# NAG Fortran Library Routine Document

## F04FEF

Note: before using this routine, please read the Users' Note for your implementation to check the interpretation of *bold italicised* terms and other implementation-dependent details.

## **1** Purpose

F04FEF solves the Yule-Walker equations for a real symmetric positive-definite Toeplitz system.

## 2 Specification

```
SUBROUTINE F04FEF(N, T, X, WANTP, P, WANTV, V, VLAST, WORK, IFAIL)INTEGERN, IFAILrealT(0:N), X(*), P(*), V(*), VLAST, WORK(*)LOGICALWANTP, WANTV
```

## **3** Description

This routine solves the equations

$$Tx = -t,$$

where T is the n by n symmetric positive-definite Toeplitz matrix

$$T = \begin{pmatrix} \tau_0 & \tau_1 & \tau_2 & \dots & \tau_{n-1} \\ \tau_1 & \tau_0 & \tau_1 & \dots & \tau_{n-2} \\ \tau_2 & \tau_1 & \tau_0 & \dots & \tau_{n-3} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \tau_{n-1} & \tau_{n-2} & \tau_{n-3} & \dots & \tau_0 \end{pmatrix}$$

and t is the vector

$$t^T = (\tau_1 \, \tau_2 \dots \tau_n).$$

The routine uses the method of Durbin (see Durbin (1960) and Golub and van Loan (1996)). Optionally the mean square prediction errors and/or the partial correlation coefficients for each step can be returned.

## 4 References

Bunch J R (1985) Stability of methods for solving Toeplitz systems of equations SIAM J. Sci. Statist. Comput. 6 349–364

Bunch J R (1987) The weak and strong stability of algorithms in numerical linear algebra *Linear Algebra Appl.* **88/89** 49–66

Cybenko G (1980) The numerical stability of the Levinson–Durbin algorithm for Toeplitz systems of equations SIAM J. Sci. Statist. Comput. 1 303–319

Durbin J (1960) The fitting of time series models Rev. Inst. Internat. Stat. 28 233

Golub G H and van Loan C F (1996) Matrix Computations (3rd Edition) Johns Hopkins University Press, Baltimore

## 5 Parameters

1: N – INTEGER

On entry: the order of the Toeplitz matrix T.

Constraint:  $N \ge 0$ . When N=0, then an immediate return is effected.

Input

#### 2: T(0:N) - real array

On entry: T(0) must contain the value  $\tau_0$  of the diagonal elements of T, and the remaining N elements of T must contain the elements of the vector t.

Constraint: T(0) > 0.0. Note that if this is not true, then the Toeplitz matrix cannot be positive-definite.

#### 3: X(\*) - real array

Note: the dimension of the array X must be at least max(1, N).

On exit: the solution vector x.

## 4: WANTP – LOGICAL

*On entry*: WANTP must be set to .TRUE. if the partial (auto)correlation coefficients are required, and must be set to .FALSE. otherwise.

#### 5: P(\*) - real array

Note: the dimension of the array P must be at least max(1, N), if WANTP = .TRUE., otherwise the dimension must be at least 1.

On exit: with WANTP as .TRUE., the *i*th element of P contains the partial (auto)correlation coefficient, or reflection coefficient,  $p_i$  for the *i*th step. (See Section 8 and Chapter G13.) If WANTP is .FALSE., then P is not referenced. Note that in any case,  $x_n = p_n$ .

#### 6: WANTV – LOGICAL

*On entry*: WANTV must be set to .TRUE. if the mean square prediction errors are required, and must be set to .FALSE. otherwise.

#### 7: V(\*) - real array

Note: the dimension of the array V must be at least max(1, N), if WANTV = .TRUE, otherwise the dimension must be at least 1.

On exit: with WANTV as .TRUE., the *i*th element of V contains the mean square prediction error, or predictor error variance ratio,  $v_i$ , for the *i*th step. (See Section 8 and Chapter G13.) If WANTV is .FALSE., then V is not referenced.

