

NAG Library Function Document

nag_rand_field_2d_predef_setup (g05zrc)

1 Purpose

nag_rand_field_2d_predef_setup (g05zrc) performs the setup required in order to simulate stationary Gaussian random fields in two dimensions, for a preset variogram, using the *circulant embedding method*. Specifically, the eigenvalues of the extended covariance matrix (or embedding matrix) are calculated, and their square roots output, for use by nag_rand_field_2d_generate (g05zsc), which simulates the random field.

2 Specification

```
#include <nag.h>
#include <nagg05.h>

void nag_rand_field_2d_predef_setup (const Integer ns[], double xmin,
    double xmax, double ymin, double ymax, const Integer maxm[], double var,
    Nag_Variogram cov, Nag_NormType norm, Integer np, const double params[],
    Nag_EmbedPad pad, Nag_EmbedScale corr, double lam[], double xx[],
    double yy[], Integer m[], Integer *approx, double *rho, Integer *icount,
    double eig[], NagError *fail)
```

3 Description

A two-dimensional random field $Z(\mathbf{x})$ in \mathbb{R}^2 is a function which is random at every point $\mathbf{x} \in \mathbb{R}^2$, so $Z(\mathbf{x})$ is a random variable for each \mathbf{x} . The random field has a mean function $\mu(\mathbf{x}) = \mathbb{E}[Z(\mathbf{x})]$ and a symmetric positive semidefinite covariance function $C(\mathbf{x}, \mathbf{y}) = \mathbb{E}[(Z(\mathbf{x}) - \mu(\mathbf{x}))(Z(\mathbf{y}) - \mu(\mathbf{y}))]$. $Z(\mathbf{x})$ is a Gaussian random field if for any choice of $n \in \mathbb{N}$ and $\mathbf{x}_1, \dots, \mathbf{x}_n \in \mathbb{R}^2$, the random vector $[Z(\mathbf{x}_1), \dots, Z(\mathbf{x}_n)]^T$ follows a multivariate Normal distribution, which would have a mean vector $\tilde{\boldsymbol{\mu}}$ with entries $\tilde{\mu}_i = \mu(\mathbf{x}_i)$ and a covariance matrix \tilde{C} with entries $\tilde{C}_{ij} = C(\mathbf{x}_i, \mathbf{x}_j)$. A Gaussian random field $Z(\mathbf{x})$ is stationary if $\mu(\mathbf{x})$ is constant for all $\mathbf{x} \in \mathbb{R}^2$ and $C(\mathbf{x}, \mathbf{y}) = C(\mathbf{x} + \mathbf{a}, \mathbf{y} + \mathbf{a})$ for all $\mathbf{x}, \mathbf{y}, \mathbf{a} \in \mathbb{R}^2$ and hence we can express the covariance function $C(\mathbf{x}, \mathbf{y})$ as a function γ of one variable: $C(\mathbf{x}, \mathbf{y}) = \gamma(\mathbf{x} - \mathbf{y})$. γ is known as a variogram (or more correctly, a semivariogram) and includes the multiplicative factor σ^2 representing the variance such that $\gamma(0) = \sigma^2$.

The functions nag_rand_field_2d_predef_setup (g05zrc) and nag_rand_field_2d_generate (g05zsc) are used to simulate a two-dimensional stationary Gaussian random field, with mean function zero and variogram $\gamma(\mathbf{x})$, over a domain $[x_{\min}, x_{\max}] \times [y_{\min}, y_{\max}]$, using an equally spaced set of $N_1 \times N_2$ points; N_1 points in the x -direction and N_2 points in the y -direction. The problem reduces to sampling a Gaussian random vector \mathbf{X} of size $N_1 \times N_2$, with mean vector zero and a symmetric covariance matrix A , which is an N_2 by N_2 block Toeplitz matrix with Toeplitz blocks of size N_1 by N_1 . Since A is in general expensive to factorize, a technique known as the *circulant embedding method* is used. A is embedded into a larger, symmetric matrix B , which is an M_2 by M_2 block circulant matrix with circulant blocks of size M_1 by M_1 , where $M_1 \geq 2(N_1 - 1)$ and $M_2 \geq 2(N_2 - 1)$. B can now be factorized as $B = WAW^* = R^*R$, where W is the two-dimensional Fourier matrix (W^* is the complex conjugate of W), A is the diagonal matrix containing the eigenvalues of B and $R = A^{\frac{1}{2}}W^*$. B is known as the embedding matrix. The eigenvalues can be calculated by performing a discrete Fourier transform of the first row (or column) of B and multiplying by $M_1 \times M_2$, and so only the first row (or column) of B is needed – the whole matrix does not need to be formed.

