

# NAG Library Function Document

## nag\_opt\_lsq\_covariance (e04ycc)

### 1 Purpose

nag\_opt\_lsq\_covariance (e04ycc) returns estimates of elements of the variance-covariance matrix of the estimated regression coefficients for a nonlinear least squares problem. The estimates are derived from the Jacobian of the function  $f(x)$  at the solution.

nag\_opt\_lsq\_covariance (e04ycc) may be used following either of the NAG C Library nonlinear least squares functions nag\_opt\_lsq\_no\_deriv (e04fcc), nag\_opt\_lsq\_deriv (e04gbc).

### 2 Specification

```
#include <nag.h>
#include <nage04.h>
void nag_opt_lsq_covariance (Integer job, Integer m, Integer n,
    double fsumsq, double cj[], Nag_E04_Opt *options, NagError *fail)
```

### 3 Description

nag\_opt\_lsq\_covariance (e04ycc) is intended for use when the nonlinear least squares function,  $F(x) = \bar{f}^T(x)f(x)$ , represents the goodness-of-fit of a nonlinear model to observed data. It assumes that the Hessian of  $F(x)$ , at the solution, can be adequately approximated by  $2J^T J$ , where  $J$  is the Jacobian of  $f(x)$  at the solution. The estimated variance-covariance matrix  $C$  is then given by

$$C = \sigma^2 (J^T J)^{-1} \quad J^T J \text{ nonsingular},$$

where  $\sigma^2$  is the estimated variance of the residual at the solution,  $\bar{x}$ , given by

$$\sigma^2 = \frac{F(\bar{x})}{m - n},$$

$m$  being the number of observations and  $n$  the number of variables.

The diagonal elements of  $C$  are estimates of the variances of the estimated regression coefficients. See the e04 Chapter Introduction, Bard (1974) and Wolberg (1967) for further information on the use of the matrix  $C$ .

When  $J^T J$  is singular then  $C$  is taken to be

$$C = \sigma^2 (J^T J)^\dagger,$$

where  $(J^T J)^\dagger$  is the pseudo-inverse of  $J^T J$ , and  $\sigma^2 = \frac{F(\bar{x})}{m - k}$ ,  $k = \text{rank}(J)$  but in this case the argument fail is returned with **fail.code** = NW\_LIN\_DEPEND as a warning to you that  $J$  has linear dependencies in its columns. The assumed rank of  $J$  can be obtained from **fail.ernum**.

The function can be used to find either the diagonal elements of  $C$ , or the elements of the  $j$ th column of  $C$ , or the whole of  $C$ .

nag\_opt\_lsq\_covariance (e04ycc) must be preceded by one of the nonlinear least squares functions mentioned in Section 1, and requires the arguments **fsumsq** and **options** to be supplied by those functions. **fsumsq** is the residual sum of squares  $F(\bar{x})$  while the structure **options** contains the members **options→s** and **options→v** which give the singular values and right singular vectors respectively in the singular value decomposition of  $J$ .