#### 8: VLAST – *real*

On exit: the value of  $v_n$ , the mean square prediction error for the final step.

#### 9: WORK(\*) – *real* array

Note: the dimension of the array WORK must be at least max(1, N - 1).

#### 10: IFAIL – INTEGER

On entry: IFAIL must be set to 0, -1 or 1. Users who are unfamiliar with this parameter should refer to Chapter P01 for details.

On exit: IFAIL = 0 unless the routine detects an error (see Section 6).

For environments where it might be inappropriate to halt program execution when an error is detected, the value -1 or 1 is recommended. If the output of error messages is undesirable, then the value 1 is recommended. Otherwise, because for this routine the values of the output parameters may be useful even if IFAIL  $\neq 0$  on exit, the recommended value is -1. When the value -1 or 1 is used it is essential to test the value of IFAIL on exit.

Workspace

Input/Output

Output

# Input

Output

Input

Output

Input

Output

#### 6 Error Indicators and Warnings

If on entry IFAIL = 0 or -1, explanatory error messages are output on the current error message unit (as defined by X04AAF).

Errors or warnings detected by the routine:

$$IFAIL = -1$$

 $\begin{array}{ll} \text{On entry,} & N < 0, \\ \text{or} & T(0) \leq 0.0. \end{array}$ 

IFAIL > 0

The principal minor of order (IFAIL + 1) of the Toeplitz matrix is not positive-definite to working accuracy. If, on exit, *x*IFAIL is close to unity, then the principal minor was close to being singular, and the sequence  $\tau_0, \tau_1, \ldots, \tau$ IFAIL may be a valid sequence nevertheless. The first IFAIL elements of X return the solution of the equations

T

IFAIL  $x = -(\tau_1, \tau_2, \ldots, \tau$  IFAIL)<sup>T</sup>, where TIFAIL is the IFAILth principal minor of T. Similarly, if WANTP and/or WANTV are true, then P and/or V return the first IFAIL elements of P and V respectively and VLAST returns vIFAIL. In particular if IFAIL = N, then the solution of the equations Tx = -t is returned in X, but  $\tau$ N is such that TN + 1 would not be positive-definite to working accuracy.

## 7 Accuracy

The computed solution of the equations certainly satisfies

r = Tx + t,

where  $||r||_1$  is approximately bounded by

$$||r||_1 \le c\epsilon \left(\prod_{i=1}^n (1+|p_i|)-1\right),$$

c being a modest function of n and  $\epsilon$  being the *machine precision*. This bound is almost certainly pessimistic, but it has not yet been established whether or not the method of Durbin is backward stable. If  $|p_n|$  is close to one, then the Toeplitz matrix is probably ill-conditioned and hence only just positive-definite. For further information on stability issues see Bunch (1985), Bunch (1987), Cybenko (1980) and Golub and van Loan (1996). The following bounds on  $||T^{-1}||_1$  hold:

$$\max\left(\frac{1}{v_{n-1}}, \frac{1}{\prod_{i=1}^{n-1}(1-p_i)}\right) \le \|T^{-1}\|_1 \le \prod_{i=1}^{n-1}\left(\frac{1+|p_i|}{1-|p_i|}\right).$$

Note:  $v_n < v_{n-1}$ . The norm of T<sup>-1</sup> may also be estimated using routine F04YCF.

#### 8 **Further Comments**

The number of floating-point operations used by this routine is approximately  $2n^2$ , independent of the values of WANTP and WANTV.

The mean square prediction error,  $v_i$ , is defined as

$$v_i = (\tau_0 + (\tau_1 \tau_2 \dots \tau_{i-1}) y_{i-1}) / \tau_0,$$

where  $y_i$  is the solution of the equations

$$T_i y_i = -(\tau_1 \tau_2 \dots \tau_i)^T$$

and the partial correlation coefficient,  $p_i$ , is defined as the *i*th element of  $y_i$ . Note that  $v_i = (1 - p_i^2)v_{i-1}$ .

