NAG C Library Function Document

nag_rngs_arma_time_series (g05pac)

1 Purpose

nag_rngs_arma_time_series (g05pac) generates a realisation of a univariate time series from an autoregressive moving average (ARMA) model. The realisation may be continued or a new realisation generated at subsequent calls to nag_rngs_arma_time_series (g05pac).

2 Specification

3 Description

Let the vector x_t , denote a time series which is assumed to follow an autoregressive moving average (ARMA) model of the form:

$$x_{t} - \mu = \phi_{1}(x_{t-1} - \mu) + \phi_{2}(x_{t-2} - \mu) + \dots + \phi_{p}(x_{t-p} - \mu) + \epsilon_{t} - \theta_{1}\epsilon_{t-1} - \theta_{2}\epsilon_{t-2} - \dots - \theta_{q}\epsilon_{t-q}$$
(1)

where ϵ_i , is a residual series of independent random perturbations assumed to be Normally distributed with zero mean and variance σ^2 . The parameters $\{\phi_i\}$, for i = 1, 2, ..., p, are called the autoregressive (AR) parameters, and $\{\theta_j\}$, for j = 1, 2, ..., q, the moving average (MA) parameters. The parameters in the model are thus the p ϕ -values, the q θ -values, the mean μ and the residual variance σ^2 .

nag_rngs_arma_time_series (g05pac) sets up a reference vector containing initial values corresponding to a stationary position using the method described in Tunnicliffe–Wilson (1979). The function can then return a realisation of x_1, x_2, \ldots, x_n . On a successful exit, the recent history is updated and saved in the reference vector **r** so that nag_rngs_arma_time_series (g05pac) may be called again to generate a realisation of x_{n+1}, x_{n+2}, \ldots , etc. See the description of the parameter **mode** in Section 5 for details.

One of the initialisation functions nag_rngs_init_repeatable (g05kbc) (for a repeatable sequence if computed sequentially) or nag_rngs_init_nonrepeatable (g05kcc) (for a non-repeatable sequence) must be called prior to the first call to nag_rngs_arma_time_series (g05pac).

4 References

Knuth D E (1981) The Art of Computer Programming (Volume 2) (2nd Edition) Addison-Wesley

Tunnicliffe–Wilson G (1979) Some efficient computational procedures for high order ARMA models J. Statist. Comput. Simulation 8 301–309

5 Parameters

```
1: mode – Integer
```

On entry: a code for selecting the operation to be performed by the function:

mode = 0

Set up reference vector only.

mode = 1

Generate terms in the time series using reference vector set up in a prior call to nag_rngs_arma_time_series (g05pac).

Input

	mode = 2	
	Set up reference vector and generate terms in the time series.	
	<i>Constraint</i> : $0 \leq \text{mode} \leq 2$.	
2:	xmean – double	Input
	On entry: the mean of the time series.	
3:	p – Integer	Input
	On entry: the number of autoregressive coefficients supplied, p.	
	Constraint: $\mathbf{p} \ge 0$.	
4:	$\mathbf{phi}[dim] - \mathbf{const}$ double	Input
	Note: the dimension, dim , of the array phi must be at least max $(1, \mathbf{p})$.	
	<i>On entry</i> : the autoregressive coefficients of the model, $\phi_1, \phi_2, \ldots, \phi_p$.	
5:	\mathbf{q} – Integer	Input
	On entry: the number of moving average coefficients supplied, q.	
	Constraint: $\mathbf{q} \geq 0$.	
6:	theta $[dim]$ – const double	Input
	Note: the dimension, dim , of the array theta must be at least max $(1, \mathbf{q})$.	Ĩ
	On entry: the moving average coefficients of the model, $\theta_1, \theta_2, \ldots, \theta_q$.	
7:	avar – double	Input
	On entry: the variance of the normal perturbations, σ^2 .	
	Constraint: $\mathbf{avar} \ge 0.0$.	
8:	var – double *	Output
	<i>On exit</i> : the proportion of the variance of a term in the series that is due to the mo (error) terms in the model. The smaller this is, the nearer is the model to non-station	ving-average
9:	n – Integer	Input
	On entry: the number of observations to be generated, n.	
	Constraint: $\mathbf{n} \ge 0$.	
10:	$\mathbf{x}[dim] - double$	Output
	Note: the dimension, dim , of the array x must be at least max $(1, \mathbf{n})$.	
	On exit: contains the next n observations from the time series.	
11:	igen – Integer	Input
	<i>On entry</i> : must contain the identification number for the generator to be used to return random number and should remain unchanged following initialisation by a prior call t functions nag_rngs_init_repeatable (g05kbc) or nag_rngs_init_nonrepeatable (g05kcc).	-
12:	iseed[4] – Integer	Input/Output
	On entry: contains values which define the current state of the selected generator.	
	On exit: contains updated values defining the new state of the selected generator.	

On exit: contains updated values defining the new state of the selected generator.

13: $\mathbf{r}[dim] - double$

Note: the dimension, dim, of the array **r** must be at least $\mathbf{p} + \mathbf{q} + 5 + \max(\mathbf{p}, \mathbf{q} + 1)$. On exit: the reference vector.

 14:
 fail – NagError *
 Input/Output

The NAG error parameter (see the Essential Introduction).

6 Error Indicators and Warnings

NE_INT

On entry, $\mathbf{p} = \langle value \rangle$. Constraint: $\mathbf{p} \ge 0$.

On entry, $\mathbf{q} = \langle value \rangle$. Constraint: $\mathbf{q} \ge 0$.

On entry, $mode = \langle value \rangle$. Constraint: $0 \leq mode > \leq$.

On entry, $\mathbf{n} = \langle value \rangle$. Constraint: $\mathbf{n} \ge 0$.