As long as all of the values of A are non-negative (i.e., B is positive semidefinite), B is a covariance matrix for a random vector \mathbf{Y} which has M_2 blocks of size M_1 . Two samples of \mathbf{Y} can now be simulated from the real and imaginary parts of $R^*(\mathbf{U} + i\mathbf{V})$, where \mathbf{U} and \mathbf{V} have elements from the standard Normal distribution. Since $R^*(\mathbf{U} + i\mathbf{V}) = W A^{\frac{1}{2}}(\mathbf{U} + i\mathbf{V})$, this calculation can be done using a discrete Fourier transform of the vector $A^{\frac{1}{2}}(\mathbf{U} + i\mathbf{V})$. Two samples of the random vector \mathbf{X} can now be recovered

by taking the first N_1 elements of the first N_2 blocks of each sample of \mathbf{Y} – because the original covariance matrix A is embedded in B , \mathbf{X} will have the correct distribution.

If B is not positive semidefinite, larger embedding matrices B can be tried; however if the size of the matrix would have to be larger than **maxm**, an approximation procedure is used. We write $A = A_+ + A_-$, where A_+ and A_- contain the non-negative and negative eigenvalues of B respectively. Then B is replaced by ρB_+ where $B_+ = W A_+ W^*$ and $\rho \in (0, 1]$ is a scaling factor. The error ϵ in approximating the distribution of the random field is given by

$$\epsilon = \sqrt{\frac{(1 - \rho)^2 \text{trace } A + \rho^2 \text{trace } A_-}{M}}.$$

Three choices for ρ are available, and are determined by the input argument **corr**:

setting **corr** = Nag_EmbedScaleTraces sets

$$\rho = \frac{\text{trace } A}{\text{trace } A_+},$$

setting **corr** = Nag_EmbedScaleSqrtTraces sets

$$\rho = \sqrt{\frac{\text{trace } A}{\text{trace } A_+}},$$

setting **corr** = Nag_EmbedScaleOne sets $\rho = 1$.

nag_rand_field_2d_predef_setup (g05zrc) finds a suitable positive semidefinite embedding matrix B and outputs its sizes in the vector **m** and the square roots of its eigenvalues in **lam**. If approximation is used, information regarding the accuracy of the approximation is output. Note that only the first row (or column) of B is actually formed and stored.

4 References

Dietrich C R and Newsam G N (1997) Fast and exact simulation of stationary Gaussian processes through circulant embedding of the covariance matrix *SIAM J. Sci. Comput.* **18** 1088–1107

Schlather M (1999) Introduction to positive definite functions and to unconditional simulation of random fields *Technical Report ST 99–10* Lancaster University

Wood A T A and Chan G (1997) Algorithm AS 312: An Algorithm for Simulating Stationary Gaussian Random Fields *Journal of the Royal Statistical Society, Series C (Applied Statistics) (Volume 46)* **1** 171–181

5 Arguments

1: **ns[2]** – const Integer *Input*

On entry: the number of sample points to use in each direction, with **ns[0]** sample points in the x -direction, N_1 and **ns[1]** sample points in the y -direction, N_2 . The total number of sample points on the grid is therefore **ns[0]** \times **ns[1]**.

Constraints:

$$\begin{aligned} \mathbf{ns}[0] &\geq 1; \\ \mathbf{ns}[1] &\geq 1. \end{aligned}$$

2: **xmin** – double *Input*

On entry: the lower bound for the x -coordinate, for the region in which the random field is to be simulated.

Constraint: **xmin** < **xmax**.

