NAG Library Routine Document

G05PJF

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

G05PJF generates a realization of a multivariate time series from a vector autoregressive moving average (VARMA) model. The realization may be continued or a new realization generated at subsequent calls to G05PJF.

2   Specification

```
SUBROUTINE G05PJF (MODE, N, K, XMEAN, IP, PHI, IQ, THETA, VAR, LDVAR, R,     &
                   LR, STATE, X, LDX, IFAIL)

INTEGER           MODE, N, K, IP, IQ, LDVAR, LR, STATE(*), LDX, IFAIL
REAL (KIND=nag_wp) XMEAN(K), PHI(K*K*IP), THETA(K*K*IQ), VAR(LDVAR,K),       &
                   R(LR), X(LDX,N)
```

3   Description

Let the vector $X_t = (x_{1t}, x_{2t}, \ldots, x_{kt})^{\mathrm{T}}$, denote a k-dimensional time series which is assumed to follow a vector autoregressive moving average (VARMA) model of the form:

$$
\begin{aligned}
X_t - \mu = \quad & \phi_1(X_{t-1} - \mu) + \phi_2(X_{t-2} - \mu) + \cdots + \phi_p(X_{t-p} - \mu) + \\
& \epsilon_t - \theta_1\epsilon_{t-1} - \theta_2\epsilon_{t-2} - \cdots - \theta_q\epsilon_{t-q}
\end{aligned}
\tag{1}
$$

where $\epsilon_t = (\epsilon_{1t}, \epsilon_{2t}, \ldots, \epsilon_{kt})^{\mathrm{T}}$, is a vector of k residual series assumed to be Normally distributed with zero mean and covariance matrix Σ. The components of ϵ_t are assumed to be uncorrelated at non-simultaneous lags. The ϕ_i's and θ_j's are k by k matrices of parameters. $\{\phi_i\}$, for $i = 1, 2, \ldots, p$, are called the autoregressive (AR) parameter matrices, and $\{\theta_j\}$, for $j = 1, 2, \ldots, q$, the moving average (MA) parameter matrices. The parameters in the model are thus the p k by k ϕ-matrices, the q k by k θ-matrices, the mean vector μ and the residual error covariance matrix Σ. Let

$$
A(\phi) = \begin{bmatrix}
\phi_1 & I & 0 & . & . & . & 0 \\
\phi_2 & 0 & I & 0 & . & . & 0 \\
. & & & & . & & \\
. & & & & & . & \\
. & & & & & . & \\
\phi_{p-1} & 0 & . & . & . & 0 & I \\
\phi_p & 0 & . & . & . & 0 & 0
\end{bmatrix}_{pk \times pk}
\quad \text{and} \quad
B(\theta) = \begin{bmatrix}
\theta_1 & I & 0 & . & . & . & 0 \\
\theta_2 & 0 & I & 0 & . & . & 0 \\
. & & & & . & & \\
. & & & & & . & \\
. & & & & & . & \\
\theta_{q-1} & 0 & . & . & . & 0 & I \\
\theta_q & 0 & . & . & . & 0 & 0
\end{bmatrix}_{qk \times qk}
$$

where I denotes the k by k identity matrix.

The model (1) must be both stationary and invertible. The model is said to be stationary if the eigenvalues of $A(\phi)$ lie inside the unit circle and invertible if the eigenvalues of $B(\theta)$ lie inside the unit circle.

For $k \geq 6$ the VARMA model (1) is recast into state space form and a realization of the state vector at time zero computed. For all other cases the routine computes a realization of the pre-observed vectors $X_0, X_{-1}, \ldots, X_{1-p}, \epsilon_0, \epsilon_{-1}, \ldots, \epsilon_{1-q}$, from (1), see Shea (1988). This realization is then used to generate a sequence of successive time series observations. Note that special action is taken for pure MA models, that is for $p = 0$.

At your request a new realization of the time series may be generated more efficiently using the information in a reference vector created during a previous call to G05PJF. See the description of the argument MODE in Section 5 for details.

The routine returns a realization of $X_1, X_2, \ldots, X_n$. On a successful exit, the recent history is updated and saved in the array R so that G05PJF may be called again to generate a realization of $X_{n+1}, X_{n+2}, \ldots$, etc. See the description of the argument MODE in Section 5 for details.

Further computational details are given in Shea (1988). Note, however, that G05PJF uses a spectral decomposition rather than a Cholesky factorization to generate the multivariate Normals. Although this method involves more multiplications than the Cholesky factorization method and is thus slightly slower it is more stable when faced with ill-conditioned covariance matrices. A method of assigning the AR and MA coefficient matrices so that the stationarity and invertibility conditions are satisfied is described in Barone (1987).

