

# NAG Library Function Document

## nag\_2d\_spline\_fit\_ts\_scatter (e02jdc)

**Note:** this function uses **optional arguments** to define choices in the problem specification and in the details of the algorithm. If you wish to use default settings for all of the optional arguments, you need only read Sections 1 to 10 of this document. If, however, you wish to reset some or all of the settings please refer to Section 11 for a detailed description of the specification of the optional arguments produced by the function.

### 1 Purpose

nag\_2d\_spline\_fit\_ts\_scatter (e02jdc) computes a spline approximation to a set of scattered data using a two-stage approximation method.

The computational complexity of the method grows linearly with the number of data points; hence large datasets are easily accommodated.

### 2 Specification

```
#include <nag.h>
#include <nage02.h>

void nag_2d_spline_fit_ts_scatter (Integer n, const double x[],
    const double y[], const double f[], Integer lsminp, Integer lsmaxp,
    Integer nxcels, Integer nycels, Integer lcoefs, double coefs[],
    Integer iopts[], double opts[], NagError *fail)
```

Before calling nag\_2d\_spline\_fit\_ts\_scatter (e02jdc), nag\_fit\_opt\_set (e02zkc) must be called with **optstr** set to "Initialize = e02jdc". Settings for optional algorithmic arguments may be specified by calling nag\_fit\_opt\_set (e02zkc) before a call to nag\_2d\_spline\_fit\_ts\_scatter (e02jdc).

### 3 Description

nag\_2d\_spline\_fit\_ts\_scatter (e02jdc) determines a smooth bivariate spline approximation to a set of data points  $(x_i, y_i, f_i)$ , for  $i = 1, 2, \dots, n$ . Here, ‘smooth’ means  $C^1$  or  $C^2$ . (You may select the degree of smoothing using the optional argument **Global Smoothing Level**.)

The approximation domain is the bounding box  $[x_{\min}, x_{\max}] \times [y_{\min}, y_{\max}]$ , where  $x_{\min}$  (respectively  $y_{\min}$ ) and  $x_{\max}$  (respectively  $y_{\max}$ ) denote the lowest and highest data values of the  $(x_i)$  (respectively  $(y_i)$ ).

The spline is computed by local approximations on a uniform triangulation of the bounding box. These approximations are extended to a smooth spline representation of the surface over the domain. The local approximation scheme is controlled by the optional argument **Local Method**. The schemes provided are: by least squares polynomial approximation (Davydov and Zeilfelder (2004)); by hybrid polynomial and radial basis function (RBF) approximation (Davydov *et al.* (2006)); or by pure RBF approximation (Davydov *et al.* (2005)).

The two-stage approximation method employed by nag\_2d\_spline\_fit\_ts\_scatter (e02jdc) is derived from the TSFIT package of O. Davydov and F. Zeilfelder.

Values of the computed spline can subsequently be computed by calling nag\_2d\_spline\_ts\_eval (e02jec) or nag\_2d\_spline\_ts\_eval\_rect (e02jfc).

## 4 References

Davydov O, Morandi R and Sestini A (2006) Local hybrid approximation for scattered data fitting with bivariate splines *Comput. Aided Geom. Design* **23** 703–721

Davydov O, Sestini A and Morandi R (2005) Local RBF approximation for scattered data fitting with bivariate splines *Trends and Applications in Constructive Approximation* M. G. de Bruin, D. H. Mache, and J. Szabados, Eds **ISNM Vol. 151** Birkhauser 91–102

Davydov O and Zeilfelder F (2004) Scattered data fitting by direct extension of local polynomials to bivariate splines *Advances in Comp. Math.* **21** 223–271

## 5 Arguments

1: **n** – Integer *Input*

*On entry:*  $n$ , the number of data values to be fitted.

*Constraint:*  $n > 1$ .

2: **x[n]** – const double *Input*

3: **y[n]** – const double *Input*

4: **f[n]** – const double *Input*

*On entry:* the  $(x_i, y_i, f_i)$  data values to be fitted.

*Constraint:*  $\mathbf{x}[j - 1] \neq \mathbf{x}[0]$  for some  $j = 2, \dots, n$  and  $\mathbf{y}[k - 1] \neq \mathbf{y}[0]$  for some  $k = 2, \dots, n$ ; i.e., there are at least two distinct  $x$  and  $y$  values.

5: **lsminp** – Integer *Input*

6: **lsmaxp** – Integer *Input*

*On entry:* are control parameters for the local approximations.

Each local approximation is computed on a local domain containing one of the triangles in the discretization of the bounding box. The size of each local domain will be adaptively chosen such that if it contains fewer than **lsminp** sample points it is expanded, else if it contains greater than **lsmaxp** sample points a thinning method is applied. **lsmaxp** mainly controls computational cost (in that working with a thinned set of points is cheaper and may be appropriate if the input data is densely distributed), while **lsminp** allows handling of different types of scattered data.

Setting **lsmaxp** < **lsminp**, and therefore forcing either expansion or thinning, may be useful for computing initial coarse approximations. In general smaller values for these arguments reduces cost.

A calibration procedure (experimenting with a small subset of the data to be fitted and validating the results) may be needed to choose the most appropriate values for **lsminp** and **lsmaxp**.

*Constraints:*

$$1 \leq \mathbf{lsminp} \leq \mathbf{n};$$

$$\mathbf{lsmaxp} \geq 1.$$

7: **nxcells** – Integer *Input*

8: **nycells** – Integer *Input*

*On entry:* **nxcells** (respectively **nycells**) is the number of cells in the  $x$  (respectively  $y$ ) direction that will be used to create the triangulation of the bounding box of the domain of the function to be fitted.

Greater efficiency generally comes when **nxcells** and **nycells** are chosen to be of the same order of magnitude and are such that  $\mathbf{n}$  is  $O(\mathbf{nxcells} \times \mathbf{nycells})$ . Thus for a ‘square’ triangulation — when **nxcells** = **nycells** — the quantities  $\sqrt{\mathbf{n}}$  and **nxcells** should be of the same order of magnitude. See also Section 9.

