

NAG C Library Function Document

nag_robust_m_regsn_param_var (g02hfc)

1 Purpose

nag_robust_m_regsn_param_var (g02hfc) calculates an estimate of the asymptotic variance-covariance matrix for the bounded influence regression estimates (M-estimates). It is intended for use with nag_robust_m_regsn_user_fn (g02hdc).

2 Specification

```
void nag_robust_m_regsn_param_var (Nag_OrderType order,
    double (*psi)(double t, Nag_Comm *comm),
    double (*psp)(double t, Nag_Comm *comm),
    Nag_RegType regtype, Nag_CovMatrixEst covmat_est, double sigma, Integer n,
    Integer m, const double x[], Integer pdx, const double rs[],
    const double wgt[], double cov[], Integer pdc, double comm_arr[],
    Nag_Comm *comm, NagError *fail)
```

3 Description

For a description of bounded influence regression see nag_robust_m_regsn_user_fn (g02hdc). Let θ be the regression parameters and let C be the asymptotic variance-covariance matrix of $\hat{\theta}$. Then for Huber type regression

$$C = f_H(X^T X)^{-1} \hat{\sigma}^2,$$

where

$$\begin{aligned} f_H &= \frac{1}{n-m} \frac{\sum_{i=1}^n \psi^2(r_i/\hat{\sigma})}{\left(\frac{1}{n} \sum \psi'(r_i/\hat{\sigma})\right)^2} \kappa^2 \\ \kappa^2 &= 1 + \frac{m}{n} \frac{\frac{1}{n} \sum_{i=1}^n (\psi'(r_i/\hat{\sigma}) - \frac{1}{n} \sum_{i=1}^n \psi'(r_i/\hat{\sigma}))^2}{\left(\frac{1}{n} \sum_{i=1}^n \psi'(r_i/\hat{\sigma})\right)^2}, \end{aligned}$$

see Huber (1981) and Marazzi (1987b).

For Mallows and Schweppe type regressions, C is of the form

$$\frac{\hat{\sigma}^2}{n} S_1^{-1} S_2 S_1^{-1},$$

where $S_1 = \frac{1}{n} X^T D X$ and $S_2 = \frac{1}{n} X^T P X$.

D is a diagonal matrix such that the i th element approximates $E(\psi'(r_i/(\sigma w_i)))$ in the Schweppe case and $E(\psi'(r_i/\sigma)w_i)$ in the Mallows case.

P is a diagonal matrix such that the i th element approximates $E(\psi^2(r_i/(\sigma w_i))w_i^2)$ in the Schweppe case and $E(\psi^2(r_i/\sigma)w_i^2)$ in the Mallows case.

Two approximations are available in nag_robust_m_regsn_param_var (g02hfc):

1. Average over the r_i

Schweppe	Mallows
$D_i = \left(\frac{1}{n} \sum_{j=1}^n \psi' \left(\frac{r_j}{\hat{\sigma} w_i} \right) \right) w_i$	$D_i = \left(\frac{1}{n} \sum_{j=1}^n \psi' \left(\frac{r_j}{\hat{\sigma}} \right) \right) w_i$
$P_i = \left(\frac{1}{n} \sum_{j=1}^n \psi^2 \left(\frac{r_j}{\hat{\sigma} w_i} \right) \right) w_i^2$	$P_i = \left(\frac{1}{n} \sum_{j=1}^n \psi^2 \left(\frac{r_j}{\hat{\sigma}} \right) \right) w_i^2$

2. Replace expected value by observed

Schweppe	Mallows
$D_i = \psi' \left(\frac{r_i}{\hat{\sigma} w_i} \right) w_i$	$D_i = \psi' \left(\frac{r_i}{\hat{\sigma}} \right) w_i$
$P_i = \psi^2 \left(\frac{r_i}{\hat{\sigma} w_i} \right) w_i^2$	$P_i = \psi^2 \left(\frac{r_i}{\hat{\sigma}} \right) w_i^2$

See Hampel *et al.* (1986) and Marazzi (1987b).

In all cases $\hat{\sigma}$ is a robust estimate of σ .

nag_robust_m_regsn_param_var (g02hfc) is based on routines in ROBETH; see Marazzi (1987b).