One of the initialization routines G05KFF (for a repeatable sequence if computed sequentially) or G05KGF (for a non-repeatable sequence) must be called prior to the first call to G05PJF.

4 References

Barone P (1987) A method for generating independent realisations of a multivariate normal stationary and invertible ARMA(p, q) process *J. Time Ser. Anal.* **8** 125–130

Shea B L (1988) A note on the generation of independent realisations of a vector autoregressive moving average process *J. Time Ser. Anal.* **9** 403–410

5 Arguments

1:   MODE – INTEGER   *Input*

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

MODE = 0
    Set up reference vector and compute a realization of the recent history.

MODE = 1
    Generate terms in the time series using reference vector set up in a prior call to G05PJF.

MODE = 2
    Combine the operations of MODE = 0 and 1.

MODE = 3
    A new realization of the recent history is computed using information stored in the reference vector, and the following sequence of time series values are generated.

If MODE = 1 or 3, then you must ensure that the reference vector R and the values of K, IP, IQ, XMEAN, PHI, THETA, VAR and LDVAR have not been changed between calls to G05PJF.

Constraint: MODE = 0, 1, 2 or 3.

2:   N – INTEGER   *Input*

On entry: n, the number of observations to be generated.

Constraint: N $\geq$ 0.

3:   K – INTEGER   *Input*

On entry: k, the dimension of the multivariate time series.

Constraint: K $\geq$ 1.

4:   XMEAN(K) – REAL (KIND=nag_wp) array   *Input*

On entry: μ, the vector of means of the multivariate time series.

5: IP – INTEGER *Input*

On entry: p, the number of autoregressive parameter matrices.

Constraint: $IP \geq 0$.

6: PHI$(K \times K \times IP)$ – REAL (KIND=nag_wp) array *Input*

On entry: must contain the elements of the $IP \times K \times K$ autoregressive parameter matrices of the model, $\phi_1, \phi_2, \ldots, \phi_p$. If PHI is considered as a three-dimensional array, dimensioned as PHI(K, K, IP), then the (i, j)th element of ϕ_l would be stored in PHI(i, j, l); that is, PHI$((l - 1) \times k \times k + (j - 1) \times k + i)$ must be set equal to the (i, j)th element of ϕ_l, for $l = 1, 2, \ldots, p$, $i = 1, 2, \ldots, k$ and $j = 1, 2, \ldots, k$.

Constraint: the elements of PHI must satisfy the stationarity condition.

7: IQ – INTEGER *Input*

On entry: q, the number of moving average parameter matrices.

Constraint: $IQ \geq 0$.

8: THETA$(K \times K \times IQ)$ – REAL (KIND=nag_wp) array *Input*

On entry: must contain the elements of the $IQ \times K \times K$ moving average parameter matrices of the model, $\theta_1, \theta_2, \ldots, \theta_q$. If THETA is considered as a three-dimensional array, dimensioned as THETA(K,K,IQ), then the (i, j)th element of θ_l would be stored in THETA(i, j, l); that is, THETA$((l - 1) \times k \times k + (j - 1) \times k + i)$ must be set equal to the (i, j)th element of θ_l, for $l = 1, 2, \ldots, q$, $i = 1, 2, \ldots, k$ and $j = 1, 2, \ldots, k$.

Constraint: the elements of THETA must be within the invertibility region.

9: VAR$(LDVAR, K)$ – REAL (KIND=nag_wp) array *Input*

On entry: VAR(i, j) must contain the (i, j)th element of Σ, for $i = 1, 2, \ldots, K$ and $j = 1, 2, \ldots, K$. Only the lower triangle is required.

Constraint: the elements of VAR must be such that Σ is positive semidefinite.

10: LDVAR – INTEGER *Input*

On entry: the first dimension of the array VAR as declared in the (sub)program from which G05PJF is called.

Constraint: $LDVAR \geq K$.

11: R(LR) – REAL (KIND=nag_wp) array *Communication Array*

On entry: if MODE $= 1$ or 3, the array R as output from the previous call to G05PJF must be input without any change.

If MODE $= 0$ or 2, the contents of R need not be set.

On exit: information required for any subsequent calls to the routine with MODE $= 1$ or 3. See Section 9.

12: LR – INTEGER *Input*

On entry: the dimension of the array R as declared in the (sub)program from which G05PJF is called.