Constraints:

$$\begin{aligned} \mathbf{nxcells} &\geq 1; \\ \mathbf{nycells} &\geq 1. \end{aligned}$$

9: **lcoefs** – Integer

*Input*

10: **coefs**[**lcoefs**] – double

*Output*

On exit: if **fail.code** = NE\_NOERROR on exit, **coefs** contains the computed spline coefficients.

Constraints:

$$\begin{aligned} &\text{if Global Smoothing Level} = 1, \\ &\mathbf{lcoefs} \geq (((\mathbf{nxcells} + 2) \times (\mathbf{nycells} + 2) + 1)/2) \times 10 + 1; \\ &\text{if Global Smoothing Level} = 2, \\ &\mathbf{lcoefs} \geq 28 \times (\mathbf{nxcells} + 2) \times (\mathbf{nycells} + 2) \times 4 + 1. \end{aligned}$$

11: **iopts**[*dim*] – Integer

*Communication Array*

**Note:** the dimension, *dim*, of this array is dictated by the requirements of associated functions that must have been previously called. This array MUST be the same array passed as argument **iopts** in the previous call to nag\_fit\_opt\_set (e02zkc).

On entry: the contents of **iopts** MUST NOT be modified in any way either directly or indirectly, by further calls to nag\_fit\_opt\_set (e02zkc), before calling either or both of the evaluation routines nag\_2d\_spline\_ts\_eval (e02jec) and nag\_2d\_spline\_ts\_eval\_rect (e02jfc).

12: **opts**[*dim*] – double

*Communication Array*

**Note:** the dimension, *dim*, of this array is dictated by the requirements of associated functions that must have been previously called. This array MUST be the same array passed as argument **opts** in the previous call to nag\_fit\_opt\_set (e02zkc).

On entry: the contents of **opts** MUST NOT be modified in any way either directly or indirectly, by further calls to nag\_fit\_opt\_set (e02zkc), before calling either or both of the evaluation routines nag\_2d\_spline\_ts\_eval (e02jec) and nag\_2d\_spline\_ts\_eval\_rect (e02jfc).

13: **fail** – NagError \*

*Input/Output*

The NAG error argument (see Section 3.6 in the Essential Introduction).

## 6 Error Indicators and Warnings

### NE\_ALL\_ELEMENTS\_EQUAL

On entry, all elements of **x** or of **y** are equal.

### NE\_ALLOC\_FAIL

Dynamic memory allocation failed.

See Section 3.2.1.2 in the Essential Introduction for further information.

### NE\_BAD\_PARAM

On entry, argument *<value>* had an illegal value.

### NE\_INITIALIZATION

Option arrays are not initialized or are corrupted.

**NE\_INT**

On entry, **lcoefs** =  $\langle value \rangle$ .

Constraint:

if **Global Smoothing Level** = 1,

**lcoefs**  $\geq (((\mathbf{nxcells} + 2) \times (\mathbf{nycels} + 2) + 1)/2) \times 10 + 1$ ;

if **Global Smoothing Level** = 2,

**lcoefs**  $\geq 28 \times (\mathbf{nxcells} + 2) \times (\mathbf{nycels} + 2) \times 4 + 1$ .

On entry, **lsmaxp** =  $\langle value \rangle$ .

Constraint: **lsmaxp**  $\geq 1$ .

On entry, **n** =  $\langle value \rangle$ .

Constraint: **n**  $> 1$ .

On entry, **nxcells** =  $\langle value \rangle$ .

Constraint: **nxcells**  $\geq 1$ .

On entry, **nycels** =  $\langle value \rangle$ .

Constraint: **nycels**  $\geq 1$ .

**NE\_INT\_2**

On entry, **lsminp** =  $\langle value \rangle$  and **n** =  $\langle value \rangle$ .

Constraint:  $1 \leq \mathbf{lsminp} \leq \mathbf{n}$ .

**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.

An unexpected algorithmic failure was encountered. Please contact NAG.

An unexpected error has been triggered by this function. Please contact NAG.  
See Section 3.6.6 in the Essential Introduction for further information.

**NE\_INVALID\_OPTION**

The selected radial basis function cannot be used with the RBF local method.

The value of optional argument **Polynomial Starting Degree** was invalid.

**NE\_NO\_LICENCE**

Your licence key may have expired or may not have been installed correctly.  
See Section 3.6.5 in the Essential Introduction for further information.

**7 Accuracy**

Technical results on error bounds can be found in Davydov and Zeilfelder (2004), Davydov *et al.* (2006) and Davydov *et al.* (2005).

Local approximation by polynomials of degree  $d$  for  $n$  data points has optimal approximation order  $n^{-(d+1)/2}$ . The improved approximation power of hybrid polynomial/RBF and of pure RBF approximations is shown in Davydov *et al.* (2006) and Davydov *et al.* (2005).

The approximation error for  $C^1$  global smoothing is  $O(n^{-2})$ . For  $C^2$  smoothing the error is  $O(n^{-7/2})$  when **Supersmooth C2** = YES and  $O(n^{-3})$  when **Supersmooth C2** = NO.

Whether maximal accuracy is achieved depends on the distribution of the input data and the choices of the algorithmic parameters. The references above contain extensive numerical tests and further technical discussions of how best to configure the method.

## 8 Parallelism and Performance

`nag_2d_spline_fit_ts_sc` (e02jdc) is threaded by NAG for parallel execution in multithreaded implementations of the NAG Library.

`nag_2d_spline_fit_ts_sc` (e02jdc) 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 function. Please also consult the Users' Note for your implementation for any additional implementation-specific information.

## 9 Further Comments

$n$ -linear complexity and memory usage can be attained for sufficiently dense input data if the triangulation parameters **nxcells** and **nycells** are chosen as recommended in their descriptions above. For sparse input data on such triangulations, if many expansion steps are required (see **lsminp**) the complexity may rise to be loglinear.

Parts of the pure RBF method used when **Local Method** = RBF have  $n$ -quadratic memory usage.

Note that if **Local Method** = HYBRID and an initial hybrid approximation is deemed unreliable (see the description of optional argument **Minimum Singular Value LHA**), a pure polynomial approximation will be used instead on that local domain.

