

nag_regsn_mult_linear_add_var (g02dec)

1. Purpose

nag_regsn_mult_linear_add_var (g02dec) adds a new independent variable to a general linear regression model.

2. Specification

```
#include <nag.h>
#include <nagg02.h>

void nag_regsn_mult_linear_add_var(Integer n, Integer ip, double q[],
    Integer tdq, double p[], double wt[], double x[], double *rss,
    double tol, NagError *fail)
```

3. Description

A linear regression model may be built up by adding new independent variables to an existing model. `nag_regsn_mult_linear_add_var` updates the QR decomposition used in the computation of the linear regression model. The QR decomposition may come from `nag_regsn_mult_linear (g02dac)` or a previous call to `nag_regsn_mult_linear_add_var`. The general linear regression model is defined by:

$$y = X\beta + \varepsilon$$

where y is a vector of n observations on the dependent variable, X is an n by p matrix of the independent variables of column rank k , β is a vector of length p of unknown parameters, and ε is a vector of length n of unknown random errors such that $\text{var } \varepsilon = V\sigma^2$, where V is a known diagonal matrix.

If $V = I$, the identity matrix, then least-squares estimation is used. If $V \neq I$, then for a given weight matrix $W \propto V^{-1}$, weighted least-squares estimation is used.

The least-squares estimates, $\hat{\beta}$ of the parameters β minimize $(y - X\beta)^T(y - X\beta)$ while the weighted least-squares estimates minimize $(y - X\beta)^TW(y - X\beta)$.

The parameter estimates may be found by computing a QR decomposition of X (or $W^{\frac{1}{2}}X$ in the weighted case), i.e.,

$$X = QR^* \quad (\text{or } W^{\frac{1}{2}}X = QR^*)$$

where $R^* = \begin{pmatrix} R \\ 0 \end{pmatrix}$ and R is a p by p upper triangular matrix and Q is an n by n orthogonal matrix.

If R is of full rank, then $\hat{\beta}$ is the solution to:

$$R\hat{\beta} = c_1$$

where $c = Q^T y$ (or $Q^T W^{\frac{1}{2}} y$) and c_1 is the first p elements of c .

If R is not of full rank a solution is obtained by means of a singular value decomposition (SVD) of R .

To add a new independent variable, x_{p+1} , R and c have to be updated. The matrix Q_{p+1} is found such that $Q_{p+1}^T[R : Q^T x_{p+1}]$ (or $Q_{p+1}^T[R : Q^T W^{\frac{1}{2}} x_{p+1}]$) is upper triangular. The vector c is then updated by multiplying by Q_{p+1}^T .

The new independent variable is tested to see if it is linearly related to the existing independent variables by checking that at least one of the values $(Q^T x_{p+1})_i$, for $i = p+2, p+3, \dots, n$ is non-zero.

The new parameter estimates, $\hat{\beta}$, can then be obtained by a call to `nag_regsn_mult_linear_upd_model (g02ddc)`.

The function can be used with $p = 0$, in which case R and c are initialized.

4. Parameters

n

Input: the number of observations, n .
 Constraint: $n \geq 1$.

ip

Input: the number of independent variables already in the model, p .
 Constraint: $ip \geq 0$ and $ip < n$.

q[n][tdq]

Input: if $ip \neq 0$, then **q** must contain the results of the QR decomposition for the model with p parameters as returned by nag_regsn_mult_linear (g02dac) or a previous call to nag_regsn_mult_linear_add_var.

If $ip = 0$, then the first column of **q** should contain the n values of the dependent variable, y .

Output: the results of the QR decomposition for the model with $p + 1$ parameters:

the first column of **q** contains the updated value of c ,

the columns 2 to $ip + 1$ are unchanged,

the first $ip + 1$ elements of column $ip + 2$ contain the new column of R , while the remaining $n - ip - 1$ elements contain details of the matrix Q_{p+1} .

tdq

Input: **tdq** the last dimension of the array **q** as declared in the function from which nag_regsn_mult_linear_add_var is called.

Constraint: $tdq \geq ip + 2$.

p[ip+1]

Input: **p** contains further details of the QR decomposition used. The first ip elements of **p** must contain the zeta values for the QR decomposition.

The first ip elements of array **p** are provided by nag_regsn_mult_linear (g02dac) or by previous calls to nag_regsn_mult_linear_add_var.

Output: the first ip elements of **p** are unchanged and the $(ip+1)$ th element contains the zeta value for Q_{p+1} .

wt[n]

Input: if weighted estimates are required, then **wt** must contain the weights to be used in the weighted regression. Otherwise **wt** need not be defined and may be set to the null pointer **NULL**, i.e., $(double *)0$.