4 References

Hampel F R, Ronchetti E M, Rousseeuw P J and Stahel W A (1986) *Robust Statistics. The Approach Based on Influence Functions* Wiley

Huber P J (1981) *Robust Statistics* Wiley

Marazzi A (1987b) Subroutines for robust and bounded influence regression in ROBETH *Cah. Rech. Doc. IUMSP, No. 3 ROB 2 Institut Universitaire de Médecine Sociale et Préventive, Lausanne*

5 Parameters

- 1: **order** – Nag_OrderType *Input*

On entry: the **order** parameter specifies the two-dimensional storage scheme being used, i.e., row-major ordering or column-major ordering. C language defined storage is specified by **order = Nag_RowMajor**. See Section 2.2.1.4 of the Essential Introduction for a more detailed explanation of the use of this parameter.

Constraint: **order = Nag_RowMajor** or **Nag_ColMajor**.

- 2: **psi** *Function*

psi must return the value of the ψ function for a given value of its argument.

Its specification is:

```
double psi (double t, Nag_Comm *comm)
```

- 1: **t** – double *Input*
On entry: the argument for which **psi** must be evaluated.
- 2: **comm** – NAG_Comm * *Input/Output*
The NAG communication parameter (see the Essential Introduction).

- 3: **psp** *Function*

psp must return the value of $\psi'(t) = \frac{d}{dt}\psi(t)$ for a given value of its argument.

Its specification is:

- double **psp** (double **t**, Nag_Comm ***comm**)
- 1: **t** – double *Input*
On entry: the argument for which **psp** must be evaluated.
- 2: **comm** – Nag_Comm * *Input/Output*
The NAG communication parameter (see the Essential Introduction).
- 4: **regtype** – Nag_RegType *Input*
On entry: the type of regression for which the asymptotic variance-covariance matrix is to be calculated.
If **regtype** = Nag_HuberReg, Huber type regression.
If **regtype** = Nag_MallowsReg, Mallows type regression.
If **regtype** = Nag_SchweppeReg, Schwepppe type regression.
- 5: **covmat_est** – Nag_CovMatrixEst *Input*
On entry: if **regtype** ≠ Nag_HuberReg, **covmat_est** must specify the approximation to be used.
If **covmat_est** = Nag_CovMatAve, averaging over residuals.
If **covmat_est** = Nag_CovMatObs, replacing expected by observed.
If **regtype** = Nag_HuberReg, **covmat_est** is not referenced.
- 6: **sigma** – double *Input*
On entry: the value of $\hat{\sigma}$, as given by nag_robust_m_regsn_user_fn (g02hdc).
Constraint: **sigma** > 0.
- 7: **n** – Integer *Input*
On entry: the number, n , of observations.
Constraint: **n** > 1.
- 8: **m** – Integer *Input*
On entry: the number, m , of independent variables.
Constraint: $1 \leq m < n$.
- 9: **x[dim]** – const double *Input*
Note: the dimension, dim , of the array **x** must be at least $\max(1, pdx \times m)$ when **order** = Nag_ColMajor and at least $\max(1, pdx \times n)$ when **order** = Nag_RowMajor.
Where $X(i, j)$ appears in this document, it refers to the array element
if **order** = Nag_ColMajor, **x**[($j - 1$) \times **pdx** + $i - 1$];
if **order** = Nag_RowMajor, **x**[($i - 1$) \times **pdx** + $j - 1$].
On entry: the values of the X matrix, i.e., the independent variables. $X(i, j)$ must contain the ij th element of X , for $i = 1, 2, \dots, n$, $j = 1, 2, \dots, m$.
- 10: **pdx** – Integer *Input*
On entry: the stride separating matrix row or column elements (depending on the value of **order**) in the array **x**.

Constraints:

if **order** = Nag_ColMajor, **pdx** $\geq \mathbf{n}$;
 if **order** = Nag_RowMajor, **pdx** $\geq \mathbf{m}$.

11: **rs[n]** – const double *Input*

On entry: the residuals from the bounded influence regression. These are given by nag_robust_m_regn_user_fn (g02hdc).

12: **wgt[n]** – const double *Input*

On entry: if **regtype** \neq Nag_HuberReg, **wgt** must contain the vector of weights used by the bounded influence regression. These should be used with nag_robust_m_regn_user_fn (g02hdc).

If **regtype** = Nag_HuberReg, **wgt** is not referenced.

Constraint: if **regtype** \neq Nag_HuberReg, **wgt**[*i*] ≥ 0.0 for *i* = 0, 1,

13: **cov[dim]** – double *Output*

Note: the dimension, *dim*, of the array **c** must be at least **pdc** \times **m**.

If **order** = Nag_ColMajor, the (*i*, *j*)th element of the matrix *C* is stored in **cov**[(*j* − 1) \times **pdc** + *i* − 1] and if **order** = Nag_RowMajor, the (*i*, *j*)th element of the matrix *C* is stored in **cov**[(*i* − 1) \times **pdc** + *j* − 1].