## 10 Example

The Franke function

$$f(x, y) = 0.75 \exp\left(-\left((9x - 2)^2 + (9y - 2)^2\right)/4\right) + \\ 0.75 \exp\left(-\left((9x + 1)^2/49 - (9y + 1)/10\right)\right) + \\ 0.5 \exp\left(-\left((9x - 7)^2 + (9y - 3)^2\right)/4\right) - \\ 0.2 \exp\left(-\left((9x - 4)^2 - (9y - 7)^2\right)\right)$$

is widely used for testing surface-fitting methods. The example program randomly generates a number of points on this surface. From these a spline is computed and then evaluated at a vector of points and on a mesh.

### 10.1 Program Text

```
/* nag_2d_spline_fit_ts_sc (e02jdc) Example Program.
 *
 * Copyright 2014 Numerical Algorithms Group.
 *
 * Mark 25, 2014.
 */

#include <math.h>
#include <nag.h>
#include <nag_stdlib.h>
#include <nage02.h>
#include <nagf16.h>
#include <nagg05.h>
#include <string.h>

static Integer generate_data(Integer *n, double **x, double **y, double **f,
                             Integer *lsminp, Integer *lsmaxp,
                             Integer *nxcells, Integer *nycells,
                             Integer *lcoefs, double **coefs,
                             Integer *gsmoothness);
static Integer handle_options(Integer gsmoothness,
                             Integer iopts[], const Integer liopts,
```

```

                                double opts[], const Integer lopts);
static Integer evaluate_at_vector(const double coefs[], const Integer iopts[],
                                const double opts[], const double pmin[],
                                const double pmax[]);
static Integer evaluate_on_mesh(const double coefs[], const Integer iopts[],
                                const double opts[], const double pmin[],
                                const double pmax[]);

int main(void)
{
    /* Scalars */
    Integer    exit_status = 0;
    Integer    liopts = 100, lopts = 100;
    Integer    gsmoothness, i, lcoefs, lsmaxp, lsminp, n, nxcels, nycels;
    /* Arrays */
    double     *coefs = 0, *f = 0, *x = 0, *y = 0;
    double     opts[100], pmax[2], pmin[2];
    Integer    iopts[100];
    /* Nag Types */
    NagError   fail;

    INIT_FAIL(fail);

    printf("nag_2d_spline_fit_ts_scad (e02jdc) Example Program Results\n");

    /* Generate the data to fit and set the compulsory algorithmic control
     * parameters.
     */
    exit_status = generate_data(&n, &x, &y, &f, &lsminp, &lsmaxp, &nxcels,
                              &nycels, &lcoefs, &coefs, &gsmoothness);
    if (exit_status != 0) goto END;

    /* Initialize the options arrays and set/get some options. */
    exit_status = handle_options(gsmoothness, iopts, liopts, opts, lopts);
    if (exit_status != 0) goto END;

    /* Compute the spline coefficients using
     * nag_2d_spline_fit_ts_scad (e02jdc).
     */
    nag_2d_spline_fit_ts_scad(n, x, y, f, lsminp, lsmaxp, nxcels, nycels, lcoefs,
                              coefs, iopts, opts, &fail);
    if (fail.code != NE_NOERROR) {
        printf("Error from nag_2d_spline_fit_ts_scad (e02jdc).\n%s\n",
              fail.message);
        exit_status = 1;
        goto END;
    }

    /* pmin and pmax form the bounding box of the spline. We must not attempt to
     * evaluate the spline outside this box. Use nag_dmin_val (f16jpc) and
     * nag_dmax_val (f16jnc) to obtain the min and max values.
     */
    nag_dmin_val(n, x, 1, &i, &pmin[0], &fail);
    if (fail.code != NE_NOERROR) {
        printf("Error from nag_dmin_val (f16jpc).\n%s\n",
              fail.message);
        exit_status = 1;
        goto END;
    }
    nag_dmin_val(n, y, 1, &i, &pmin[1], &fail);
    if (fail.code != NE_NOERROR) {
        printf("Error from nag_dmin_val (f16jpc).\n%s\n",
              fail.message);
        exit_status = 1;
        goto END;
    }
    nag_dmax_val(n, x, 1, &i, &pmax[0], &fail);
    if (fail.code != NE_NOERROR) {
        printf("Error from nag_dmax_val (f16jnc).\n%s\n",
              fail.message);
        exit_status = 1;
    }
}

```

```

    goto END;
}
nag_dmax_val(n, y, 1, &i, &pmax[1], &fail);
if (fail.code != NE_NOERROR) {
    printf("Error from nag_dmax_val (f16jnc).\n%s\n",
        fail.message);
    exit_status = 1;
    goto END;
}

NAG_FREE(x);
NAG_FREE(y);
NAG_FREE(f);

/* Evaluate the approximation at a vector of values. */
exit_status = evaluate_at_vector(coefs, iopts, opts, pmin, pmax);
if (exit_status != 0) goto END;

/* Evaluate the approximation on a mesh. */
exit_status = evaluate_on_mesh(coefs, iopts, opts, pmin, pmax);
if (exit_status != 0) goto END;

END:

NAG_FREE(coefs);

return exit_status;
}
static Integer generate_data(Integer *n, double **x, double **y, double **f,
                            Integer *lsminp, Integer *lsmaxp, Integer *nxcels,
                            Integer *nycels, Integer *lcoefs, double **coefs,
                            Integer *gsmoothness)
{
    /* Reads n from a data file and then generates an x and a y vector of n
     * pseudorandom uniformly distributed values on (0,1]. These are passed
     * to the bivariate function of R. Franke to create the data set to fit.
     * The remaining input data for the fitter are set to suitable values for
     * this problem, as discussed by Davydov and Zeilfelder.
     * Reads the global smoothing level from a data file. This value determines
     * the minimum required length of the array of spline coefficients, coefs.
     */

    /* Scalars */
    Integer exit_status = 0;
    Integer lseed = 4, mstate = 21;
    Integer i, lstate, subid;
    /* Arrays */
    Integer seed[4], state[21];
    /* Nag Types */
    NagError fail;

    INIT_FAIL(fail);

    /* Read the size of the data set to be generated and fitted.
     * (Skip the heading in the data file.)
     */
#ifdef _WIN32
    scanf_s("%m[^\\n]");
#else
    scanf("%m[^\\n]");
#endif
#ifdef _WIN32
    scanf_s("%NAG_IFMT"%m[^\\n] ", n);
#else
    scanf("%NAG_IFMT"%m[^\\n] ", n);
#endif

    if (!( *x = NAG_ALLOC(*n, double) ) ||
        !( *y = NAG_ALLOC(*n, double) ) ||
        !( *f = NAG_ALLOC(*n, double) ) )
    {