If $wt[i] = 0.0$, then the i th observation is not included in the model, in which case the effective number of observations is the number of observations with non-zero weights.

If **wt** = **NULL**, then the effective number of observations is n .

Constraint: **wt** = **NULL** or $wt[i] \geq 0.0$, for $i = 0, 1, \dots, n - 1$.

x[n]

Input: the new independent variable, x .

rss

Output: the residual sum of squares for the new fitted model.

Note: this will only be valid if the model is of full rank, see Section 6.

tol

Input: the value of **tol** is used to decide if the new independent variable is linearly related to independent variables already included in the model. If the new variable is linearly related then c is not updated. The smaller the value of **tol** the stricter the criterion for deciding if there is a linear relationship.

Suggested value: **tol** = 0.000001.

Constraint: **tol** > 0.0.

fail

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

5. Error Indications and Warnings

NE_INT_ARG_LT

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

On entry, **ip** must not be less than 0: **ip** = $\langle value \rangle$.

NE_2_INT_ARG_GE

On entry **ip** = $\langle value \rangle$ while **n** = $\langle value \rangle$. These parameters must satisfy **ip** < **n**.

NE_2_INT_ARG_LT

On entry **tdq** = $\langle value \rangle$ while **ip**+2 = $\langle value \rangle$. These parameters must satisfy **tdq** \geq **ip**+2.

NE_REAL_ARG_LT

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

NE_REAL_ARG_LE

On entry, **tol** must not be less than or equal to 0.0: **tol** = $\langle value \rangle$.

NE_NVAR_NOT_IND

The new independent variable is a linear combination of existing variables. The (**ip**+1)th column of **q** is, therefore, null.

6. Further Comments

It should be noted that the residual sum of squares produced by `nag_regsn_mult_linear_add_var` may not be correct if the model to which the new independent variable is added is not of full rank. In such a case `nag_regsn_mult_linear_upd_model` (g02ddc) should be used to calculate the residual sum of squares.

6.2. References

Draper N R and Smith H (1985) *Applied Regression Analysis* (2nd Edn) Wiley.

Golub G H and Van Loan C F (1983) *Matrix Computations* Johns Hopkins University Press, Baltimore.

Hammarling S (1985) The Singular Value Decomposition in Multivariate Statistics *ACM Signum Newsletter* **20** (3) 2–25.

McCullagh P and Nelder J A (1983) *Generalized Linear Models* Chapman and Hall.

Searle S R (1971) *Linear Models* Wiley.

7. See Also

`nag_regsn_mult_linear` (g02dac)

`nag_regsn_mult_linear_upd_model` (g02ddc)

`nag_regsn_mult_linear_delete_var` (g02dfc)

8. Example

A data set consisting of 12 observations is read in. The four independent variables are stored in the array **x** while the dependent variable is read into the first column of **q**. If the character variable **meanc** indicates that a mean should be included in the model, a variable taking the value 1.0 for all observations is set up and fitted. Subsequently, one variable at a time is selected to enter the model as indicated by the input value of **indx**. After the variable has been added the parameter estimates are calculated by `nag_regsn_mult_linear_upd_model` (g02ddc) and the results printed. This is repeated until the input value of **indx** is 0.

8.1. Program Text

```

/* nag_regsn_mult_linear_add_var(g02dec) Example Program
 *
 * Copyright 1991 Numerical Algorithms Group.
 *
 * Mark 2, 1991.
 */
#include <nag.h>
#include <stdio.h>
#include <nag_stdlib.h>
#include <nagg02.h>

#define NMAX 12
#define MMAX 5
#define TDX MMAX
#define TDQ MMAX+1

main()
{
    double rss, rsst, tol;
    Integer i, indx, ip, rank, j, m, n;
    double df;
    Boolean svd;
    char    meanc, weight;
    Nag_IncludeMean mean;
    double  b[MMAX], cov[MMAX*(MMAX+1)/2], p[MMAX*(MMAX+2)],
    q[NMAX][MMAX+1], se[MMAX], wt[NMAX], x[NMAX][MMAX], xe[NMAX];
    double  *wtptr;
    static  NagError fail;

    Vprintf("g02dec Example Program Results\n");
    /* Skip heading in data file */
    Vscanf("%*[\n]");
    Vscanf("%ld %ld %c %c", &n, &m, &weight, &meanc);
    if (meanc=='m')
        mean = Nag_MeanInclude;
    else
        mean = Nag_MeanZero;

    if (weight=='w')
        wtptr = wt;
    else
        wtptr = (double *)0;

    if (n<=NMAX && m<MMAX)
    {
        if (wtptr)
        {
            for (i=0; i<n; i++)
            {
                for (j=0; j<m; j++)
                    Vscanf("%lf", &x[i][j]);
                Vscanf("%lf%lf", &q[i][0], &wt[i]);
            }
        }
        else
        {
            for (i=0; i<n; i++)
            {
                for (j=0; j<m; j++)
                    Vscanf("%lf", &x[i][j]);
                Vscanf("%lf", &q[i][0]);
            }
        }
        /* Set tolerance */
        tol = 0.000001e0;
        ip = 0;
        if (mean==Nag_MeanInclude)
        {