Constraints:

if $K \geq 6$, $LR \geq (5r^2 + 1) \times K^2 + (4r + 3) \times K + 4$;

if $K < 6$, $LR \geq ((IP + IQ)^2 + 1) \times K^2 +$

$(4 \times (IP + IQ) + 3) \times K + \max\{Kr(Kr + 2), K^2(IP + IQ)^2 + l(l + 3) + K^2(IQ + 1)\} + 4$.

W h e r e $r = \max(\text{IP}, \text{IQ})$ a n d i f $\text{IP} = 0$, $l = \text{K}(\text{K} + 1)/2$, o r i f $\text{IP} \geq 1$, $l = \text{K}(\text{K} + 1)/2 + (\text{IP} - 1)\text{K}^2$.

See Section 9 for some examples of the required size of the array R.

13: STATE($*$) – INTEGER array   *Communication Array*

Note: the actual argument supplied **must** be the array STATE supplied to the initialization routines G05KFF or G05KGF.

On entry: contains information on the selected base generator and its current state.

On exit: contains updated information on the state of the generator.

14: X(LDX, N) – REAL (KIND=nag_wp) array   *Output*

On exit: $\text{X}(i, t)$ will contain a realization of the ith component of X_t, for $i = 1, 2, \ldots, k$ and $t = 1, 2, \ldots, n$.

15: LDX – INTEGER   *Input*

On entry: the first dimension of the array X as declared in the (sub)program from which G05PJF is called.

Constraint: $\text{LDX} \geq \text{K}$.

16: IFAIL – INTEGER   *Input/Output*

On entry: IFAIL must be set to 0, -1 or 1. If you are unfamiliar with this argument you should refer to Section 3.4 in How to Use the NAG Library and its Documentation for details.

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, if you are not familiar with this argument, the recommended value is 0. **When the value -1 or 1 is used it is essential to test the value of IFAIL on exit.**

On exit: $\text{IFAIL} = 0$ unless the routine detects an error or a warning has been flagged (see Section 6).

6   Error Indicators and Warnings

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

Errors or warnings detected by the routine:

$\text{IFAIL} = 1$

On entry, $\text{MODE} = \langle value \rangle$.
Constraint: $\text{MODE} = 0$, 1, 2 or 3.

$\text{IFAIL} = 2$

On entry, $\text{N} = \langle value \rangle$.
Constraint: $\text{N} \geq 0$.

$\text{IFAIL} = 3$

On entry, $\text{K} = \langle value \rangle$.
Constraint: $\text{K} \geq 1$.

IFAIL = 5

On entry, IP = ⟨*value*⟩.
Constraint: IP ≥ 0.

IFAIL = 6

On entry, the AR parameters are outside the stationarity region.

IFAIL = 7

On entry, IQ = ⟨*value*⟩.
Constraint: IQ ≥ 0.

IFAIL = 8

On entry, the moving average parameter matrices are such that the model is non-invertible.

IFAIL = 9

On entry, the covariance matrix VAR is not positive semidefinite to **machine precision**.

IFAIL = 10

On entry, LDVAR = ⟨*value*⟩ and K = ⟨*value*⟩.
Constraint: LDVAR ≥ K.

IFAIL = 11

K is not the same as when R was set up in a previous call.
Previous value of K = ⟨*value*⟩ and K = ⟨*value*⟩.

IFAIL = 12

On entry, LR is not large enough, LR = ⟨*value*⟩: minimum length required = ⟨*value*⟩.

IFAIL = 13

On entry, STATE vector has been corrupted or not initialized.

IFAIL = 15

On entry, LDX = ⟨*value*⟩ and K = ⟨*value*⟩.
Constraint: LDX ≥ K.

IFAIL = 20

An excessive number of iterations were required by the NAG routine used to evaluate the eigenvalues of the matrices used to test for stationarity or invertibility.

IFAIL = 21

The reference vector cannot be computed because the AR parameters are too close to the boundary of the stationarity region.

IFAIL = 22

An excessive number of iterations were required by the NAG routine used to evaluate the eigenvalues of the covariance matrix.

IFAIL = 23

An excessive number of iterations were required by the NAG routine used to evaluate the eigenvalues stored in the reference vector.

IFAIL $= -99$

An unexpected error has been triggered by this routine. Please contact NAG.

See Section 3.9 in How to Use the NAG Library and its Documentation for further information.

IFAIL $= -399$

Your licence key may have expired or may not have been installed correctly.

See Section 3.8 in How to Use the NAG Library and its Documentation for further information.

IFAIL $= -999$

Dynamic memory allocation failed.

See Section 3.7 in How to Use the NAG Library and its Documentation for further information.