```

```

    printf("Allocation failure\n");
    exit_status = -1;
    goto END;
}

/* Initialize the random number generator using
 * nag_rand_init_repeatable (g05kfc) and then generate the data using
 * nag_rand_basic (g05sac).
 */
subid = 53;
seed[0] = 32958;
seed[1] = 39838;
seed[2] = 881818;
seed[3] = 45812;
lstate = mstate;
nag_rand_init_repeatable(Nag_WichmannHill_I, subid, seed, lseed, state,
                        &lstate, &fail);
if (fail.code != NE_NOERROR) {
    printf("Error from nag_rand_init_repeatable (g05kfc).\n%s\n",
          fail.message);
    exit_status = 1;
    goto END;
}
nag_rand_basic(*n, state, *x, &fail);
if (fail.code != NE_NOERROR) {
    printf("Error from nag_rand_basic (g05sac).\n%s\n",
          fail.message);
    exit_status = 1;
    goto END;
}
nag_rand_basic(*n, state, *y, &fail);
if (fail.code != NE_NOERROR) {
    printf("Error from nag_rand_basic (g05sac).\n%s\n",
          fail.message);
    exit_status = 1;
    goto END;
}

for (i = 0; i < *n; i++)
    (*f)[i] = 0.75*
        exp(-(pow((9.*(x)[i]-2.), 2)+pow((9.*(y)[i]-2.), 2))/4.) +
        0.75*exp(-pow((9.*(x)[i]+1.), 2)/49.-(9.*(y)[i]+1.)/10.) +
        0.5*exp(-pow((9.*(x)[i]-7.), 2)+pow((9.*(y)[i]-3.), 2))/4.) -
        0.2*exp(-pow((9.*(x)[i]-4.), 2)-pow((9.*(y)[i]-7.), 2));

/* Set the grid size for the approximation. */
*nxcells = 6;
*nycells = *nxcells;

/* Read the required level of global smoothing. */
#ifdef _WIN32
    scanf_s("%NAG_IFMT"%"^\n] ", gsmoothness);
#else
    scanf("%NAG_IFMT"%"^\n] ", gsmoothness);
#endif

/* Identify the computation. */
printf("\n"
      "Computing the coefficients of a C^%"NAG_IFMT" spline approximation to"
      " Franke's function\n"
      "Using a %"NAG_IFMT" by %"NAG_IFMT" grid\n",
      *gsmoothness, *nxcells, *nycells);

/* Set the local-approximation control parameters. */
*lsminp = 3;
*lsmaxp = 100;

/* Set up the array to hold the computed spline coefficients. */
switch (*gsmoothness)
{
    case 1:

```



```

        *lcoefs = (((*nxcels+2)*(*nycels+2)+1)/2)*10 + 1;
        break;
    case 2:
        *lcoefs = 28*(*nxcels+2)*(*nycels+2)*4 + 1;
        break;
    default:
        *lcoefs = 0;
    }

    if (!(*coefs = NAG_ALLOC(*lcoefs, double)))
    {
        printf("Allocation failure\n");
        exit_status = -1;
        goto END;
    }

    END:

    return exit_status;
}
static Integer handle_options(Integer gsmoothness,
                              Integer iopts[], const Integer liopts,
                              double opts[], const Integer lopts)
{
    /* Auxiliary routine for initializing the options arrays and
     * for demonstrating how to set and get optional parameters using
     * nag_fit_opt_set (e02zkc) and nag_fit_opt_get (e02zlc) respectively.
     */

    /* Scalars */
    Integer      exit_status = 0;
    double       rvalue;
    Integer      ivalue, lcvalue;
    /* Arrays */
    char         cvalue[16+1], optstr[80+1], supersmooth[9+1];
    /* Nag Types */
    Nag_Boolean  supersmooth_enum;
    Nag_VariableType optype;
    NagError     fail;

    INIT_FAIL(fail);

    nag_fit_opt_set("Initialize = e02jdc", iopts, liopts, opts, lopts, &fail);
    if (fail.code != NE_NOERROR) {
        printf("Error from nag_fit_opt_set (e02zkc).\n%s\n",
              fail.message);
        exit_status = 1;
        goto END;
    }

    /* Configure the global approximation method. */
#ifdef _WIN32
    sprintf_s(optstr, _countof(optstr), "Global Smoothing Level = %1"NAG_IFMT,
              gsmoothness);
#else
    sprintf(optstr, "Global Smoothing Level = %1"NAG_IFMT,
            gsmoothness);
#endif
    nag_fit_opt_set(optstr, iopts, liopts, opts, lopts, &fail);
    if (fail.code != NE_NOERROR) {
        printf("Error from nag_fit_opt_set (e02zkc).\n%s\n",
              fail.message);
        exit_status = 1;
        goto END;
    }

    /* If C^2 smoothing is requested, compute the spline using additional
     * super-smoothness constraints?
     * (The default is 'No'.)
     */
#ifdef _WIN32