```

```

    for (i = 0; i<n; ++i)
        xe[i] = 1.0;

    g02dec(n, ip, (double *)q, (Integer)(TDQ), p, wtptr, xe, &rss,
          tol, NAGERR_DEFAULT);

    ip = 1;
}
while (scanf("%ld", &indx) != EOF)
{
    if (indx > 0)
    {
        for (i=0; i<n; i++)
            xe[i] = x[i][indx-1];
        g02dec(n, ip, (double *)q, (Integer)(TDQ), p, wtptr, xe, &rss,
              tol, &fail);
        if (fail.code == NE_NOERROR)
        {
            ip += 1;
            Vprintf("Variable %4ld added\n", indx);
            rsst = 0.0;

            g02ddc(n, ip, (double *)q, (Integer)(TDQ), &rsst, &df, b, se,
                  cov, &svd, &rank, p, tol, NAGERR_DEFAULT);

            if (svd)
                Vprintf("Model not of full rank\n\n");
            Vprintf("Residual sum of squares = %13.4e\n", rsst);
            Vprintf("Degrees of freedom = %3.1f\n", df);
            Vprintf("Variable      Parameter estimate      Standard error\n\n");
            for (j=0; j<ip; j++)
                Vprintf("%6ld%20.4e%20.4e\n", j+1, b[j], se[j]);
            Vprintf("\n");
        }
        else if (fail.code == NE_NVAR_NOT_IND)
            Vprintf(" * New variable not added *\n");
        else
        {
            Vprintf("%s\n", fail.message);
            exit(EXIT_FAILURE);
        }
    }
}
}
else
{
    Vfprintf(stderr, "One or both of m and n are out of range:\n
m = %-3ld while n = %-3ld\n", m, n);
    exit(EXIT_FAILURE);
}
exit(EXIT_SUCCESS);
}

```

8.2. Program Data

g02dec Example Program Data

```

12 4 u m
1.0 1.4 0.0 0.0 4.32
1.5 2.2 0.0 0.0 5.21
2.0 4.5 0.0 0.0 6.49
2.5 6.1 0.0 0.0 7.10
3.0 7.1 0.0 0.0 7.94
3.5 7.7 0.0 0.0 8.53
4.0 8.3 1.0 4.0 8.84
4.5 8.6 1.0 4.5 9.02
5.0 8.8 1.0 5.0 9.27
5.5 9.0 1.0 5.5 9.43
6.0 9.3 1.0 6.0 9.68
6.5 9.2 1.0 6.5 9.83
1
3

```

4
2
0

8.3. Program Results

g02dec Example Program Results

Variable 1 added
Residual sum of squares = 4.0164e+00
Degrees of freedom = 10.0

Variable	Parameter estimate	Standard error
1	4.4100e+00	4.3756e-01
2	9.4979e-01	1.0599e-01

Variable 3 added
Residual sum of squares = 3.8872e+00
Degrees of freedom = 9.0

Variable	Parameter estimate	Standard error
1	4.2236e+00	5.6734e-01
2	1.0554e+00	2.2217e-01
3	-4.1962e-01	7.6695e-01

Variable 4 added
Residual sum of squares = 1.8702e-01
Degrees of freedom = 8.0

Variable	Parameter estimate	Standard error
1	2.7605e+00	1.7592e-01
2	1.7057e+00	7.3100e-02
3	4.4575e+00	4.2676e-01
4	-1.3006e+00	1.0338e-01

Variable 2 added
Residual sum of squares = 8.4066e-02
Degrees of freedom = 7.0

Variable	Parameter estimate	Standard error
1	3.1440e+00	1.8181e-01
2	9.0748e-01	2.7761e-01
3	2.0790e+00	8.6804e-01
4	-6.1589e-01	2.4530e-01
5	2.9224e-01	9.9810e-02