# nag\_ref\_vec\_multi\_normal (g05eac)

# 1. Purpose

**nag\_ref\_vec\_multi\_normal** (g05eac) sets up a reference vector for a multivariate Normal distribution with mean vector a and variance-covariance matrix C, so that **nag\_ref\_vec\_multi\_normal** (g05eac) may be used to generate pseudo-random vectors.

#### 2. Specification

# 3. Description

When the variance-covariance matrix is non-singular (i.e., strictly positive-definite), the distribution has probability density function

$$f(x) = \sqrt{\frac{|C^{-1}|}{(2\pi)^n}} \exp\left\{-(x-a)^T C^{-1}(x-a)\right\}$$

where n is the number of dimensions, C is the variance-covariance matrix, a is the vector of means and x is the vector of positions.

Variance-covariance matrices are symmetric and positive semi-definite. Given such a matrix C, there exists a lower triangular matrix L such that  $LL^T = C$ . L is not unique, if C is singular.

nag\_ref\_vec\_multi\_normal decomposes C to find such an L. It then stores n, a and L in the reference vector r for later use by nag\_return\_multi\_normal (g05ezc). nag\_return\_multi\_normal (g05ezc) generates a vector x of independent standard Normal pseudo-random numbers. It then returns the vector a + Lx, which has the required multivariate Normal distribution.

It should be noted that this routine will work with a singular variance-covariance matrix C, provided C is positive semi-definite, despite the fact that the above formula for the probability density function is not valid in that case. Wilkinson (1965) should be consulted if further information is required.

#### 4. Parameters

a[n]

Input: the vector of means, a, of the distribution.

 $\mathbf{n}$ 

Input: the number of dimensions, n, of the distribution.

Constraint:  $\mathbf{n} > 0$ .

c[n][tdc]

Input: the variance-covariance matrix of the distribution. Only the upper triangle need be set

tdc

Input: the second dimension of the array  $\mathbf{c}$  as declared in the function from which nag\_ref\_vec\_multi\_normal is called.

Constraint:  $tdc \geq n$ .

eps

Input: the maximum error in any element of C, relative to the largest element of C. Constraint:  $0.0 \le \text{eps} \le 0.1/\text{n}$ .

 $\mathbf{r}$ 

Output: reference vector for which memory will be allocated internally. This reference vector will subsequently be used by nag\_return\_multi\_normal (g05ezc). If no memory is allocated to  $\bf r$  (e.g. when an input error is detected) then  $\bf r$  will be NULL on return.

[NP3275/5/pdf] 3.g05eac.1

fail

The NAG error parameter, see the Essential Introduction to the NAG C Library.

# 5. Error Indications and Warnings

#### NE\_INT\_ARG\_LT

On entry, **n** must not be less than 1:  $\mathbf{n} = \langle value \rangle$ .

#### NE\_2\_INT\_ARG\_LT

On entry,  $\mathbf{tdc} = \langle value \rangle$  while  $\mathbf{n} = \langle value \rangle$ . These parameters must satisfy  $\mathbf{tdc} \geq \mathbf{n}$ .

#### NE\_REAL\_ARG\_LT

On entry, **eps** must not be less than 0.0: **eps** =  $\langle value \rangle$ .

#### NE\_2\_REAL\_ARG\_GT

On entry,  $eps = \langle value \rangle$  while  $0.1/n = \langle value \rangle$ . These parameters must satisfy  $eps \leq 0.1/n$ .

#### NE\_ALLOC\_FAIL

Memory allocation failed.

# NE\_NOT\_POS\_SEM\_DEF

Matrix C is not positive semi-definite.

#### 6. Further Comments

The time taken by the routine is of order  $n^3$ .

It is recommended that the diagonal elements of C should not differ too widely in order of magnitude. This may be achieved by scaling the variables if necessary. The actual matrix decomposed is  $C + E = LL^T$ , where E is a diagonal matrix with small positive diagonal elements. This ensures that, even when C is singular, or nearly singular, the Cholesky Factor L corresponds to a positive-definite variance-covariance matrix that agrees with C within a tolerance determined by  $\operatorname{eps}$ .

#### 6.1. Accuracy

The maximum absolute error in  $LL^T$ , and hence in the variance-covariance matrix of the resulting vectors, is less than  $(n \times \max(\mathbf{eps}, \varepsilon) + (n+3)\varepsilon/2)$  times the maximum element of C, where  $\varepsilon$  is the **machine precision**. Under normal circumstances, the above will be small compared to sampling error.

# 6.2. References

Knuth D E (1981) The Art of Computer Programming (Vol 2) (2nd Edn) Addison-Wesley. Wilkinson J H (1965) The Algebraic Eigenvalue Problem Clarendon Press, Oxford.

#### 7. See Also

nag\_random\_init\_repeatable (g05cbc) nag\_random\_init\_nonrepeatable (g05ccc) nag\_random\_normal (g05ddc) nag\_return\_multi\_normal (g05ezc)

#### 8. Example

The example program prints five pseudo-random observations from a bivariate Normal distribution with means vector

$$\begin{bmatrix} 1.0 \\ 2.0 \end{bmatrix}$$

and variance-covariance matrix

$$\begin{bmatrix} 2.0 & 1.0 \\ 1.0 & 3.0 \end{bmatrix},$$

generated by nag\_ref\_vec\_multi\_normal and nag\_return\_multi\_normal (g05ezc) after initialisation by nag\_random\_init\_repeatable (g05ebc).

3.905eac. 2 [NP3275/5/pdf]

#### 8.1. Program Text

```
/* nag_ref_vec_multi_normal(g05eac) Example Program
   Copyright 1991 Numerical Algorithms Group.
 * Mark 2, 1991.
 * Mark 3 revised, 1994.
#include <nag.h>
#include <stdio.h>
#include <nag_stdlib.h>
#include <nagg05.h>
#define N 2
#define TDC N
main()
{
  Integer i, j;
double a[N], c[N][TDC], z[N];
double *r = (double *)0;
  double eps = 0.01;
  Vprintf("g05eac Example Program Results\n");
  a[0] = 1.0;
  a[1] = 2.0;

c[0][0] = 2.0;
  c[1][1] = 3.0;
  c[0][1] = 1.0;
  c[1][0] = 1.0;
  g05cbc((Integer)0);
  g05eac(a, (Integer)N, (double *)c, (Integer)TDC,
          eps, &r, NAGERR_DEFAULT);
  for (i=1; i<=5; i++)
       g05ezc(z, r);
       for (j=0; j<2; j++)
Vprintf("%10.4f",z[j]);
       Vprintf("\n");
  NAG_FREE(r);
  exit(EXIT_SUCCESS);
```

# 8.2. Program Data

None.

# 8.3. Program Results

[NP3275/5/pdf] 3.g05eac.3