```

```

    scanf_s("%9s%*[\n] ", supersmooth, _countof(supersmooth));
#else
    scanf("%9s%*[\n] ", supersmooth);
#endif
    supersmooth_enum = (Nag_Boolean) nag_enum_name_to_value(supersmooth);
    if (gsmoothness == 2 && supersmooth_enum == Nag_TRUE)
    {
        nag_fit_opt_set("Supersmooth C2 = Yes", iopts, liopts, opts, lopts,
            &fail);
        if (fail.code != NE_NOERROR) {
            printf("Error from nag_fit_opt_set (e02zkc).\n%s\n",
                fail.message);
            exit_status = 1;
            goto END;
        }
    }

    nag_fit_opt_set("Averaged Spline = Yes", iopts, liopts, opts, lopts, &fail);
    if (fail.code != NE_NOERROR) {
        printf("Error from nag_fit_opt_set (e02zkc).\n%s\n",
            fail.message);
        exit_status = 1;
        goto END;
    }

    /* Configure the local approximation method.
     * (The default is 'Polynomial'.)
     */
    nag_fit_opt_set("Local Method = Polynomial", iopts, liopts, opts, lopts,
        &fail);
    if (fail.code != NE_NOERROR) {
        printf("Error from nag_fit_opt_set (e02zkc).\n%s\n",
            fail.message);
        exit_status = 1;
        goto END;
    }

#ifdef _WIN32
    sprintf_s(optstr, _countof(optstr), "Minimum Singular Value LPA = %16.9e",
        1./32.);
#else
    sprintf(optstr, "Minimum Singular Value LPA = %16.9e",
        1./32.);
#endif
    nag_fit_opt_set(optstr, iopts, liopts, opts, lopts, &fail);
    if (fail.code != NE_NOERROR) {
        printf("Error from nag_fit_opt_set (e02zkc).\n%s\n",
            fail.message);
        exit_status = 1;
        goto END;
    }

    switch (gsmoothness)
    {
        case 1:
#ifdef _WIN32
            sprintf_s(optstr, _countof(optstr), "Polynomial Starting Degree = 3");
#else
            sprintf(optstr, "Polynomial Starting Degree = 3");
#endif
            break;
        case 2:
            if (supersmooth_enum == Nag_TRUE)
            {
                /* We can benefit from starting with local polynomials of greater
                 * degree than with regular C^2 smoothing.
                 */
                printf("Using super-smoothing\n");
#ifdef _WIN32
                sprintf_s(optstr, _countof(optstr), "Polynomial Starting Degree = 6");
#else
                sprintf_s(optstr, _countof(optstr), "Polynomial Starting Degree = 6");
#endif
            }
    }

```

```

        sprintf(optstr, "Polynomial Starting Degree = 6");
#endif
    }
    else
#ifdef _WIN32
        sprintf_s(optstr, _countof(optstr), "Polynomial Starting Degree = 5");
#else
        sprintf(optstr, "Polynomial Starting Degree = 5");
#endif
    break;
}
nag_fit_opt_set(optstr, iopts, liopts, opts, lopts, &fail);
if (fail.code != NE_NOERROR) {
    printf("Error from nag_fit_opt_set (e02zkc).\n%s\n",
        fail.message);
    exit_status = 1;
    goto END;
}

/* As an example of how to get the value of an optional parameter,
 * display whether averaging of local approximations is in operation.
 */
lcvalue = 16 + 1;
nag_fit_opt_get("Averaged Spline", &ival, &rval, cvalue, lcvalue,
    &optype, iopts, opts, &fail);
if (fail.code != NE_NOERROR) {
    printf("Error from nag_fit_opt_get (e02zlc).\n%s\n", fail.message);
    exit_status = 1;
    goto END;
}

if (!strcmp(cvalue, "YES"))
    printf("Using an averaged local approximation\n");

END:
return exit_status;
}

static Integer evaluate_at_vector(const double coefs[], const Integer iopts[],
    const double opts[], const double pmin[],
    const double pmax[])
{
    /* Evaluates the approximation at a vector of values using
     * nag_2d_spline_ts_eval (e02jec).
     */

    /* Scalars */
    Integer exit_status = 0;
    Integer i, nevalv;
    /* Arrays */
    double *fevalv = 0, *xevalv = 0, *yevalv = 0;
    /* Nag Types */
    NagError fail;

    INIT_FAIL(fail);

#ifdef _WIN32
    scanf_s("%"NAG_IFMT"%*[\n] ", &nevalv);
#else
    scanf("%"NAG_IFMT"%*[\n] ", &nevalv);
#endif

    if (!(xevalv = NAG_ALLOC(nevalv, double)) ||
        !(yevalv = NAG_ALLOC(nevalv, double)) ||
        !(fevalv = NAG_ALLOC(nevalv, double)))
    {
        printf("Allocation failure\n");
        exit_status = -1;
        goto END;
    }
}

```

```

#ifdef _WIN32
    for (i = 0; i < nevalv; i++) scanf_s("%lf%lf%*[\n]", &xevalv[i], &yevalv[i]);
#else
    for (i = 0; i < nevalv; i++) scanf("%lf%lf%*[\n]", &xevalv[i], &yevalv[i]);
#endif

/* Force the points to be within the bounding box of the spline. */
for (i = 0; i < nevalv; i++)
{
    xevalv[i] = MAX(xevalv[i], pmin[0]);
    xevalv[i] = MIN(xevalv[i], pmax[0]);
    yevalv[i] = MAX(yevalv[i], pmin[1]);
    yevalv[i] = MIN(yevalv[i], pmax[1]);
}

/* Evaluate using nag_2d_spline_ts_eval (e02jec). */
nag_2d_spline_ts_eval(nevalv, xevalv, yevalv, coefs, fevalv, iopts, opts,
                    &fail);
if (fail.code != NE_NOERROR) {
    printf("Error from nag_2d_spline_ts_eval (e02jec).\n%s\n", fail.message);
    exit_status = 1;
    goto END;
}

printf("\nValues of computed spline at (x_i,y_i):\n"
       "\n%12s%12s%12s\n", "x_i", "y_i", "f(x_i,y_i)");

for (i = 0; i < nevalv; i++)
    printf("%12.2f%12.2f%12.2f\n", xevalv[i], yevalv[i], fevalv[i]);

END:

NAG_FREE(xevalv);
NAG_FREE(yevalv);
NAG_FREE(fevalv);

return exit_status;
}

static Integer evaluate_on_mesh(const double coefs[], const Integer iopts[],
                              const double opts[], const double pmin[],
                              const double pmax[])
{
    /* Evaluates the approximation on a mesh of n_x * n_y values. */

    /* Scalars */
    Integer exit_status = 0;
    Integer i, j, nxeval, nyeval;
    /* Arrays */
    char print_mesh[9+1];
    double *fevalm = 0, *xevalm = 0, *yevalm = 0;
    double h[2], ll_corner[2], ur_corner[2];
    /* Nag Types */
    Nag_Boolean print_mesh_enum;
    NagError fail;

    INIT_FAIL(fail);

#ifdef _WIN32
    scanf_s("%"NAG_IFMT%"NAG_IFMT"%*[\n] ", &nxeval, &nyeval);
#else
    scanf("%"NAG_IFMT%"NAG_IFMT"%*[\n] ", &nxeval, &nyeval);
#endif

    if (!(xevalm = NAG_ALLOC(nxeval, double)) ||
        !(yevalm = NAG_ALLOC(nyeval, double)) ||
        !(fevalm = NAG_ALLOC(nxeval*nyeval, double)))
    {
        printf("Allocation failure\n");
        exit_status = -1;
        goto END;
    }
}

