
GaussNoise, 2-205
Impulse, 2-238
Pulse, 2-332 to 2-333
Ramp, 2-341 to 2-342
SawtoothWave, 2-351 to 2-352
Sinc, 2-362
SinePattern, 2-363 to 2-364
SineWave, 2-365 to 2-366
SquareWave, 2-380 to 2-381

Triangle, 2-402
TriangleWave, 2-403 to 2-404
Uniform, 2-406
WhiteNoise, 2-409
signal processing functions. See FIR digital filter functions; frequency domain functions; IIR digital filter functions;
time domain functions; windows functions.
Sinc function, 2-362
SinePattern function, 2-363 to 2-364
SineWave function, 2-365 to 2-366
Sort function, 2-367
source file code stored in Interactive window, 1-10
SpecialMatrix function, 2-368 to 2-370
Spectrum function, 2-371
SpectrumUnitConversion function, 2-372 to 2-375
SpInterp function, 2-376 to 2-377
Spline function, 2-378 to 2-379
Split-Radix algorithm, 1-12
SquareWave function, 2-380 to 2-381
statistics functions
ANOVA1Way, 2-10 to 2-15
ANOVA2Way, 2-16 to 2-26
ANOVA3Way, 2-27 to 2-41
Contingency_Table, 2-88 to 2-91
definition, 1-9
F_Dist, 2-192
function tree, 1-6
GenLSFit, 2-215 to 2-223
Histogram, 2-232 to 2-233
InvF_Dist, 2-255 to 2-256
InvN_Dist, 2-262
InvT_Dist, 2-263
InvXX_Dist, 2-264
Mean, 2-291
Median, 2-292
Mode, 2-293
Moment, 2-294 to 2-295
N-Dist, 2-298

RMS, 2-350
Sort, 2-367
StdDev, 2-382
T_Dist, 2-393
Variance, 2-408
XX_Dist, 2-418
StdDev function, 2-382
Sub1D function, 2-383
Sub2D function, 2-384
Subset1D function, 2-385
Sum1D function, 2-386
Sum2D function, 2-387
SVD function, 2-388 to 2-389
SVDS function, 2-390
SymEigenValueVector function, 2-391 to 2-392

## T

T_Dist function, 2-393
technical support, B-1 to B-2
telephone and fax support numbers, B-2 time domain functions

Clip, 2-85
Convolve, 2-92 to 2-93
Correlate, 2-95 to 2-96
Decimate, 2-156
Deconvolve, 2-157
definition, 1-9
Difference, 2-159 to 2-160
function tree, 1-4
Integrate, 2-241 to 2-242
PulseParam, 2-234 to 2-336
Reverse, 2-349
Shift, 2-360 to 2-361
ToPolar function, 2-394
ToPolar1D function, 2-395
ToRect function, 2-396
ToRect1D function, 2-397
Trace function, 2-398
TransferFunction function, 2-399

Transpose function, 2-401
Triangle function, 2-402
TriangleWave function, 2-403 to 2-404
TriWin function, 2-405
two-dimensional array operation functions
Add2D, 2-5
definition, 1-8
Div2D, 2-162
function tree, 1-3
LinEv2D, 2-277
MaxMin2D, 2-289 to 2-290
Mul2D, 2-297
PolyEv2D, 2-320 to 2-321
QScale2D, 2-340
Scale2D, 2-355 to 2-356
Sub2D, 2-384

## U

Uniform function, 2-406
UnWrap1D function, 2-407

## V

Variance function, 2-408
vector and matrix algebra functions
BackSub, 2-46 to 2-47
CheckPosDef, 2-80
Cholesky, 2-82 to 2-83
ConditionNumber, 2-86 to 2-87
CxCheckPosDef, 2-104
CxCholesky, 2-105 to 2-106
CxConditionNumber, 2-107 to 2-108
CxDeterminant, 2-109 to 2-110
CxDotProduct, 2-113
CxEigenValueVector, 2-114 to 2-115
CxGenInvMatrix, 2-117 to 2-118
CxGenLinEqs, 2-119 to 2-120
CxLU, 2-125 to 2-126
CxMatrixMul, 2-127 to 2-128
CxMatrixNorm, 2-129 to 2-130

CxMatrixRank, 2-131 to 2-132
CxOuterProduct, 2-135 to 2-136
CxPolyRoots, 2-137 to 2-138
CxPseudoInverse, 2-140 to 2-141
CxQR, 2-142 to 2-143
CxSpecialMatrix, 2-145 to 2-147
CxSVD, 2-151 to 2-152
CxSVDS, 2-153
CxTrace, 2-154
CxTranspose, 2-155
definition, 1-10
Determinant, 2-158
DotProduct, 2-163
ForwSub, 2-201 to 2-202
function tree, 1-7 to 1-8
GenDeterminant, 2-207 to 2-208
GenEigenValueVector, 2-209 to 2-210
GenInvMatrix, 2-211 to 2-212
GenLinEqs, 2-213 to 2-214
InvMatrix, 2-261
LinEqs, 2-275
LU, 2-280 to 2-281
MatrixMul, 2-282 to 2-283
MatrixNorm, 2-284 to 2-285
MatrixRank, 2-286 to 2-287
NonLinearFitWith MaxIters, 2-305
Normal1D, 2-307 to 2-308
Normal2D, 2-309 to 2-310
OuterProduct, 2-314
PseudoInverse, 2-330 to 2-331
purpose and use, 1-21
QR, 2-337 to 2-338
SpecialMatrix, 2-368 to 2-370
SVD, 2-388 to 2-389
SVDS, 2-390
SymEigenValueVector, 2-391 to 2-392
Trace, 2-398
Transpose, 2-401

## W

WhiteNoise function, 2-409
Wind_BPF function, 2-410 to 2-411
Wind_BSF function, 2-412 to 2-413
Wind_HPF function, 2-414 to 2-415
Wind_LPF function, 2-416 to 2-417
windowing, $1-13$ to $1-15$
windows functions
BkmanWin, 2-52
BlkHarrisWin, 2-53
CosTaperedWin, 2-97
definition, 1-9
ExBkmanWin, 2-188
ExpWin, 2-191
FlatTopWin, 2-199
ForceWin, 2-200
function tree, 1-5 to 1-6
GenCosWin, 2-206
HamWin, 2-228
HanWin, 2-229
KsrWin, 2-273 to 2-274
TriWin, 2-405
windType parameter, 1-16

## X

XX_Dist function, 2-418


[^0]:    $\sqrt{3}$
    Note
    The y values must be all positive or all negative to perform an exponential fit.