On exit: the estimate of the variance-covariance matrix.

14: **pdc** – Integer *Input*

On entry: the stride separating matrix row or column elements (depending on the value of **order**) in the array **cov**.

Constraint: **pdc** $\geq \mathbf{m}$.

15: **comm_arr[dim]** – double *Output*

Note: the dimension, *dim*, of the array **comm_arr** must be at least **m** \times (**n** + **m** + 1) + 2 \times **n**.

On exit: if **regtype** \neq Nag_HuberReg, **comm_arr**[*i* − 1], for *i* = 1, 2, ..., **n**, will contain the diagonal elements of the matrix *D* and **comm_arr**[*i* − 1], for *i* = **n** + 1, **n** + 2, ..., 2**n**, will contain the diagonal elements of matrix *P*.

16: **comm** – NAG_Comm * *Input/Output*

The NAG communication parameter (see the Essential Introduction).

17: **fail** – NagError * *Input/Output*

The NAG error parameter (see the Essential Introduction).

6 Error Indicators and Warnings

NE_INT

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

Constraint: **n** > 1.

On entry, **pdx** = $\langle\text{value}\rangle$.

Constraint: **pdx** > 0.

On entry, **pdc** = $\langle\text{value}\rangle$.

Constraint: **pdc** > 0.

On entry, **m** = $\langle\text{value}\rangle$.

Constraint: **m** ≥ 1 .

NE_INT_2

On entry, $\mathbf{m} = \langle value \rangle$, $\mathbf{n} = \langle value \rangle$.

Constraint: $1 \leq \mathbf{m} < \mathbf{n}$.

On entry, $\mathbf{pdx} = \langle value \rangle$, $\mathbf{n} = \langle value \rangle$.

Constraint: $\mathbf{pdx} \geq \mathbf{n}$.

On entry, $\mathbf{pdx} = \langle value \rangle$, $\mathbf{m} = \langle value \rangle$.

Constraint: $\mathbf{pdx} \geq \mathbf{m}$.

On entry, $\mathbf{pdc} = \langle value \rangle$, $\mathbf{m} = \langle value \rangle$.

Constraint: $\mathbf{pdc} \geq \mathbf{m}$.

On entry, $\mathbf{m} = \langle value \rangle$, $\mathbf{pdc} = \langle value \rangle$.

Constraint: $\mathbf{pdc} \geq \mathbf{m}$.

On entry, $\mathbf{n} \leq \mathbf{m}$: $\mathbf{n} = \langle value \rangle$, $\mathbf{m} = \langle value \rangle$.

NE_ENUM_INT

On entry, $\mathbf{regtype} = \langle value \rangle$, $\mathbf{wgt} = \langle value \rangle$.

Constraint: if $\mathbf{regtype} \neq \text{Nag_HuberReg}$, $\mathbf{wgt}[i] \geq 0.0$ for $i = 0, \dots,$

NE_CORRECTION_FACTOR

Correction factor = 0 (Huber type regression).

NE_POS_DEF

$X^T X$ matrix not positive definite.

NE_REAL

On entry, $\mathbf{sigma} = \langle value \rangle$.

Constraint: $\mathbf{sigma} \geq 0$.

NE_REAL_ARRAY_ELEM_CONS

On entry, an element of $\mathbf{wgt} < 0$.

NE_SINGULAR

S_1 matrix is singular or almost singular.

NE_ALLOC_FAIL

Memory allocation failed.

NE_BAD_PARAM

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

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 consult NAG for assistance.

7 Accuracy

In general, the accuracy of the variance-covariance matrix will depend primarily on the accuracy of the results from nag_robust_m_regrn_user_fn (g02hdc).

8 Further Comments

This routine is only for situations in which X has full column rank.

Care has to be taken in the choice of the ψ function since if $\psi'(t) = 0$ for too wide a range then either the value of f_H will not exist or too many values of D_i will be zero and it will not be possible to calculate C .

9 Example

The asymptotic variance-covariance matrix is calculated for a Schweppe type regression. The values of X , $\hat{\sigma}$ and the residuals and weights are read in. The averaging over residuals approximation is used.