```

```

    }

    /* Define the mesh by its lower-left and upper-right corners, which must
    * lie within the bounding box of the spline.
    */

#ifdef _WIN32
    scanf_s("%lf%lf%*[\n] ", &ll_corner[0], &ll_corner[1]);
#else
    scanf("%lf%lf%*[\n] ", &ll_corner[0], &ll_corner[1]);
#endif
#ifdef _WIN32
    scanf_s("%lf%lf%*[\n] ", &ur_corner[0], &ur_corner[1]);
#else
    scanf("%lf%lf%*[\n] ", &ur_corner[0], &ur_corner[1]);
#endif

    for (i = 0; i < 2; i++) {
        ll_corner[i] = MAX(ll_corner[i], pmin[i]);
        ur_corner[i] = MIN(ur_corner[i], pmax[i]);
    }

    /* Set the mesh spacing and the evaluation points. */
    h[0] = (ur_corner[0]-ll_corner[0])/(double) (nxeval-1);
    h[1] = (ur_corner[1]-ll_corner[1])/(double) (nyeval-1);

    for (i = 0; i < nxeval; i++) xevalm[i] = ll_corner[0] + (double) i*h[0];
    for (j = 0; j < nyeval; j++) yevalm[j] = ll_corner[1] + (double) j*h[1];

    /* Evaluate using nag_2d_spline_ts_eval_rect (e02jfc). */
    nag_2d_spline_ts_eval_rect(nxeval, nyeval, xevalm, yevalm, coefs, fevalm,
        iopts, opts, &fail);
    if (fail.code != NE_NOERROR) {
        printf("Error from nag_2d_spline_ts_eval_rect (e02jfc).\n%s\n",
            fail.message);
        exit_status = 1;
        goto END;
    }

    /* Output the computed function values? */
#ifdef _WIN32
    scanf_s("%9s%*[\n] ", print_mesh, _countof(print_mesh));
#else
    scanf("%9s%*[\n] ", print_mesh);
#endif
    print_mesh_enum = (Nag_Boolean) nag_enum_name_to_value(print_mesh);

    printf("\n");
    if (!print_mesh_enum) {
        printf("Outputting of the function values on the mesh is disabled\n");
    } else {
        printf("Values of computed spline at (x_i,y_j):\n"
            "\n%12s%12s%12s\n", "x_i", "y_j", "f(x_i,y_j)");
        for (j = 0; j < nyeval; j++)
            for (i = 0; i < nxeval; i++)
                printf("%12.2f%12.2f%12.2f\n", xevalm[i], yevalm[j],
                    fevalm[j*nxeval+i]);
    }

    END:

    NAG_FREE(xevalm);
    NAG_FREE(yevalm);
    NAG_FREE(fevalm);

    return exit_status;
}

```

## 10.2 Program Data

```
nag_2d_spline_fit_ts_scatter (e02jdc) Example Program Data
100                               : number of data points to fit
1                                 : global smoothing level
Nag_FALSE                         : if C^2 smoothing, supersmooth?
1                                 : no. points for vector evaluation
0 0                               : (x_i,y_i) vector to eval.
101 101                          : (n_x,n_y) size for mesh eval.
0 0                               : mesh lower-left corner
1 1                               : mesh upper-right corner
Nag_FALSE                         : display the computed mesh vals?
```

## 10.3 Program Results

nag\_2d\_spline\_fit\_ts\_scatter (e02jdc) Example Program Results

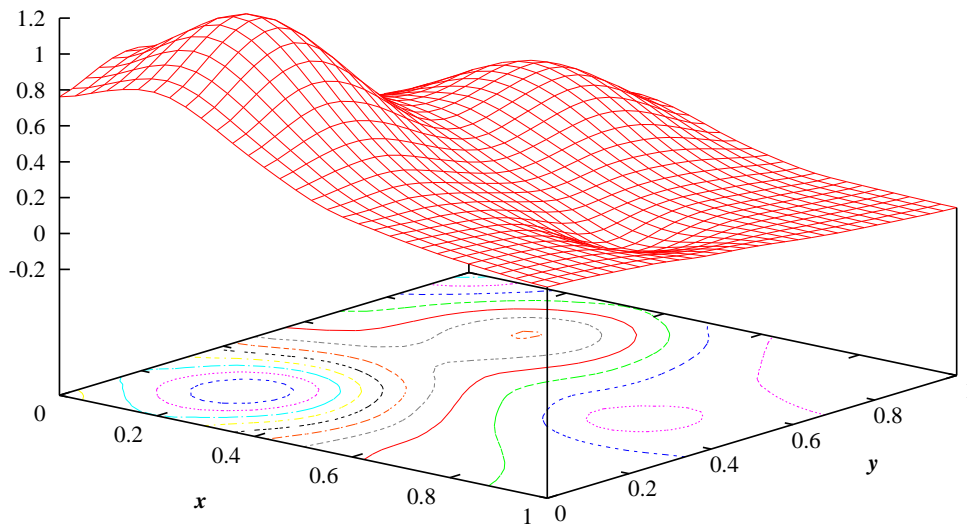
Computing the coefficients of a C<sup>1</sup> spline approximation to Franke's function  
 Using a 6 by 6 grid  
 Using an averaged local approximation

Values of computed spline at (x<sub>i</sub>,y<sub>i</sub>):

x <sub>i</sub>	y <sub>i</sub>	f(x <sub>i</sub> ,y <sub>i</sub> )
0.01	0.01	0.73

Outputting of the function values on the mesh is disabled

**Example Program**  
 Calculation and Evaluation of Bivariate Spline Fit  
 from Scattered Data using Two-Stage Approximation



## 11 Optional Arguments

Several optional arguments in nag\_2d\_spline\_fit\_ts\_scatter (e02jdc) control aspects of the algorithm, methodology used, logic or output. Their values are contained in the arrays **iopts** and **opts**; these must be initialized before calling nag\_2d\_spline\_fit\_ts\_scatter (e02jdc) by first calling nag\_fit\_opt\_set (e02zkc) with **optstr** set to "Initialize = e02jdc".