7   Accuracy

The accuracy is limited by the matrix computations performed, and this is dependent on the condition of the argument and covariance matrices.

8   Parallelism and Performance

G05PJF is threaded by NAG for parallel execution in multithreaded implementations of the NAG Library.

G05PJF makes calls to BLAS and/or LAPACK routines, which may be threaded within the vendor library used by this implementation. Consult the documentation for the vendor library for further information.

Please consult the X06 Chapter Introduction for information on how to control and interrogate the OpenMP environment used within this routine. Please also consult the Users' Note for your implementation for any additional implementation-specific information.

9   Further Comments

Note that, in reference to IFAIL $= 8$, G05PJF will permit moving average parameters on the boundary of the invertibility region.

The elements of R contain amongst other information details of the spectral decompositions which are used to generate future multivariate Normals. Note that these eigenvectors may not be unique on different machines. For example the eigenvectors corresponding to multiple eigenvalues may be permuted. Although an effort is made to ensure that the eigenvectors have the same sign on all machines, differences in the signs may theoretically still occur.

The following table gives some examples of the required size of the array R, specified by the argument LR, for $k = 1, 2$ or 3, and for various values of p and q.

		q			
		0	1	2	3
	0	13	20	31	46
		36	56	92	144
		85	124	199	310
	1	19	30	45	64
		52	88	140	208
		115	190	301	448
p	2	35	50	69	92
		136	188	256	340
		397	508	655	838
	3	57	76	99	126
		268	336	420	520
		877	1024	1207	1426

Note that G13DXF may be used to check whether a VARMA model is stationary and invertible.

The time taken depends on the values of p, q and especially n and k.

10   Example

This program generates two realizations, each of length 48, from the bivariate AR(1) model

$$X_t - \mu = \phi_1(X_{t-1} - \mu) + \epsilon_t$$

with

$$\phi_1 = \begin{bmatrix} 0.80 & 0.07 \\ 0.00 & 0.58 \end{bmatrix},$$

$$\mu = \begin{bmatrix} 5.00 \\ 9.00 \end{bmatrix},$$

and

$$\Sigma = \begin{bmatrix} 2.97 & 0 \\ 0.64 & 5.38 \end{bmatrix}.$$

The pseudorandom number generator is initialized by a call to G05KFF. Then, in the first call to G05PJF, MODE = 2 in order to set up the reference vector before generating the first realization. In the subsequent call MODE = 3 and a new recent history is generated and used to generate the second realization.

10.1 Program Text

```
    Program g05pjfe

!     G05PJF Example Program Text

!     Mark 26 Release. NAG Copyright 2016.

!       .. Use Statements ..
      Use nag_library, Only: g05kff, g05pjf, nag_wp
!       .. Implicit None Statement ..
      Implicit None
!       .. Parameters ..
      Integer, Parameter                :: lseed = 1, nin = 5, nout = 6
!       .. Local Scalars ..
      Integer                           :: genid, i, ifail, ii, ip, iq, j, k,   &
```

```
                                                  k2, l, ldvar, ldx, lphi, lr, lstate, &
                                                  ltheta, mode, n, nreal, rn, subid
!       .. Local Arrays ..
        Real (Kind=nag_wp), Allocatable :: phi(:), r(:), theta(:), var(:,:),    &
                                                  x(:,:), xmean(:)
        Integer                            :: seed(lseed)
        Integer, Allocatable               :: state(:)
!       .. Intrinsic Procedures ..
        Intrinsic                          :: max
!       .. Executable Statements ..
        Write (nout,*) 'G05PJF Example Program Results'
        Write (nout,*)

!       Skip heading in data file
        Read (nin,*)

!       Read in the base generator information and seed
        Read (nin,*) genid, subid, seed(1)

!       Initial call to initializer to get size of STATE array
        lstate = 0
        Allocate (state(lstate))
        ifail = 0
        Call g05kff(genid,subid,seed,lseed,state,lstate,ifail)

!       Reallocate STATE
        Deallocate (state)
        Allocate (state(lstate))

!       Initialize the generator to a repeatable sequence
        ifail = 0
        Call g05kff(genid,subid,seed,lseed,state,lstate,ifail)

!       Read in the sample size and number of realizations
        Read (nin,*) n, nreal

!       Read in the number of coefficients
        Read (nin,*) k, ip, iq

        k2 = k**2
        rn = max(ip,iq)
        l = k*(k+1)/2
        If (ip>0) Then
          l = l + (ip-1)*k2
        End If
        If (k>=6) Then
          lr = (5*rn**2+1)*k2 + (4*rn+3) + 4
        Else
          lr = ((ip+iq)**2+1)*k2 + (4*(ip+iq)+3)*k + max(k*rn*(k*rn+2),k2*(ip+iq &
          )**2+l*(l+3)+k2*(iq+1)) + 4
        End If
        lphi = ip*k*k
        ltheta = iq*k*k
        ldvar = k
        ldx = k
        Allocate (phi(lphi),theta(ltheta),var(ldvar,k),r(lr),x(ldx,n),xmean(k))

!       Read in the AR parameters
        Do l = 1, ip
          Do i = 1, k
            ii = (l-1)*k*k + i
            Read (nin,*)(phi(ii+k*(j-1)),j=1,k)
          End Do
        End Do

!       Read in the MA parameters
        Do l = 1, iq
          Do i = 1, k
            ii = (l-1)*k*k + i
            Read (nin,*)(theta(ii+k*(j-1)),j=1,k)
          End Do
```

```
        End Do

!       Read in the means
        Read (nin,*) xmean(1:k)

!       Read in the variance / covariance matrix
        Read (nin,*)(var(i,1:i),i=1,k)

!       For the first realization we need to set up the reference vector
!       as well as generate the series
        mode = 2

!       Generate NREAL realizations
d_lp: Do rn = 1, nreal

          ifail = 0
          Call g05pjf(mode,n,k,xmean,ip,phi,iq,theta,var,ldvar,r,lr,state,x,ldx, &
            ifail)

!         Display the results
          Write (nout,99999) ' Realization Number ', rn
          Do i = 1, k
            Write (nout,*)
            Write (nout,99999) '  Series number ', i
            Write (nout,*) '  -------------'
            Write (nout,*)
            Write (nout,99998) x(i,1:n)
          End Do
          Write (nout,*)

!         For subsequent realizations we use previous reference vector
          mode = 3

        End Do d_lp

99999 Format (1X,A,I0)
99998 Format (8(2X,F8.3))
    End Program g05pjfe
```