9.1 Program Text

```
/* nag_robust_m_regsn_param_var (g02hfc) Example Program.
*
* Copyright 2002 Numerical Algorithms Group.
*
* Mark 7, 2002.
*/
#include <math.h>
#include <stdio.h>
#include <nag.h>
#include <nag_stdlib.h>
#include <nagg02.h>

static double psi(double t, Nag_Comm *comm);
static double psp(double t, Nag_Comm *comm);
int main(void)
{
    /* Scalars */
    double sigma;
    Integer exit_status, i, ic, ix, j, k, m, n;
    Integer pdc, pdx;
    NagError fail;
    Nag_OrderType order;
    Nag_Comm comm;

    /* Arrays */
    double *cov=0, *rs=0, *wgt=0, *comm_arr=0, *x=0;

#ifdef NAG_COLUMN_MAJOR
#define COV(I,J) cov[(J-1)*pdc + I - 1]
#define X(I,J) x[(J-1)*pdx + I - 1]
    order = Nag_ColMajor;
#else
#define COV(I,J) cov[(I-1)*pdc + J - 1]
#define X(I,J) x[(I-1)*pdx + J - 1]
    order = Nag_RowMajor;
#endif
    exit_status = 0;
    INIT_FAIL(fail);

    Vprintf("g02hfc Example Program Results\n");

    /* Skip heading in data file */
    Vscanf("%*[^\n] ");

    /* Read in the dimensions of X */
    Vscanf("%ld%ld%*[^\n] ", &n, &m);

    /* Allocate memory */
    if ( !(cov = NAG_ALLOC(m * m, double)) ||
        !(rs = NAG_ALLOC(n, double)) ||
        !(wgt = NAG_ALLOC(n, double)) ||
        !(comm_arr = NAG_ALLOC(m*(n+m+1)+2*n, double)) ||
        !(x = NAG_ALLOC(n, double)) )
        exit_status = 1;
}
```

```

    !(x = NAG_ALLOC(n * m, double)) )
{
    Vprintf("Allocation failure\n");
    exit_status = -1;
    goto END;
}

#ifndef NAG_COLUMN_MAJOR
    pdc = m;
    pdx = n;
#else
    pdc = m;
    pdx = m;
#endif

Vprintf("\n");

/* Read in the X matrix */
for (i = 1; i <= n; ++i)
{
    for (j = 1; j <= m; ++j)
    {
        Vscanf("%lf", &x(i,j));
    }
    Vscanf("%*[^\n] ");
}

/* Read in sigma */
Vscanf("%lf%*[^\n] ", &sigma);

/* Read in weights and residuals */
for (i = 1; i <= n; ++i)
{
    Vscanf("%lf%lf%*[^\n] ", &wgt[i - 1], &rs[i - 1]);
}

/* Set other parameter values */
ix = 5;
ic = 3;
/* Set parameters for Schweppe type regression */
g02hfc(order, psi, psp, Nag_SchweppeReg, Nag_CovMatAve, sigma, n, m, x, pdx,
        rs, wgt, cov, pdc, comm_arr, &comm, &fail);
if (fail.code != NE_NOERROR)
{
    Vprintf("Error from g02hfc.\n%s\n", fail.message);
    exit_status = 1;
    goto END;
}

Vprintf("Covariance matrix\n");
for (j = 1; j <= m; ++j)
{
    for (k = 1; k <= m; ++k)
    {
        Vprintf("%10.4f%s", COV(j,k), k%6 == 0 || k == m ? "\n": " ");
    }
}

END:
if (cov) NAG_FREE(cov);
if (rs) NAG_FREE(rs);
if (wgt) NAG_FREE(wgt);
if (comm_arr) NAG_FREE(comm_arr);
if (x) NAG_FREE(x);

return exit_status;
}

static double psi(double t, Nag_Comm *comm)
{

```

```

double ret_val;

if (t <= -1.5)
{
    ret_val = -1.5;
}
else if (fabs(t) < 1.5)
{
    ret_val = t;
}
else
{
    ret_val = 1.5;
}
return ret_val;
}

static double psp(double t, Nag_Comm *comm)
{
    double ret_val;

    ret_val = 0.0;
    if (fabs(t) < 1.5)
    {
        ret_val = 1.0;
    }
    return ret_val;
}

```

9.2 Program Data

g02hfc Example Program Data

```

5      3                  : N   M
1.0  -1.0  -1.0          : X1  X2  X3
1.0  -1.0   1.0
1.0   1.0  -1.0
1.0   1.0   1.0
1.0   0.0   3.0          : End of X1 X2 and X3 values

20.7783                 : SIGMA

0.4039   0.5643          : Weights and residuals, WGT and RS
0.5012  -1.1286
0.4039   0.5643
0.5012  -1.1286
0.3862   1.1286          : End of weights and residuals

```

9.3 Program Results

g02hfc Example Program Results

```

Covariance matrix
  0.2070   0.0000   -0.0478
  0.0000   0.2229   -0.0000
 -0.0478  -0.0000    0.0796

```