#### 9 Example

To find the solution of the Yule–Walker equations Tx = -t, where

$$T = \begin{pmatrix} 4 & 3 & 2 & 1 \\ 3 & 4 & 3 & 2 \\ 2 & 3 & 4 & 3 \\ 1 & 2 & 3 & 4 \end{pmatrix} \text{ and } t = \begin{pmatrix} 3 \\ 2 \\ 1 \\ 0 \end{pmatrix}.$$

#### 9.1 Program Text

**Note:** the listing of the example program presented below uses *bold italicised* terms to denote precision-dependent details. Please read the Users' Note for your implementation to check the interpretation of these terms. As explained in the Essential Introduction to this manual, the results produced may not be identical for all implementations.

```
FO4FEF Example Program Text
*
*
      Mark 15 Release. NAG Copyright 1991.
      .. Parameters ..
*
      INTEGER
                       NIN, NOUT
      PARAMETER
                       (NIN=5,NOUT=6)
      INTEGER
                       NMAX
      PARAMETER
                       (NMAX=100)
      .. Local Scalars ..
      real
                       VLAST
                       I, IFAIL, N
      INTEGER
      LOGICAL
                       WANTP, WANTV
      .. Local Arrays ..
      real
                       P(NMAX), T(0:NMAX), V(NMAX), WORK(NMAX-1),
     +
                       X(NMAX)
      .. External Subroutines ..
      EXTERNAL
                      F04FEF
      .. Executable Statements ..
      WRITE (NOUT, *) 'F04FEF Example Program Results'
      Skip heading in data Ûle
      READ (NIN,*)
      READ (NIN,*) N
      WRITE (NOUT, *)
      IF ((N.LT.O) .OR. (N.GT.NMAX)) THEN
         WRITE (NOUT,99999) 'N is out of range. N = ', N
      ELSE
         READ (NIN, *) (T(I), I=0, N)
         WANTP = .TRUE.
         WANTV = .TRUE.
*
         IFAIL = -1
*
         CALL F04FEF(N,T,X,WANTP,P,WANTV,V,VLAST,WORK,IFAIL)
*
         IF (IFAIL.EQ.0) THEN
            WRITE (NOUT, *)
            WRITE (NOUT, *) 'Solution vector'
            WRITE (NOUT, 99998) (X(I), I=1, N)
            IF (WANTP) THEN
               WRITE (NOUT, *)
               WRITE (NOUT,*) 'Refection coefûcients'
               WRITE (NOUT, 99998) (P(I), I=1, N)
            END IF
            IF (WANTV) THEN
               WRITE (NOUT, *)
               WRITE (NOUT,*) 'Mean square prediction errors'
               WRITE (NOUT, 99998) (V(I), I=1, N)
            END IF
         ELSE IF (IFAIL.GT.O) THEN
            WRITE (NOUT, *)
            WRITE (NOUT, 99999) 'Solution for system of order', IFAIL
            WRITE (NOUT, 99998) (X(I), I=1, IFAIL)
            IF (WANTP) THEN
               WRITE (NOUT, *)
               WRITE (NOUT,*) 'Refection coefûcients'
```

```
WRITE (NOUT,99998) (P(I),I=1,IFAIL)
END IF
IF (WANTV) THEN
WRITE (NOUT,*)
WRITE (NOUT,*) 'Mean square prediction errors'
WRITE (NOUT,99998) (V(I),I=1,IFAIL)
END IF
END IF
END IF
STOP
*
99999 FORMAT (1X,A,I5)
99998 FORMAT (1X,5F9.4)
END
```

## 9.2 Program Data

FO4FEF Example Program Data

4 :Value of N 4.0 3.0 2.0 1.0 0.0 :End of vector T

## 9.3 Program Results

F04FEF Example Program Results

Solution vector -0.8000 0.0000 -0.0000 0.2000 Refection coefÛcients -0.7500 0.1429 0.1667 0.2000 Mean square prediction errors 0.4375 0.4286 0.4167 0.4000