Each optional argument has an associated default value; to set any of them to a non-default value, or to reset any of them to the default value, use `nag_fit_opt_set` (e02zkc). The current value of an optional argument can be queried using `nag_fit_opt_get` (e02zlc).

The remainder of this section can be skipped if you wish to use the default values for all optional arguments.

The following is a list of the optional arguments available. A full description of each optional argument is provided in Section 11.1.

**Averaged Spline**

**Global Smoothing Level**

**Interpolation Only RBF**

**Local Method**

**Minimum Singular Value LHA**

**Minimum Singular Value LPA**

**Polynomial Starting Degree**

**Radial Basis Function**

**Scaling Coefficient RBF**

**Separation LRBFA**

**Supersmooth C2**

## 11.1 Description of the Optional Arguments

For each option, we give a summary line, a description of the optional argument and details of constraints.

The summary line contains:

the keywords;

a parameter value, where the letters  $a$ ,  $i$  and  $r$  denote options that take character, integer and real values respectively;

the default value.

Keywords and character values are case insensitive.

For `nag_2d_spline_fit_ts_scatter` (e02jdc) the maximum length of the parameter **cvalue** used by `nag_fit_opt_get` (e02zlc) is 16.

**Averaged Spline**  $a$  Default = NO

When the bounding box is triangulated there are 8 equivalent configurations of the mesh. Setting **Averaged Spline** = YES will use the averaged value of the 8 possible local polynomial approximations over each triangle in the mesh. This usually gives better results but at (about 8 times) higher computational cost.

Constraint: **Averaged Spline** = YES or NO.

**Global Smoothing Level**  $i$  Default = 1

The smoothness level for the global spline approximation.

**Global Smoothing Level = 1**

Will use  $C^1$  piecewise cubics.

**Global Smoothing Level = 2**

Will use  $C^2$  piecewise sextics.

Constraint: **Global Smoothing Level** = 1 or 2.

**Interpolation Only RBF** *a* Default = YES

If **Interpolation Only RBF** = YES, each local RBF approximation is computed by interpolation.

If **Interpolation Only RBF** = NO, each local RBF approximation is computed by a discrete least squares approach. This is likely to be more accurate and more expensive than interpolation.

If **Local Method** = HYBRID or POLYNOMIAL, this option setting is ignored.

*Constraint:* **Interpolation Only RBF** = YES or NO.

**Local Method** *a* Default = POLYNOMIAL

The local approximation scheme to use.

**Local Method** = POLYNOMIAL

Uses least squares polynomial approximations.

**Local Method** = HYBRID

Uses hybrid polynomial and RBF approximations.

**Local Method** = RBF

Uses pure RBF approximations.

In general POLYNOMIAL is less computationally expensive than HYBRID is less computationally expensive than RBF with the reverse ordering holding for accuracy of results.

*Constraint:* **Local Method** = POLYNOMIAL, HYBRID or RBF.

**Minimum Singular Value LHA** *r* Default = 1.0

A tolerance measure for accepting or rejecting a local hybrid approximation (LHA) as reliable.

The solution of a local least squares problem solved on each triangle subdomain is accepted as reliable if the minimum singular value  $\sigma$  of the collocation matrix (of polynomial and radial basis function terms) associated with the least squares problem satisfies **Minimum Singular Value LHA**  $\leq \sigma$ .

In general the approximation power will be reduced as **Minimum Singular Value LHA** is reduced. (A small  $\sigma$  indicates that the local data has hidden redundancies which prevent it from carrying enough information for a good approximation to be made.) Setting **Minimum Singular Value LHA** very large may have the detrimental effect that only approximations of low degree are deemed reliable.

A calibration procedure (experimenting with a small subset of the data to be fitted and validating the results) may be needed to choose the most appropriate value for this parameter.

If **Local Method** = POLYNOMIAL or RBF, this option setting is ignored.

*Constraint:* **Minimum Singular Value LHA**  $\geq 0.0$ .

**Minimum Singular Value LPA** *r* Default = 1.0

A tolerance measure for accepting or rejecting a local polynomial approximation (LPA) as reliable. Clearly this setting is relevant when **Local Method** = POLYNOMIAL, but it also may be used when **Local Method** = HYBRID (see Section 9.)

The solution of a local least squares problem solved on each triangle subdomain is accepted as reliable if the minimum singular value  $\sigma$  of the matrix (of Bernstein polynomial values) associated with the least squares problem satisfies **Minimum Singular Value LPA**  $\leq \sigma$ .

In general the approximation power will be reduced as **Minimum Singular Value LPA** is reduced. (A small  $\sigma$  indicates that the local data has hidden redundancies which prevent it from carrying enough information for a good approximation to be made.) Setting **Minimum Singular Value LPA** very large may have the detrimental effect that only approximations of low degree are deemed reliable.

**Minimum Singular Value LPA** will have no effect if **Polynomial Starting Degree** = 0, and it will have little effect if the input data is ‘smooth’ (e.g., from a known function).

A calibration procedure (experimenting with a small subset of the data to be fitted and validating the results) may be needed to choose the most appropriate value for this parameter.



If **Local Method** = RBF, this option setting is ignored.

*Constraint:* **Minimum Singular Value LPA**  $\geq 0.0$ .

**Polynomial Starting Degree** *i*      Default = 5 if **Local Method** = HYBRID,  
Default = 1 otherwise

The degree to be used for the polynomial part in the initial step of each local approximation.

At this initial step the method will attempt to fit with a local approximation having polynomial part of degree **Polynomial Starting Degree**. If **Local Method** = POLYNOMIAL and the approximation is deemed unreliable (according to **Minimum Singular Value LPA**), the degree will be decremented by one and a new local approximation computed, ending with a constant approximation if no other is reliable. If **Local Method** = HYBRID and the approximation is deemed unreliable (according to **Minimum Singular Value LHA**), a pure polynomial approximation of this degree will be tried instead. The method then proceeds as in the POLYNOMIAL case.

