NAG C Library Function Document

nag_all_regsn (g02eac)

1   Purpose

nag_all_regsn (g02eac) calculates the residual sums of squares for all possible linear regressions for a given set of independent variables.

2   Specification

```
void nag_all_regsn (Nag_OrderType order, Nag_IncludeMean mean, Integer n,
    Integer m, const double x[], Integer pdx, const char *var_names[],
    const Integer sx[], const double y[], const double wt[], Integer *nmod,
    const char *model[], double rss[], Integer nterms[], Integer mrank[],
    NagError *fail)
```

3   Description

For a set of k possible independent variables there are 2^k linear regression models with from zero to k independent variables in each model. For example if $k = 3$ and the variables are A, B and C then the possible models are:

(i)   null model

(ii)   A

(iii) B

(iv)   C

(v)   A and B

(vi)   A and C

(vii) B and C

(viii) A, B and C.

nag_all_regsn (g02eac) calculates the residual sums of squares from each of the 2^k possible models. The method used involves a QR decomposition of the matrix of possible independent variables. Independent variables are then moved into and out of the model by a series of Givens rotations and the residual sums of squares computed for each model; see Clark (1981) and Smith and Bremner (1989).

The computed residual sums of squares are then ordered first by increasing number of terms in the model, then by decreasing size of residual sums of squares. So the first model will always have the largest residual sum of squares and the 2^kth will always have the smallest. This aids the user in selecting the best possible model from the given set of independent variables.

nag_all_regsn (g02eac) allows the user to specify some independent variables that must be in the model, the forced variables. The other independent variables from which the possible models are to be formed are the free variables.

4   References

Clark M R B (1981) A Givens algorithm for moving from one linear model to another without going back to the data *Appl. Statist.* **30** 198–203

Smith D M and Bremner J M (1989) All possible subset regressions using the QR decomposition *Comput. Statist. Data Anal.* **7** 217–236

Weisberg S (1985) *Applied Linear Regression* Wiley

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:    **mean** – Nag_IncludeMean   *Input*

On entry: indicates if a mean term is to be included.

If **mean = Nag_MeanInclude** (Mean), a mean term, intercept, will be included in the models.

If **mean = Nag_MeanZero** (Zero), the models will pass through the origin, zero-point.

Constraint: **mean = Nag_MeanInclude** or **Nag_MeanZero**.

3:    **n** – Integer   *Input*

On entry: the number of observations.

Constraint: $\mathbf{n} \geq 2$.

4:    **m** – Integer   *Input*

On entry: the maximum number of variables contained in **x**.

Constraint: $\mathbf{m} \geq 2$.

5:    **x**[dim] – const double   *Input*

Note: the dimension, dim, of the array **x** must be at least $\max(1, \mathbf{pdx} \times \mathbf{m})$ when **order = Nag_ColMajor** and at least $\max(1, \mathbf{pdx} \times \mathbf{n})$ when **order = Nag_RowMajor**.

Where $\mathbf{X}(i, j)$ appears in this document, it refers to the array element

if **order = Nag_ColMajor**, $\mathbf{x}[(j - 1) \times \mathbf{pdx} + i - 1]$;

if **order = Nag_RowMajor**, $\mathbf{x}[(i - 1) \times \mathbf{pdx} + j - 1]$.

On entry: $\mathbf{X}(i, j)$ must contain the ith observation for the jth independent variable, for $i = 1, 2, \ldots, \mathbf{n}$; $j = 1, 2, \ldots, \mathbf{m}$.

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

7:    **var_names**[**m**] – char *   *Input*

On entry: **var_names**$[i - 1]$ must contain the name of the independent variable in row i of **x**, for $i = 1, 2, \ldots, \mathbf{m}$.

8:    **sx**[**m**] – const Integer   *Input*

On entry: indicates which independent variables are to be considered in the model.

If $\mathbf{sx}[j - 1] \geq 2$, then the variable contained in the jth column of **x** is included in all regression models, i.e., is a forced variable.