- 3: **xmax** – double *Input*
On entry: the upper bound for the x -coordinate, for the region in which the random field is to be simulated.
Constraint: **xmin** < **xmax**.
- 4: **ymin** – double *Input*
On entry: the lower bound for the y -coordinate, for the region in which the random field is to be simulated.
Constraint: **ymin** < **ymax**.
- 5: **ymax** – double *Input*
On entry: the upper bound for the y -coordinate, for the region in which the random field is to be simulated.
Constraint: **ymin** < **ymax**.
- 6: **maxm[2]** – const Integer *Input*
On entry: determines the maximum size of the circulant matrix to use – a maximum of **maxm[0]** elements in the x -direction, and a maximum of **maxm[1]** elements in the y -direction. The maximum size of the circulant matrix is thus **maxm[0]** × **maxm[1]**.
Constraint: **maxm**[i] ≥ 2^k , where k is the smallest integer satisfying $2^k \geq 2(\mathbf{ns}[i] - 1)$, for $i = 0, 1$.
- 7: **var** – double *Input*
On entry: the multiplicative factor σ^2 of the variogram $\gamma(\mathbf{x})$.
Constraint: **var** ≥ 0.0.
- 8: **cov** – Nag_Variogram *Input*
On entry: determines which of the preset variograms to use. The choices are given below. Note that $x' = \left\| \frac{x}{\ell_1}, \frac{y}{\ell_2} \right\|$, where ℓ_1 and ℓ_2 are correlation lengths in the x and y directions respectively and are parameters for most of the variograms, and σ^2 is the variance specified by **var**.

cov = Nag_VgmSymmStab
 Symmetric stable variogram

$$\gamma(\mathbf{x}) = \sigma^2 \exp(-(x')^\nu),$$

where

$$\begin{aligned} \ell_1 &= \mathbf{params}[0], \ell_1 > 0, \\ \ell_2 &= \mathbf{params}[1], \ell_2 > 0, \\ \nu &= \mathbf{params}[2], 0 < \nu \leq 2. \end{aligned}$$

cov = Nag_VgmCauchy
 Cauchy variogram

$$\gamma(\mathbf{x}) = \sigma^2 \left(1 + (x')^2\right)^{-\nu},$$

where

$$\begin{aligned} \ell_1 &= \mathbf{params}[0], \ell_1 > 0, \\ \ell_2 &= \mathbf{params}[1], \ell_2 > 0, \\ \nu &= \mathbf{params}[2], \nu > 0. \end{aligned}$$

cov = Nag_VgmDifferential
Differential variogram with compact support

$$\gamma(\mathbf{x}) = \begin{cases} \sigma^2 \left(1 + 8x' + 25(x')^2 + 32(x')^3\right)(1 - x')^8, & x' < 1, \\ 0, & x' \geq 1, \end{cases}$$

where

$$\ell_1 = \mathbf{params}[0], \ell_1 > 0,$$

$$\ell_2 = \mathbf{params}[1], \ell_2 > 0.$$

cov = Nag_VgmExponential
Exponential variogram

$$\gamma(\mathbf{x}) = \sigma^2 \exp(-x'),$$

where

$$\ell_1 = \mathbf{params}[0], \ell_1 > 0,$$

$$\ell_2 = \mathbf{params}[1], \ell_2 > 0.$$

cov = Nag_VgmGauss
Gaussian variogram

$$\gamma(\mathbf{x}) = \sigma^2 \exp\left(-(x')^2\right),$$

where

$$\ell_1 = \mathbf{params}[0], \ell_1 > 0,$$

$$\ell_2 = \mathbf{params}[1], \ell_2 > 0.$$

cov = Nag_VgmNugget
Nugget variogram

$$\gamma(\mathbf{x}) = \begin{cases} \sigma^2, & \mathbf{x} = \mathbf{0}, \\ 0, & \mathbf{x} \neq \mathbf{0}. \end{cases}$$

No parameters need be set for this value of **cov**.