## 4 References

Bard Y (1974) *Nonlinear Parameter Estimation* Academic Press

Wolberg J R (1967) *Prediction Analysis* Van Nostrand

## 5 Arguments

- 1: **job** – Integer *Input*  
*On entry:* indicates which elements of  $C$  are returned as follows:  
**job** =  $-1$   
The  $n$  by  $n$  symmetric matrix  $C$  is returned.  
**job** =  $0$   
The diagonal elements of  $C$  are returned.  
**job** >  $0$   
The elements of column **job** of  $C$  are returned.  
*Constraint:*  $-1 \leq \text{job} \leq \mathbf{n}$ .
- 2: **m** – Integer *Input*  
*On entry:* the number  $m$  of observations (residuals  $f_i(x)$ ).  
*Constraint:*  $\mathbf{m} \geq \mathbf{n}$ .
- 3: **n** – Integer *Input*  
*On entry:* the number  $n$  of variables ( $x_j$ ).  
*Constraint:*  $1 \leq \mathbf{n} \leq \mathbf{m}$ .
- 4: **fsumsq** – double *Input*  
*On entry:* the sum of squares of the residuals,  $F(\bar{x})$ , at the solution  $\bar{x}$ , as returned by the nonlinear least squares function.  
*Constraint:* **fsumsq**  $\geq 0.0$ .
- 5: **cj[n]** – double *Output*  
*On exit:* with **job** =  $0$ , **cj** returns the  $n$  diagonal elements of  $C$ . With **job** =  $j > 0$ , **cj** returns the  $n$  elements of the  $j$ th column of  $C$ . When **job** =  $-1$ , **cj** is not referenced.
- 6: **options** – Nag\_E04\_Opt \* *Input/Output*  
*On entry/exit:* the structure used in the call to the nonlinear least squares function. The following members are relevant to nag\_opt\_lsq\_covariance (e04ycc), their values should not be altered between the call to the least squares function and the call to nag\_opt\_lsq\_covariance (e04ycc).
- s** – double *Input*  
*On entry:* the pointer to the  $n$  singular values of the Jacobian as returned by the nonlinear least squares function.
- v** – double *Input/Output*  
*On entry:* the pointer to the  $n$  by  $n$  right-hand orthogonal matrix (the right singular vectors) of  $J$  as returned by the nonlinear least squares function.  
*On exit:* when **job**  $\geq 0$  then **v** is unchanged.

When  $\mathbf{job} = -1$  then the leading  $n$  by  $n$  part of  $\mathbf{v}$  is overwritten by the  $n$  by  $n$  matrix  $C$ . Matrix element  $i, j$  is held in  $\mathbf{v}[(i-1) \times \mathbf{tdv} + j - 1]$  for  $i = 1, 2, \dots, n$  and  $j = 1, 2, \dots, n$ .

**tdv** – Integer *Input*

*On entry:* the trailing dimension used by  $\mathbf{v}$ .

7: **fail** – NagError \* *Input/Output*

The NAG error argument (see Section 2.7 in How to Use the NAG Library and its Documentation).

## 6 Error Indicators and Warnings

### NE\_2\_INT\_ARG\_GT

On entry,  $\mathbf{job} = \langle value \rangle$  while  $\mathbf{n} = \langle value \rangle$ . These arguments must satisfy  $\mathbf{job} \leq \mathbf{n}$ .

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

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

### NE\_INT\_ARG\_LT

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

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

### NE\_REAL\_ARG\_LT

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

### NE\_SINGULAR\_VALUES

The singular values are all zero, so that at the solution the Jacobian matrix has rank 0.

### NW\_LIN\_DEPEND

At the solution the Jacobian matrix contains linear, or near linear, dependencies amongst its columns.  $J$  assumed to have rank  $\langle value \rangle$ .

In this case the required elements of  $C$  have still been computed based upon  $J$  having an assumed rank given by **fail.errnum**. The rank is computed by regarding singular values **options.s[j]** that are not larger than  $10\epsilon \times \mathbf{options.s[0]}$  as zero, where  $\epsilon$  is the **machine precision** (see nag\_machine\_precision (X02AJC)). If you expect near linear dependencies at the solution and are happy with this tolerance in determining rank you should not call nag\_opt\_lsq\_covariance (e04ycc) with the null pointer NAGERR\_DEFAULT as the argument **fail** but should specifically declare and initialize a NagError structure for the argument **fail**.

### Overflow

If overflow occurs then either an element of  $C$  is very large, or the singular values or singular vectors have been incorrectly supplied.

## 7 Accuracy

The computed elements of  $C$  will be the exact covariances corresponding to a closely neighbouring Jacobian matrix  $J$ .

## 8 Parallelism and Performance

nag\_opt\_lsq\_covariance (e04ycc) is not threaded in any implementation.

## 9 Further Comments

When  $\mathbf{job} = -1$  the time taken by the function is approximately proportional to  $n^3$ . When  $\mathbf{job} \geq 0$  the time taken by the function is approximately proportional to  $n^2$ .