If $\mathbf{sx}[j-1] = 1$, then the variable contained in the jth column of $\mathbf{x}$ is included in the set from which the regression models are chosen, i.e., is a free variable.

If $\mathbf{sx}[j-1] = 0$, then the variable contained in the jth column of $\mathbf{x}$ is not included in the models.

We denote the total number of free variables a k, where k is the number of free variables in the model, see **nmod** to **mrank**.

Constraint: $\mathbf{sx}[j-1] \geq 0$, for $j = 1, 2, \ldots, \mathbf{m}$ and at least one value of $\mathbf{sx} = 1$.

9: **y**[**n**] – const double   *Input*

   On entry: $\mathbf{y}[i-1]$ must contain the ith observation on the dependent variable, y_i, for $i = 1, 2, \ldots, n$.

10: **wt**[dim] – const double   *Input*

   On entry: optionally, the weights to be used in the weighted regression.

   If $\mathbf{wt}[i-1] = 0.0$, then the ith 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 weights are not provided then **wt** must be set to the **NULL** pointer, i.e., (double *)0, and the effective number of observations is **n**.

   Constraint: if **wt** is not **NULL**, $\mathbf{wt}[i] \geq 0.0$ for $i = 0, 1, \ldots, n - 1$.

11: **nmod** – Integer *   *Output*

   On exit: the total number of models for which residual sums of squares have been calculated. If there are k independent variables then $\mathbf{nmod} = 2^k$.

12: **model**[dim] – char *   *Output*

   Note: the dimension, dim, of the array **model** must be at least big enough to hold the names of all the free independent variables which appear in all the models. This will never exceed $2^k \times \mathbf{m}$, where k is the number of free variables in the model.

   On exit: the names of the independent variables in each model, represented as pointers to the names provided by the user in **var_names**. The model names are stored as follows:

   If the first model has three names i.e., $\mathbf{nterms}[0] = 3$; then $\mathbf{model}[0]$, $\mathbf{model}[1]$ and $\mathbf{model}[2]$ will contain these three names;

   If the second model has two names i.e., $\mathbf{nterms}[1] = 2$; then $\mathbf{model}[3]$, $\mathbf{model}[4]$ will contain these two names.

13: **rss**[dim] – double   *Output*

   Note: the dimension, dim, of the array **rss** must be at least $\max(2^k, \mathbf{m})$.

   On exit: $\mathbf{rss}[i-1]$ contains the residual sum of squares for the ith model, for $i = 1, 2, \ldots, \mathbf{nmod}$.

14: **nterms**[dim] – Integer   *Output*

   Note: the dimension, dim, of the array **nterms** must be at least $\max(2^k, \mathbf{m})$.

   On exit: $\mathbf{nterms}[i-1]$ contains the number of independent variables in the ith model, not including the mean if one is fitted, for $i = 1, 2, \ldots, \mathbf{nmod}$.

15: **mrank**[dim] – Integer   *Output*

   Note: the dimension, dim, of the array **mrank** must be at least $\max(2^k, \mathbf{m})$.

   On exit: $\mathbf{mrank}[i-1]$ contains the rank of the residual sum of squares for the ith model, i.e., model with smallest sum of squares has rank 1.

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

The NAG error parameter (see the Essential Introduction).

6   Error Indicators and Warnings

NE_INT

On entry, $\mathbf{n} = \langle value \rangle$.
Constraint: $\mathbf{n} \geq 2$.

On entry, $\mathbf{m} = \langle value \rangle$.
Constraint: $\mathbf{m} \geq 2$.

On entry, $\mathbf{pdx} = \langle value \rangle$.
Constraint: $\mathbf{pdx} > 0$.

NE_INT_2

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

NE_FREE_VARS

There are no free **x** variables.

NE_FULL_RANK

Full model is not of full rank.

NE_INDEP_VARS_OBS

Number of requested x-variables $\geq$ number of observations.

NE_INT_ARRAY_ELEM_CONS

On entry, $\mathbf{sx}[\langle value \rangle] < 0$.

NE_REAL_ARRAY_ELEM_CONS

On entry, $\mathbf{wt}[\langle value \rangle] < 0.0$.