cov = Nag_VgmSpherical
Spherical variogram

$$\gamma(\mathbf{x}) = \begin{cases} \sigma^2 \left(1 - 1.5x' + 0.5(x')^3\right), & x' < 1, \\ 0, & x' \geq 1, \end{cases}$$

where

$$\ell_1 = \mathbf{params}[0], \ell_1 > 0,$$

$$\ell_2 = \mathbf{params}[1], \ell_2 > 0.$$

cov = Nag_VgmBessel
Bessel variogram

$$\gamma(\mathbf{x}) = \sigma^2 \frac{2^\nu \Gamma(\nu + 1) J_\nu(x')}{(x')^\nu},$$

where

$J_\nu(\cdot)$ is the Bessel function of the first kind,

$$\ell_1 = \mathbf{params}[0], \ell_1 > 0,$$

$$\ell_2 = \mathbf{params}[1], \ell_2 > 0,$$

$$\nu = \mathbf{params}[2], \nu \geq 0.$$

cov = Nag_VgmHole
Hole effect variogram

$$\gamma(\mathbf{x}) = \sigma^2 \frac{\sin(x')}{x'},$$

where

$$\ell_1 = \mathbf{params}[0], \ell_1 > 0,$$

$$\ell_2 = \mathbf{params}[1], \ell_2 > 0.$$

cov = Nag_VgmWhittleMatern
Whittle-Matérn variogram

$$\gamma(\mathbf{x}) = \sigma^2 \frac{2^{1-\nu} (x')^\nu K_\nu(x')}{\Gamma(\nu)},$$

where

$K_\nu(\cdot)$ is the modified Bessel function of the second kind,

$$\ell_1 = \mathbf{params}[0], \ell_1 > 0,$$

$$\ell_2 = \mathbf{params}[1], \ell_2 > 0,$$

$$\nu = \mathbf{params}[2], \nu > 0.$$

cov = Nag_VgmContParam
Continuously parameterised variogram with compact support

$$\gamma(\mathbf{x}) = \begin{cases} \sigma^2 \frac{2^{2-\nu} (x')^\nu K_\nu(x')}{\Gamma(\nu)} (1 + 8x'' + 25(x'')^2 + 32(x'')^3) (1 - x'')^8, & x'' < 1, \\ 0, & x'' \geq 1, \end{cases}$$

where

$$x'' = \left\| \frac{x'}{\ell_1 s_1}, \frac{y'}{\ell_2 s_2} \right\|,$$

$K_\nu(\cdot)$ is the modified Bessel function of the second kind,

$$\ell_1 = \mathbf{params}[0], \ell_1 > 0,$$

$$\ell_2 = \mathbf{params}[1], \ell_2 > 0,$$

$$s_1 = \mathbf{params}[2], s_1 > 0,$$

$$s_2 = \mathbf{params}[3], s_2 > 0,$$

$$\nu = \mathbf{params}[4], \nu > 0.$$

cov = Nag_VgmGenHyp
Generalized hyperbolic distribution variogram

$$\gamma(\mathbf{x}) = \sigma^2 \frac{(\delta^2 + (x')^2)^{\frac{\lambda}{2}}}{\delta^\lambda K_\lambda(\kappa\delta)} K_\lambda\left(\kappa(\delta^2 + (x')^2)^{\frac{\lambda}{2}}\right),$$

where

$K_\lambda(\cdot)$ is the modified Bessel function of the second kind,

$$\ell_1 = \mathbf{params}[0], \ell_1 > 0,$$

$$\ell_2 = \mathbf{params}[1], \ell_2 > 0,$$

$$\lambda = \mathbf{params}[2], \text{no constraint on } \lambda,$$

$$\delta = \mathbf{params}[3], \delta > 0,$$

$$\kappa = \mathbf{params}[4], \kappa > 0.$$

Constraint: **cov** = Nag_VgmSymmStab, Nag_VgmCauchy, Nag_VgmDifferential, Nag_VgmExponential, Nag_VgmGauss, Nag_VgmNugget, Nag_VgmSpherical, Nag_VgmBessel, Nag_VgmHole, Nag_VgmWhittleMatern, Nag_VgmContParam or Nag_VgmGenHyp.