10.2 Program Data

```
G05PJF Example Program Data
1  1  1762543      :: GENID,SUBID,SEED(1)
48 2               :: N,NREAL
2  1  0            :: K, IP, IQ
0.80  0.07
0.00  0.58         :: End of PHI
5.00  9.00         :: XMEAN
2.97
0.64  5.38         :: End of VAR (lower triangle)
```

10.3 Program Results

```
 G05PJF Example Program Results

  Realization Number 1

   Series number 1
   -------------

     4.833     2.813     3.224     3.825     1.023     1.415     2.184     3.005
     5.547     4.832     4.705     5.484     9.407    10.335     8.495     7.478
     6.373     6.692     6.698     6.976     6.200     4.458     2.520     3.517
     3.054     5.439     5.699     7.136     5.750     8.497     9.563    11.604
     9.020    10.063     7.976     5.927     4.992     4.222     3.982     7.107
     3.554     7.045     7.025     4.106     5.106     5.954     8.026     7.212

   Series number 2
   -------------
```

```
   8.458      9.140     10.866     10.975      9.245      5.054      5.023     12.486
  10.534     10.590     11.376      8.793     14.445     13.237     11.030      8.405
   7.187      8.291      5.920      9.390     10.055      6.222      7.751     10.604
  12.441     10.664     10.960      8.022     10.073     12.870     12.665     14.064
  11.867     12.894     10.546     12.754      8.594      9.042     12.029     12.557
   9.746      5.487      5.500      8.629      9.723      8.632      6.383     12.484

 Realization Number 2

  Series number 1
  -------------

   5.396      4.811      2.685      5.824      2.449      3.563      5.663      6.209
   3.130      4.308      4.333      4.903      1.770      1.278      1.340     -0.527
   1.745      3.211      4.478      5.170      5.365      4.852      6.080      6.464
   2.765      2.148      6.641      7.224     10.316      7.102      5.604      3.934
   4.839      3.698      5.210      5.384      7.652      7.315      7.332      7.561
   7.537      7.788      6.868      7.575      6.108      6.188      8.132     10.310

  Series number 2
  -------------

  11.345     10.070     13.654     12.409     11.329     13.054     12.465      9.867
  10.263     13.394     10.553     10.331      7.814      8.747     10.025     11.167
  10.626      9.366      9.607      9.662     10.492     10.766     11.512     10.813
  10.799      8.780      9.221     14.245     11.575     10.620      8.282      5.447
   9.935      9.386     11.627     10.066     11.394      7.951      7.907     12.616
  15.246      9.962     13.216     11.350     11.227      6.021      6.968     12.428
```