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

For a discussion of the improved accuracy obtained by using a method based on the QR decomposition see Smith and Bremner (1989).

8   Further Comments

nag_cp_stat (g02ecc) may be used to compute R^2 and C_p-values from the results of nag_all_regsn (g02eac).

If a mean has been included in the model and no variables are forced in then **rss**[0] contains the total sum of squares and in many situations a reasonable estimate of the variance of the errors is given by **rss**[**nmod** − 1]/(**n** − 1 − **nterms**[**nmod** − 1]).

9   Example

The data for this example is given in Weisberg (1985). The independent variables and the dependent variable are read, as are the names of the variables. These names are as given in Weisberg (1985). The residual sums of squares computed and printed with the names of the variables in the model.

9.1   Program Text

```
/* nag_all_regsn (g02eac) Example Program.
 *
 * Copyright 2002 Numerical Algorithms Group.
 *
 * Mark 7, 2002.
 */

#include <math.h>
#include <stdio.h>
#include <string.h>
#include <nag.h>
#include <nag_stdlib.h>
#include <nagg02.h>

int main(void)
{

/* Scalars */
  Integer exit_status, free_vars, i, ii, j, m, n, nmod, pdx;
  NagError fail;
  Nag_OrderType order;

  /* Arrays */
  const char **model=0;
  const char *var_names[] = { "DAY", "BOD", "TKN", "TS", "TVS", "COD" };
  double *rss=0, *x=0, *y=0, *wtptr=0;
  Integer *sx=0, *mrank=0, *nterms=0;

  /* For iteration over model */
  Integer model_index=0;

#ifdef NAG_COLUMN_MAJOR
#define X(I,J) x[(J-1)*pdx + I - 1]
  order = Nag_ColMajor;
#else
#define X(I,J) x[(I-1)*pdx + J - 1]
  order = Nag_RowMajor;
#endif

  INIT_FAIL(fail);
  exit_status = 0;
  Vprintf("g02eac Example Program Results\n");

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

  Vscanf("%ld%ld%*[^\n] ", &n, &m);

/* Allocate memory */
  if ( !(x = NAG_ALLOC(n * m, double)) ||
       !(y = NAG_ALLOC(n, double)) ||
```

```
        !(sx = NAG_ALLOC(m, Integer)) )
     {
       Vprintf("Allocation failure\n");
       exit_status = -1;
       goto END;
     }

#ifdef NAG_COLUMN_MAJOR
  pdx = n;
  order = Nag_ColMajor;
#else
  pdx = m;
  order = Nag_RowMajor;
#endif

  for (i = 1; i <= n; ++i)
     {
       for (j = 1; j <= m; ++j)
         Vscanf("%lf", &X(i,j));
       Vscanf("%lf%*[^\n] ", &y[i - 1]);
     }

  free_vars = 1;
  for (j = 1; j <= m; ++j)
     {
       Vscanf("%ld", &sx[j - 1]);
       if (sx[j - 1] == 1)
          {
            free_vars <<= 1;
          }
     }
  Vscanf("%*[^\n] ");

  if ( !(model = NAG_ALLOC(free_vars*m, const char *)) ||
       !(rss = NAG_ALLOC(free_vars, double)) ||
       !(mrank = NAG_ALLOC(free_vars, Integer)) ||
       !(nterms = NAG_ALLOC(free_vars, Integer)) )
     {
       Vprintf("Allocation failure\n");
       exit_status = -1;
       goto END;
     }

  g02eac(order, Nag_MeanInclude, n, m, x, pdx, var_names, sx, y, wtptr,
         &nmod, model, rss, nterms, mrank, &fail);
  if (fail.code != NE_NOERROR)
     {
       Vprintf("Error from g02eac.\n%s\n", fail.message);
       exit_status = 1;
       goto END;
     }

  Vprintf("\n");
  Vprintf("Number of     Rss    Rank        Model\n");
  Vprintf("parameters\n");
  for (i = 1; i <= nmod; ++i)
     {
       ii = nterms[i - 1];
       Vprintf("%8ld%11.4f%4ld   ", ii, rss[i - 1], mrank[i - 1]);
       for (j = 1; j <= ii; ++j)
         Vprintf("%-3.3s %s", model[model_index++],
                 j%5 == 0 || j == 5 ?"\n":" ");
       Vprintf("\n");
     }
 END:
  if (rss) NAG_FREE(rss);
  if (x) NAG_FREE(x);
  if (y) NAG_FREE(y);
  if (sx) NAG_FREE(sx);
  if (mrank) NAG_FREE(mrank);
  if (nterms) NAG_FREE(nterms);
```

```
  if ( model) NAG_FREE(model);
  return exit_status;
}
```