NE_REAL

On entry, $\mathbf{avar} = \langle value \rangle$. Constraint: $\mathbf{avar} \ge 0.0$.

NE_STATIONARY_AR

phi does not define a stationary autoregresive process.

NE_BAD_PARAM

On entry, parameter $\langle value \rangle$ had an illegal value.

NE_INTERNAL_ERROR

An internal error has occurred in this function. Check the function call and any array sizes. If the call is correct then please consult NAG for assistance.

7 Accuracy

The errors in the initialisation process should be very much smaller than the error term; see Tunnicliffe–Wilson (1979).

8 Further Comments

The time taken by nag_rngs_arma_time_series (g05pac) is essentially of order $(\mathbf{p})^2$.

Note: nag_rngs_init_repeatable (g05kbc) and nag_rngs_init_nonrepeatable (g05kcc) must be used with care if this function is used as well. The reference vector, as mentioned before, contains a copy of the recent history of the series. This will not be altered properly by calls to any of the above functions. A call to nag_rngs_init_repeatable (g05kbc) or nag_rngs_init_nonrepeatable (g05kcc) should be followed by a call to nag_rngs_arma_time_series (g05pac) with **mode** = 0 to re-initialise the time series reference vector in use. To maintain repeatability with nag_rngs_init_repeatable (g05kbc), the calls to nag_rngs_arma_time_series (g05pac) should be performed in the same order and at the same point or points in the simulation every time nag_rngs_init_repeatable (g05kbc) is used. When the generator state is saved and restored using the parameters **igen** and **iseed**, the time series reference vector must be saved and restored as well.

Input/Output

The ARMA model for a time series can also be written as:

$$(x_n - E) = A_1(x_{n-1} - E) + \dots + A_{NA}(x_{n-NA} - E) + B_1a_n + \dots + B_{NB}a_{n-NB+1}$$

where

 x_n is the observed value of the time series at time n,

NA is the number of autoregressive parameters, A_i ,

NB is the number of moving average parameters, B_i ,

E is the mean of the time series,

and

 a_t is a series of independent random Standard Normal perturbations.

This is related to the form given in Section 3 by:

$$B_1^2 = \sigma^2,$$

$$B_{i+1} = -\theta_i \sigma = -\theta_i B_1, \quad i = 1, 2, \dots, q,$$

$$NB = q + 1,$$

$$E = c,$$

$$A_i = \phi_i, \quad i = 1, 2, \dots, p,$$

$$NA = p.$$

9 Example

This example program calls nag_rngs_arma_time_series (g05pac) to set up the reference vector for an autoregressive model after initialisation by nag_rngs_init_repeatable (g05kbc). The model is given by

$$x_t = 0.4x_{t-1} + 0.2x_{t-2} + \epsilon_t$$

where ϵ_t is a series of independent random Normal perturbations with variance 1.0. nag_rngs_arma_time_series (g05pac) is then called generate a sample of ten observations, which are printed.

9.1 Program Text

```
/* nag_rngs_arma_time_series(g05pac) Example Program.
 * Copyright 2001 Numerical Algorithms Group.
 * Mark 7, 2001.
 */
#include <stdio.h>
#include <nag.h>
#include <nag_stdlib.h>
#include <nagg05.h>
int main(void)
{
  /* Scalars */
  double avar, var, xmean;
 Integer i, igen, ip, iq, n, nr;
Integer exit_status=0;
  NagError fail;
  /* Arrays */
  double *phi=0, *r=0, *theta=0, *x=0;
  Integer iseed[4];
  INIT_FAIL(fail);
  Vprintf("g05pac Example Program Results\n\n");
```

```
ip=2;
 iq=0;
 n=10;
 nr=ip+iq+5+ip;
 /* allocate memory */
 if ( !(phi = NAG_ALLOC(ip, double)) ||
      !(r = NAG_ALLOC(nr, double)) ||
      !(theta = NAG_ALLOC(1, double)) ||
      !(x = NAG_ALLOC(n, double)) )
   {
     Vprintf("Allocation failure\n");
     exit_status = -1;
     goto END;
   }
 /* Set the ARMA model parameters */
 xmean = 0.0;
 phi[0] = 0.4;
 phi[1] = 0.2;
 avar = 1.0;
 /* Initialise the seed to a repeatable sequence */
 iseed[0] = 1762543;
 iseed[1] = 9324783;
 iseed[2] = 4234401;
 iseed[3] = 742355;
 /* igen identifies the stream. */
 igen = 1;
 g05kbc(&igen, iseed);
 /* Set up the reference vector */
 gO5pac(O, xmean, ip, phi, iq, theta, avar, &var, n, x, igen,
 iseed, r, &fail);
if (fail.code != NE_NOERROR)
   {
     Vprintf("Error from g05pac.\n%s\n", fail.message);
     exit_status = 1;
     goto END;
   }
 /* Generate a sample of 10 observations */
 gO5pac(1, xmean, ip, phi, iq, theta, avar, &var, n, x, igen,
 iseed, r, &fail);
if (fail.code != NE_NOERROR)
     Vprintf("Error from g05pac.\n%s\n", fail.message);
     exit_status = 1;
     goto END;
   }
 for (i = 0; i < n; ++i)
   {
     Vprintf("%12.4f\n", x[i]);
  }
END:
 if (phi) NAG_FREE(phi);
 if (r) NAG_FREE(r);
if (theta) NAG_FREE(theta);
 if (x) NAG_FREE(x);
 return exit_status;
```

9.2 Program Data

None.

}

9.3 Program Results

g05pac Example Program Results

-1.0654 -0.2828 -2.0924 -2.3304 -2.5998 -1.7143 -2.4882 -1.3882 -2.2722 -1.8806