**Polynomial Starting Degree** is bounded from above by the maximum possible spline degree, which is 6 (when performing  $C^2$  global super-smoothing). Note that the best-case approximation error (see Section 7) for  $C^2$  smoothing with **Supersmooth C2** = NO is achieved for local polynomials of degree 5; that is, for this level of global smoothing no further benefit is gained by setting **Polynomial Starting Degree** = 6.

The default value gives a good compromise between efficiency and accuracy. In general the best approximation can be obtained by setting:

If **Local Method** = POLYNOMIAL

    if **Global Smoothing Level** = 1, **Polynomial Starting Degree** = 3;

    if **Global Smoothing Level** = 2;

        if **Supersmooth C2** = NO, **Polynomial Starting Degree** = 5;

        otherwise **Polynomial Starting Degree** = 6.

If **Local Method** = HYBRID, **Polynomial Starting Degree** as small as possible.

If **Local Method** = RBF, this option setting is ignored.

*Constraints:*

if **Local Method** = HYBRID,

    if **Radial Basis Function** = MQ2, MQ3, TPS or POLYHARMONIC3,  
    **Polynomial Starting Degree**  $\geq 1$ ;

    if **Radial Basis Function** = TPS4 or POLYHARMONIC5,  
    **Polynomial Starting Degree**  $\geq 2$ ;

    if **Radial Basis Function** = TPS6 or POLYHARMONIC7,  
    **Polynomial Starting Degree**  $\geq 3$ ;

    if **Radial Basis Function** = POLYHARMONIC9,  
    **Polynomial Starting Degree**  $\geq 4$ ;

otherwise **Polynomial Starting Degree**  $\geq 0$ ;

if **Local Method** = POLYNOMIAL and **Global Smoothing Level** = 1,  
**Polynomial Starting Degree**  $\leq 3$ ;

otherwise **Polynomial Starting Degree**  $\leq 6$ .

**Radial Basis Function** *a*      Default = MQ  
**Scaling Coefficient RBF** *r*      Default = 1.0

**Radial Basis Function** selects the RBF to use in each local RBF approximation, while **Scaling Coefficient RBF** selects the scale factor to use in its evaluation, as described below.

A calibration procedure (experimenting with a small subset of the data to be fitted and validating the results) may be needed to choose the most appropriate scale factor and RBF.

If **Local Method** = POLYNOMIAL, these option settings are ignored.

If **Local Method** = HYBRID or RBF, the following (conditionally) positive definite functions may be chosen.

Define  $R = \sqrt{x^2 + y^2}$  and  $\rho = R/r$ .

GAUSS	Gaussian $\exp(-\rho^2)$
IMQ	inverse multiquadric $1/\sqrt{r^2 + R^2}$
IMQ2	inverse multiquadric $1/(r^2 + R^2)$
IMQ3	inverse multiquadric $1/(r^2 + R^2)^{(3/2)}$
IMQ0.5	inverse multiquadric $1/(r^2 + R^2)^{(1/4)}$
WENDLAND31	H. Wendland's $C^2$ function $\max(0, 1 - \rho)^4(4\rho + 1)$
WENDLAND32	H. Wendland's $C^4$ function $\max(0, 1 - \rho)^6(35\rho^2 + 18\rho + 3)$
WENDLAND33	H. Wendland's $C^6$ function $\max(0, 1 - \rho)^8(32\rho^3 + 25\rho^2 + 8\rho + 1)$
BUHMANN3	M. Buhmann's $C^3$ function $112/45\rho^{(9/2)} + 16/3\rho^{(7/2)} - 7\rho^4 - 14/15\rho^2 + 1/9$ if $\rho \leq 1$ , 0 otherwise
MQ	multiquadric $\sqrt{r^2 + R^2}$
MQ1.5	multiquadric $(r^2 + R^2)^{(1.5/2)}$
POLYHARMONIC1.5	polyharmonic spline $\rho^{1.5}$
POLYHARMONIC1.75	polyharmonic spline $\rho^{1.75}$

If **Local Method** = HYBRID the following conditionally positive definite functions may also be chosen.

MQ2	multiquadric $(r^2 + R^2)\log(r^2 + R^2)$
MQ3	multiquadric $(r^2 + R^2)^{(3/2)}$
TPS	thin plate spline $\rho^2\log\rho^2$
POLYHARMONIC3	polyharmonic spline $\rho^3$
TPS4	thin plate spline $\rho^4\log\rho^2$
POLYHARMONIC5	polyharmonic spline $\rho^5$
TPS6	thin plate spline $\rho^6\log\rho^2$
POLYHARMONIC7	polyharmonic spline $\rho^7$
POLYHARMONIC9	polyharmonic spline $\rho^9$

*Constraints:*

if **Radial Basis Function** = MQ2, MQ3, TPS or POLYHARMONIC3,  
**Local Method** = HYBRID and **Polynomial Starting Degree**  $\geq 1$ ;  
 if **Radial Basis Function** = TPS4 or POLYHARMONIC5,  
**Local Method** = HYBRID and **Polynomial Starting Degree**  $\geq 2$ ;  
 if **Radial Basis Function** = TPS6 or POLYHARMONIC7,  
**Local Method** = HYBRID and **Polynomial Starting Degree**  $\geq 3$ ;  
 if **Radial Basis Function** = POLYHARMONIC9,  
**Local Method** = HYBRID and **Polynomial Starting Degree**  $\geq 4$ ;  
**Scaling Coefficient RBF**  $> 0.0$ .

**Separation LRBFA**  $r$  Default = 16.0/**Scaling Coefficient RBF**

A knot-separation parameter used to control the condition number of the matrix used in each local RBF approximation (LRBFA). A smaller value may mean greater numerical stability but fewer knots.

If **Local Method** = HYBRID or POLYNOMIAL, this option setting is ignored.

*Constraint:* **Separation LRBFA** > 0.0.

**Supersmooth C2**

*a*

Default = NO

If **Supersmooth C2** = YES, the  $C^2$  spline is generated using additional smoothness constraints. This usually gives better results but at higher computational cost.

If **Global Smoothing Level** = 1 this option setting is ignored.

*Constraint:* **Supersmooth C2** = YES or NO.

---