## 10 Example

This example estimates the variance-covariance matrix  $C$  for the least squares estimates of  $x_1$ ,  $x_2$  and  $x_3$  in the model

$$y = x_1 + \frac{t_1}{x_2 t_2 + x_3 t_3}$$

using the 15 sets of data given in the following table:

$y$	0.14	1.0	endgroup	15.0	1.00.182.0	endgroup	14.02.00.223.0	endgroup	13.03.00.254.0	endgroup	12.04.00.295.0	endgroup	11.05
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The program uses (0.5,1.0,1.5) as the initial guess at the position of the minimum and computes the least squares solution using nag\_opt\_lsq\_no\_deriv (e04fcc). Note that the structure **options** is initialized by nag\_opt\_init (e04xcc) before calling nag\_opt\_lsq\_no\_deriv (e04fcc). See the function documents for nag\_opt\_lsq\_no\_deriv (e04fcc), nag\_opt\_init (e04xcc) and nag\_opt\_free (e04xzc) for further information.

### 10.1 Program Text

```
/* nag_opt_lsq_covariance (e04ycc) Example Program.
*
* NAGPRODCODE Version.
*
* Copyright 2016 Numerical Algorithms Group.
*
* Mark 26, 2016.
*/
#include <nag.h>
#include <stdio.h>
#include <string.h>
#include <math.h>
#include <nag_stlib.h>
#include <nage04.h>

#ifndef __cplusplus
extern "C"
{
#endif
    static void NAG_CALL lsqfun(Integer m, Integer n, const double x[],
                                double fvec[], Nag_Comm *comm);
#ifndef __cplusplus
}
#endif

/* Define a user structure template to store data in lsqfun */
struct user
{
    double *y;
    double *t;
};

#define T(I, J) t[(I) *tdt + J]

int main(void)
{
    Integer exit_status = 0, i, j, job, m, n, nt, tdj, tdt;
    NagError fail;
    Nag_Comm comm;
```

```

Nag_E04_Opt options;
double *cj = 0, *fjac = 0, fsumsq, *fvec = 0, *x = 0;
struct user s;

INIT_FAIL(fail);

s.y = 0;
s.t = 0;
printf("nag_opt_lsq_covariance (e04ycc) Example Program Results\n");
#ifndef _WIN32
scanf_s("%*[^\n]"); /* Skip heading in data file */
#else
scanf("%*[^\n]"); /* Skip heading in data file */
#endif
n = 3;
m = 15;
nt = 3;
if (n >= 1 && n <= m) {
    if (!(fjac = NAG_ALLOC(m * n, double)) ||
        !(fvec = NAG_ALLOC(m, double)) ||
        !(x = NAG_ALLOC(n, double)) ||
        !(cj = NAG_ALLOC(n, double)) ||
        !(s.y = NAG_ALLOC(m, double)) || !(s.t = NAG_ALLOC(m * nt, double)))
    {
        printf("Allocation failure\n");
        exit_status = -1;
        goto END;
    }
    tdj = n;
    tdt = nt;
}
else {
    printf("Invalid n or m.\n");
    exit_status = 1;
    return exit_status;
}
/* Read data into structure.
 * Observations t (j = 0, 1, 2) are held in s->t[i][j]
 * (i = 0, 1, 2, . . . , 14)
 */
for (i = 0; i < m; ++i) {
#ifdef _WIN32
    scanf_s("%lf", &s.y[i]);
#else
    scanf("%lf", &s.y[i]);
#endif
#ifdef _WIN32
    for (j = 0; j < nt; ++j)
        scanf_s("%lf", &s.T(i, j));
#else
    for (j = 0; j < nt; ++j)
        scanf("%lf", &s.T(i, j));
#endif
}
/* Set up the starting point */
x[0] = 0.5;
x[1] = 1.0;
x[2] = 1.5;

/* nag_opt_init (e04xxc).
 * Initialization function for option setting
 */
nag_opt_init(&options); /* Initialize options structure */

/* Turn off most monitoring information from e04fcc */
options.list = Nag_FALSE;
options.print_level = Nag_Soln;

/* Assign address of user defined structure to
 * comm.p for communication to lsqfun().
```

```

/*
comm.p = (Pointer) &s;

/* nag_opt_lsq_no_deriv (e04fcc).
 * Unconstrained nonlinear least squares (no derivatives
 * required)
 */
fflush(stdout);
nag_opt_lsq_no_deriv(m, n, lsqfun, x, &fsumsq, fvec, fjac, tdj,
                      &options, &comm, &fail);
if (fail.code != NE_NOERROR && fail.code != NW_COND_MIN) {
    printf("Error from nag_opt_lsq_no_deriv (e04fcc).\n%s\n", fail.message);
    exit_status = 1;
    goto END;
}

job = 0;
/* nag_opt_lsq_covariance (e04ycc).
 * Covariance matrix for nonlinear least squares
 */
nag_opt_lsq_covariance(job, m, n, fsumsq, cj, &options, &fail);
if (fail.code != NE_NOERROR) {
    printf("Error from nag_opt_lsq_covariance (e04ycc).\n%s\n", fail.message);
    exit_status = 1;
    goto END;
}

printf("\nEstimates of the variances of the sample regression");
printf(" coefficients are:\n");
for (i = 0; i < n; ++i)
    printf(" %15.5e", cj[i]);
printf("\n");

/* Free memory allocated to pointers s and v */
/* nag_opt_free (e04xzc).
 * Memory freeing function for use with option setting
 */
nag_opt_free(&options, "all", &fail);
if (fail.code != NE_NOERROR) {
    printf("Error from nag_opt_free (e04xzc).\n%s\n", fail.message);
    exit_status = 1;
    goto END;
}
END:
NAG_FREE(fjac);
NAG_FREE(fvec);
NAG_FREE(x);
NAG_FREE(cj);
NAG_FREE(s.y);
NAG_FREE(s.t);

return exit_status;
}

static void NAG_CALL lsqfun(Integer m, Integer n, const double x[],
                           double fvec[], Nag_Comm *comm)
{
/* Function to evaluate the residuals.
 *
 * The address of the user defined structure is recovered in each call
 * to lsqfun() from comm->p and the structure used in the calculation
 * of the residuals.
 */

Integer i, tdt;
struct user *s = (struct user *) comm->p;
tdt = n;