9.2 Program Data

```
g02eac Example Program Data
  20 6
   0. 1125.0 232.0 7160.0 85.9 8905.0  1.5563
   7.  920.0 268.0 8804.0 86.5 7388.0  0.8976
  15.  835.0 271.0 8108.0 85.2 5348.0  0.7482
  22. 1000.0 237.0 6370.0 83.8 8056.0  0.7160
  29. 1150.0 192.0 6441.0 82.1 6960.0  0.3010
  37.  990.0 202.0 5154.0 79.2 5690.0  0.3617
  44.  840.0 184.0 5896.0 81.2 6932.0  0.1139
  58.  650.0 200.0 5336.0 80.6 5400.0  0.1139
  65.  640.0 180.0 5041.0 78.4 3177.0 -0.2218
  72.  583.0 165.0 5012.0 79.3 4461.0 -0.1549
  80.  570.0 151.0 4825.0 78.7 3901.0  0.0000
  86.  570.0 171.0 4391.0 78.0 5002.0  0.0000
  93.  510.0 243.0 4320.0 72.3 4665.0 -0.0969
 100.  555.0 147.0 3709.0 74.9 4642.0 -0.2218
 107.  460.0 286.0 3969.0 74.4 4840.0 -0.3979
 122.  275.0 198.0 3558.0 72.5 4479.0 -0.1549
 129.  510.0 196.0 4361.0 57.7 4200.0 -0.2218
 151.  165.0 210.0 3301.0 71.8 3410.0 -0.3979
 171.  244.0 327.0 2964.0 72.5 3360.0 -0.5229
 220.   79.0 334.0 2777.0 71.9 2599.0 -0.0458
   0    1    1    1    1    1
```

9.3 Program Results

```
g02eac Example Program Results
```

Number of parameters	Rss	Rank	Model				
0	5.0634	32					
1	5.0219	31	TKN				
1	2.5044	30	TVS				
1	2.0338	28	BOD				
1	1.5563	25	COD				
1	1.5370	24	TS				
2	2.4381	29	TKN	TVS			
2	1.7462	27	BOD	TVS			
2	1.5921	26	BOD	TKN			
2	1.4963	23	BOD	COD			
2	1.4707	22	TKN	TS			
2	1.4590	21	TS	TVS			
2	1.4397	20	BOD	TS			
2	1.4388	19	TKN	COD			
2	1.3287	15	TVS	COD			
2	1.0850	8	TS	COD			
3	1.4257	18	BOD	TKN	TVS		
3	1.3900	17	TKN	TS	TVS		
3	1.3894	16	BOD	TS	TVS		
3	1.3204	14	BOD	TVS	COD		
3	1.2764	13	BOD	TKN	COD		
3	1.2582	12	BOD	TKN	TS		
3	1.2179	10	TKN	TVS	COD		
3	1.0644	7	BOD	TS	COD		
3	1.0634	6	TS	TVS	COD		
3	0.9871	4	TKN	TS	COD		
4	1.2199	11	BOD	TKN	TS	TVS	
4	1.1565	9	BOD	TKN	TVS	COD	
4	1.0388	5	BOD	TS	TVS	COD	
4	0.9871	3	BOD	TKN	TS	COD	
4	0.9653	2	TKN	TS	TVS	COD	
5	0.9652	1	BOD	TKN	TS	TVS	COD