9: **norm** – Nag_NormType *Input*

On entry: determines which norm to use when calculating the variogram.

norm = Nag_OneNorm

The 1-norm is used, i.e., $\|x, y\| = |x| + |y|$.

norm = Nag_TwoNorm

The 2-norm (Euclidean norm) is used, i.e., $\|x, y\| = \sqrt{x^2 + y^2}$.

Suggested value: **norm** = Nag_TwoNorm.

Constraint: **norm** = Nag_OneNorm or Nag_TwoNorm.

10: **np** – Integer *Input*

On entry: the number of parameters to be set. Different covariance functions need a different number of parameters.

cov = Nag_VgmNugget

np must be set to 0.

cov = Nag_VgmDifferential, Nag_VgmExponential, Nag_VgmGauss, Nag_VgmSpherical or Nag_VgmHole

np must be set to 2.

cov = Nag_VgmSymmStab, Nag_VgmCauchy, Nag_VgmBessel or Nag_VgmWhittleMatern

np must be set to 3.

cov = Nag_VgmContParam or Nag_VgmGenHyp

np must be set to 5.

11: **params[np]** – const double *Input*

On entry: the parameters for the variogram as detailed in the description of **cov**.

Constraint: see **cov** for a description of the individual parameter constraints.

12: **pad** – Nag_EmbedPad *Input*

On entry: determines whether the embedding matrix is padded with zeros, or padded with values of the variogram. The choice of padding may affect how big the embedding matrix must be in order to be positive semidefinite.

pad = Nag_EmbedPadZeros

The embedding matrix is padded with zeros.

pad = Nag_EmbedPadValues

The embedding matrix is padded with values of the variogram.

Suggested value: **pad** = Nag_EmbedPadValues.

Constraint: **pad** = Nag_EmbedPadZeros or Nag_EmbedPadValues.

13: **corr** – Nag_EmbedScale *Input*

On entry: determines which approximation to implement if required, as described in Section 3.

Suggested value: **corr** = Nag_EmbedScaleTraces.

Constraint: **corr** = Nag_EmbedScaleTraces, Nag_EmbedScaleSqrtTraces or Nag_EmbedScaleOne.

- 14: **lam**[**maxm**[0] × **maxm**[1]] – double *Output*
On exit: contains the square roots of the eigenvalues of the embedding matrix.
- 15: **xx**[**ns**[0]] – double *Output*
On exit: the points of the x -coordinates at which values of the random field will be output.
- 16: **yy**[**ns**[1]] – double *Output*
On exit: the points of the y -coordinates at which values of the random field will be output.
- 17: **m**[2] – Integer *Output*
On exit: **m**[0] contains M_1 , the size of the circulant blocks and **m**[1] contains M_2 , the number of blocks, resulting in a final square matrix of size $M_1 \times M_2$.
- 18: **approx** – Integer * *Output*
On exit: indicates whether approximation was used.
approx = 0
 No approximation was used.
approx = 1
 Approximation was used.
- 19: **rho** – double * *Output*
On exit: indicates the scaling of the covariance matrix. **rho** = 1.0 unless approximation was used with **corr** = Nag_EmbedScaleTraces or Nag_EmbedScaleSqrtTraces.
- 20: **icount** – Integer * *Output*
On exit: indicates the number of negative eigenvalues in the embedding matrix which have had to be set to zero.
- 21: **eig**[3] – double *Output*
On exit: indicates information about the negative eigenvalues in the embedding matrix which have had to be set to zero. **eig**[0] contains the smallest eigenvalue, **eig**[1] contains the sum of the squares of the negative eigenvalues, and **eig**[2] contains the sum of the absolute values of the negative eigenvalues.
- 22: **fail** – NagError * *Input/Output*
 The NAG error argument (see Section 3.6 in the Essential Introduction).

6 Error Indicators and Warnings

NE_BAD_PARAM

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

NE_ENUM_INT

On entry, **np** = $\langle value \rangle$.
 Constraint: for **cov** = $\langle value \rangle$, **np** = $\langle value \rangle$.

NE_ENUM_REAL_1

On entry, **params**[$\langle value \rangle$] = $\langle value \rangle$.
 Constraint: dependent on **cov**, see documentation.