```

```

for (i = 0; i < m; ++i)
    fvec[i] = x[0] +
        s->T(i, 0) / (x[1] * s->T(i, 1) + x[2] * s->T(i, 2)) - s->y[i];
} /* lsqfun */

```

## 10.2 Program Data

```
nag_opt_lsq_covariance (e04ycc) Example Program Data
0.14 1.0 15.0 1.0
0.18 2.0 14.0 2.0
0.22 3.0 13.0 3.0
0.25 4.0 12.0 4.0
0.29 5.0 11.0 5.0
0.32 6.0 10.0 6.0
0.35 7.0 9.0 7.0
0.39 8.0 8.0 8.0
0.37 9.0 7.0 7.0
0.58 10.0 6.0 6.0
0.73 11.0 5.0 5.0
0.96 12.0 4.0 4.0
1.34 13.0 3.0 3.0
2.10 14.0 2.0 2.0
4.39 15.0 1.0 1.0
```

## 10.3 Program Results

```
nag_opt_lsq_covariance (e04ycc) Example Program Results
```

Final solution:

x	g	Residuals
8.24106e-02	3.0423e-10	-5.8811e-03
1.13304e+00	-2.0975e-10	-2.6534e-04
2.34370e+00	-7.1256e-11	2.7469e-04
		6.5415e-03
		-8.2299e-04
		-1.2995e-03
		-4.4631e-03
		-1.9963e-02
		8.2216e-02
		-1.8212e-02
		-1.4811e-02
		-1.4710e-02
		-1.1208e-02
		-4.2040e-03
		6.8079e-03

The sum of squares is 8.2149e-03.

Estimates of the variances of the sample regression coefficients are:  
1.53120e-04      9.48024e-02      8.77806e-02

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