NE_INT_ARRAY

On entry, **maxm** = [*value*, *value*].

Constraint: the minimum calculated value for **maxm** are [*value*, *value*].

Where the minima of **maxm**[*i* - 1] is given by 2^k , where *k* is the smallest integer satisfying $2^k \geq 2(\mathbf{ns}[i - 1] - 1)$, for *i* = 1, 2.

On entry, **ns** = [*value*, *value*].

Constraint: **ns**[0] ≥ 1, **ns**[1] ≥ 1.

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 contact NAG for assistance.

NE_REAL

On entry, **var** = *value*.

Constraint: **var** ≥ 0.0.

NE_REAL_2

On entry, **xmin** = *value* and **xmax** = *value*.

Constraint: **xmin** < **xmax**.

On entry, **ymin** = *value* and **ymax** = *value*.

Constraint: **ymin** < **ymax**.

7 Accuracy

If on exit **approx** = 1, see the comments in Section 3 regarding the quality of approximation; increase the values in **maxm** to attempt to avoid approximation.

8 Parallelism and Performance

nag_rand_field_2d_predef_setup (g05zrc) is not threaded by NAG in any implementation.

nag_rand_field_2d_predef_setup (g05zrc) 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 Users' Note for your implementation for any additional implementation-specific information.

9 Further Comments

None.

10 Example

This example calls nag_rand_field_2d_predef_setup (g05zrc) to calculate the eigenvalues of the embedding matrix for 25 sample points on a 5 by 5 grid of a two-dimensional random field characterized by the symmetric stable variogram (**cov** = Nag_VgmSymmStab).

10.1 Program Text

```
/* nag_rand_field_2d_predef_setup (g05zrc) Example Program.
 *
 * Copyright 2013 Numerical Algorithms Group.
 *
 * Mark 24, 2013.
 */
#include <math.h>
```



```

#include <nag.h>
#include <nag_stdlib.h>
#include <nagg05.h>

static void display_results(Integer approx, Integer *m, double rho,
                           double *eig, Integer icount, double *lam);
static void read_input_data(Nag_Variogram *cov, Integer *np, double *params,
                            Nag_NormType *norm, double *var, double *xmin,
                            double *xmax, double *ymin, double *ymax,
                            Integer *ns, Integer *maxm, Nag_EmbedScale *corr,
                            Nag_EmbedPad *pad);

int main(void)
{
    /* Scalars */
    Integer      exit_status = 0;
    double       rho, var, xmax, xmin, ymax, ymin;
    Integer      approx, icount, np;
    /* Arrays */
    double       eig[3], params[5];
    double       *lam = 0, *xx = 0, *yy = 0;
    Integer      m[2], maxm[2], ns[2];
    /* Nag types */
    Nag_Variogram cov;
    Nag_NormType  norm;
    Nag_EmbedPad  pad;
    Nag_EmbedScale corr;
    NagError      fail;

    INIT_FAIL(fail);

    printf("nag_rand_field_2d_predef_setup (g05zrc) Example Program Results\n\n");
    /* Get problem specifications from data file*/
    read_input_data(&cov, &np, params, &norm, &var, &xmin, &xmax, &ymin, &ymax,
                   ns, maxm, &corr, &pad);
    if (!(lam = NAG_ALLOC(maxm[0]*maxm[1], double)) ||
        !(xx = NAG_ALLOC(ns[0], double)) ||
        !(yy = NAG_ALLOC(ns[1], double)))
    {
        printf("Allocation failure\n");
        exit_status = -1;
        goto END;
    }
    /* Get square roots of the eigenvalues of the embedding matrix. These are
     * obtained from the setup for simulating two-dimensional random fields,
     * with a predefined variogram, by the circulant embedding method using
     * nag_rand_field_2d_predef_setup (g05zrc).
     */
    nag_rand_field_2d_predef_setup(ns, xmin, xmax, ymin, ymax, maxm, var, cov,
                                   norm, np, params, pad, corr, lam, xx, yy, m,
                                   &approx, &rho, &icount, eig, &fail);
    if (fail.code != NE_NOERROR) {
        printf("Error from nag_rand_field_2d_predef_setup (g05zrc).\n%s\n",
              fail.message);
        exit_status = 1;
        goto END;
    }
    /* Output results*/
    display_results(approx, m, rho, eig, icount, lam);
END:
    NAG_FREE(lam);
    NAG_FREE(xx);
    NAG_FREE(yy);
    return exit_status;
}

void read_input_data(Nag_Variogram *cov, Integer *np, double *params,
                    Nag_NormType *norm, double *var, double *xmin,
                    double *xmax, double *ymin, double *ymax,
                    Integer *ns, Integer *maxm, Nag_EmbedScale *corr,
                    Nag_EmbedPad *pad)

```

```

{
  Integer j;
  char   nag_enum_arg[40];

  /* Read in covariance function name and convert to value using
   * nag_enum_name_to_value (x04nac).
   */
  scanf("%s%[\n] %39s%[\n]", nag_enum_arg);
  *cov = (Nag_Variogram) nag_enum_name_to_value(nag_enum_arg);
  /* Read in parameters */
  scanf("%ld%[\n]", np);
  for (j=0; j<*np; j++)
    scanf("%lf", &params[j]);
  scanf("%s%[\n]");
  /* Read choice of norm to use, and convert name to value. */
  scanf(" %39s%[\n]", nag_enum_arg);
  *norm = (Nag_NormType) nag_enum_name_to_value(nag_enum_arg);
  /* Read in variance of random field*/
  scanf("%lf%[\n]", var);
  /* Read in domain endpoints*/
  scanf("%lf %lf%[\n]", xmin, xmax);
  scanf("%lf %lf%[\n]", ymin, ymax);
  /* Read in number of sample points in each direction*/
  scanf("%ld %ld%[\n]", &ns[0], &ns[1]);
  /* Read in maximum size of embedding matrix*/
  scanf("%ld %ld%[\n]", &maxm[0], &maxm[1]);
  /* Read name of scaling in case of approximation and convert to value. */
  scanf(" %39s%[\n]", nag_enum_arg);
  *corr = (Nag_EmbedScale) nag_enum_name_to_value(nag_enum_arg);
  /* Read in choice of padding and convert name to value. */
  scanf(" %39s%[\n]", nag_enum_arg);
  *pad = (Nag_EmbedPad) nag_enum_name_to_value(nag_enum_arg);
}

void display_results(Integer approx, Integer *m, double rho, double *eig,
                    Integer icount, double *lam)
{
  /* Scalars */
  Integer i, j;

  /* Display size of embedding matrix*/
  printf("\nSize of embedding matrix = %ld\n\n", m[0]*m[1]);
  /* Display approximation information if approximation used. */
  if (approx==1) {
    printf("Approximation required\n\n");
    printf("rho = %10.5f\n", rho);
    printf("eig = ");
    for (j=0; j<3; j++)
      printf("%10.5f", eig[j]);
    printf("\nicount = %ld\n", icount);
  } else {
    printf("Approximation not required\n");
  }
  /* Display square roots of the eigenvalues of the embedding matrix. */
  printf("\nSquare roots of eigenvalues of embedding matrix:\n\n");
  for ( i=0; i<m[0]; i++) {
    for (j=0; j<m[1]; j++) {
      printf("%8.4f", lam[i+j*m[0]]);
    }
    printf("\n");
  }
}

```

10.2 Program Data

```
nag_rand_field_2d_predef_setup (g05zrc) Example Program Data
Nag_VgmSymmStab      : cov
  3                   : np (3 parameters for 2D Nag_VgmSymmStab)
  0.1  0.15  1.2     : params (c1, c2 and nu)
Nag_TwoNorm          : norm
  0.5                 : var
-1.0  1.0           : xmin, xmax
-0.5  0.5           : ymin, ymax
  5  5               : ns
  64  64             : maxm
Nag_EmbedScaleOne    : corr
Nag_EmbedPadValues   : pad
```

10.3 Program Results

```
nag_rand_field_2d_predef_setup (g05zrc) Example Program Results
```

Size of embedding matrix = 64

Approximation not required

Square roots of eigenvalues of embedding matrix:

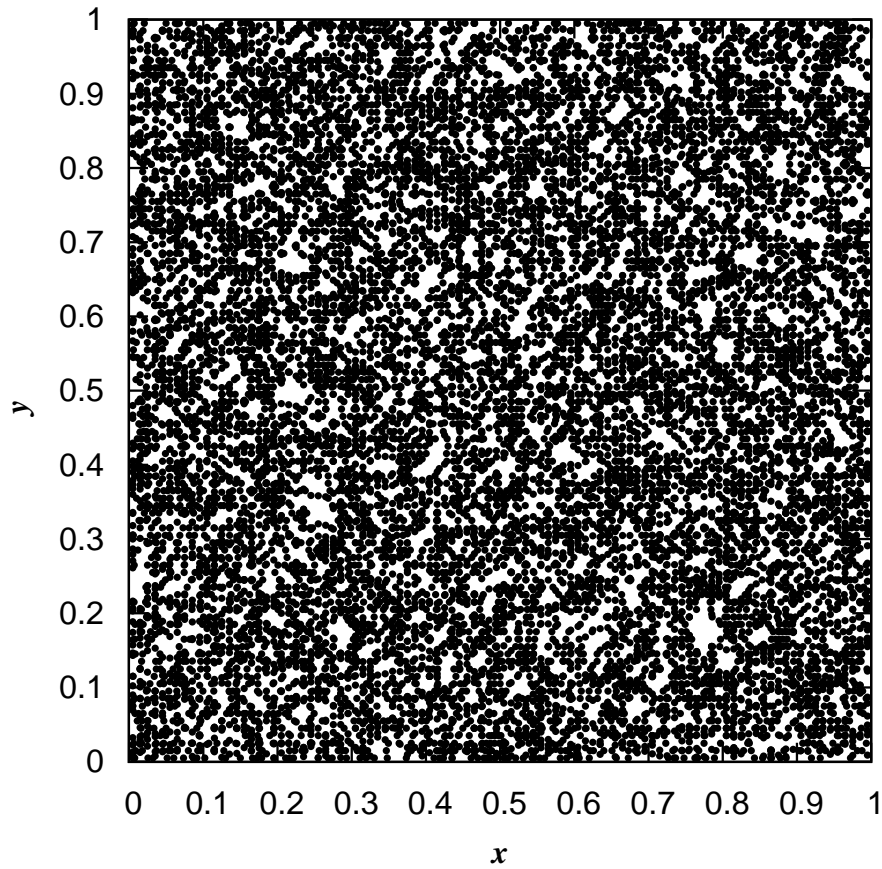
```
0.8966  0.8234  0.6810  0.5757  0.5391  0.5757  0.6810  0.8234
0.8940  0.8217  0.6804  0.5756  0.5391  0.5756  0.6804  0.8217
0.8877  0.8175  0.6792  0.5754  0.5391  0.5754  0.6792  0.8175
0.8813  0.8133  0.6780  0.5751  0.5390  0.5751  0.6780  0.8133
0.8787  0.8116  0.6774  0.5750  0.5390  0.5750  0.6774  0.8116
0.8813  0.8133  0.6780  0.5751  0.5390  0.5751  0.6780  0.8133
0.8877  0.8175  0.6792  0.5754  0.5391  0.5754  0.6792  0.8175
0.8940  0.8217  0.6804  0.5756  0.5391  0.5756  0.6804  0.8217
```

The two plots shown below illustrate the random fields that can be generated by `nag_rand_field_2d_generate` (g05zsc) using the eigenvalues calculated by `nag_rand_field_2d_predef_setup` (g05zrc). These are for two realizations of a two-dimensional random field, based on eigenvalues of the embedding matrix for points on a 100 by 100 grid. The random field is

characterized by the exponential variogram (`cov = Nag_VgmExponential`) with correlation lengths both equal to 0.1.

Example Program 1

First realization of two-dimensional Random Field
exponential variogram, correlation lengths = 0.1



Example Program 2
Second realization of two-dimensional Random Field
exponential variogram, correlation lengths = 0.1

