nag_regsn_std_resid_influence (g02fac)

1.   Purpose

nag_regsn_std_resid_influence (g02fac) calculates two types of standardised residuals and two measures of influence for a linear regression.

2.   Specification

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

void nag_regsn_std_resid_influence(Integer n, Integer ip, Integer nres,
    double res[], double h[], double rms, double sres[], NagError *fail)
```

3.   Description

For the general linear regression model is defined by

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

where y is a vector of length n of the dependent variable,
       X is an n by p matrix of the independent variables,
       β is a vector of length p of unknown parameters,
and ε is a vector of length n of unknown random errors such that $\operatorname{var}\varepsilon = \sigma^2 I$.

The residuals are given by

$$r = y - \hat{y} = y - X\hat{\beta}.$$

The fitted values, $\hat{y} = X\hat{\beta}$, can be written as Hy for an n by n matrix H. The ith diagonal element of H, h_i, gives a measure of the influence of the ith value of the independent variables on the fitted regression model. The values of r and the h_i are returned by nag_regsn_mult_linear (g02dac).

nag_regsn_std_resid_influence calculates statistics which help to indicate if an observation is extreme and having an undue influence on the fit of the regression model. Two types of standardised residual are calculated

(a)   The ith residual is standardised by its variance when the estimate of σ^2, s^2, is calculated from all the data; known as internal studentization.

$$RI_i = \frac{r_i}{s\sqrt{1 - h_i}}.$$

(b)   The ith residual is standardised by its variance when the estimate of σ^2, s^2_{-i} is calculated from the data excluding the ith observation; known as external studentization.

$$RE_i = \frac{r_i}{s_{-i}\sqrt{1 - h_i}} = r_i\sqrt{\frac{n - p - 1}{n - p - RI_i^2}}.$$

The two measures of influence are:

(a)   Cook's D

$$D_i = \frac{1}{p}RE_i^2\frac{h_i}{1 - h_i}$$

(b)   Atkinson's T

$$T_i = |RE_i|\sqrt{\left(\frac{n - p}{p}\right)\left(\frac{h_i}{1 - h_i}\right)}.$$

4.   Parameters

n

Input: number of observations included in the regression, n.
Constraint: $\mathbf{n} > \mathbf{ip} + 1$.

ip

Input: the number of linear parameters estimated in the regression model, p.
Constraint: $\mathbf{ip} \geq 1$.

nres

Input: the number of residuals.
Constraint: $1 \leq \mathbf{nres} \leq \mathbf{n}$.

res[nres]

Input: the residuals, r_i.

h[nres]

Input: the diagonal elements of H, h_i, corresponding to the residuals in **res**.
Constraint: $0.0 < \mathbf{h}[i] < 1.0$, for $i = 0, 1, \ldots,\mathbf{nres} - 1$.

rms

Input: the estimate of σ^2 based on all n observations, s^2, i.e., the residual mean square.
Constraint: $\mathbf{rms} > 0.0$.

sres[nres][4]

Output: the standardised residuals and influence statistics.
For the observation with residual given in **res**[i]:

      sres[i][0] is the internally studentized residual
      sres[i][1] is the externally studentized residual
      sres[i][2] is Cook's D statistic
      sres[i][3] is Atkinson's T statistic.

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

NE_2_INT_ARG_LE

On entry, **n** = $\langle value \rangle$ while **ip** + 1 = $\langle value \rangle$. These parameters must satisfy $\mathbf{n} > \mathbf{ip} + 1$.

NE_2_INT_ARG_LT

On entry, **nres** = $\langle value \rangle$ while **n** = $\langle value \rangle$. These parameters must satisfy $\mathbf{nres} \geq \mathbf{n}$.

NE_REAL_ARG_LE

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

NE_REAL_ARG_GE

On entry, **h**[$\langle value \rangle$] must not be greater than or equal to 1.0: **h**[$\langle value \rangle$] = $\langle value \rangle$.

NE_RESID_LARG

On entry, the value of a residual is too large for the given value of **rms**: **res**[$\langle value \rangle$] = $\langle value \rangle$,
rms = $\langle value \rangle$.

6.   Further Comments

6.1. Accuracy

Accuracy is sufficient for all practical purposes.

6.2. References

Atkinson A C (1981) Two Graphical displays for outlying and influential observations in regression
Biometrika **68** 13–20.
Cook R D and Weisberg S (1982) *Residuals and Influence in Regression* Chapman and Hall.

7.   See Also

nag_regsn_mult_linear (g02dac)

8.   Example

A set of 24 residuals and h_i values from an 11 parameter model fitted to the cloud seeding data
considered in Cook and Weisberg (1982) are input and the standardised residuals etc calculated
and printed for the first 10 observations.

8.1.   Program Text

```
/* nag_regsn_std_resid_influence(g02fac) Example Program
 *
 * Copyright 1990 Numerical Algorithms Group.
 *
 * Mark 1, 1990.
 */
#include <nag.h>
#include <stdio.h>
#include <nag_stdlib.h>
#include <nagg02.h>

#define NMAX 24

main()
{
  double rms;
  Integer i, ip, j, n, nres;
  double h[NMAX], res[NMAX], sres[NMAX][4];

  Vprintf("g02fac Example Program Results\n");
  /*  Skip heading in data file */
  Vscanf("%*[^\n]");
  Vscanf("%ld %ld %ld %lf", &n, &ip, &nres, &rms);
  if (nres<NMAX)
    {
      for (i=0; i<nres; i++)
        Vscanf("%lf%lf",&res[i],&h[i]);
      g02fac(n, ip, nres, res, h, rms, (double *)sres, NAGERR_DEFAULT);
      Vprintf("\n");
      Vprintf("          Internally     Externally\n");
      Vprintf(" Obs.    studentized    studentized    Cook's D    Atkinson's T\
\n");
      Vprintf("          residuals      residuals\n\n");
      for (i=0; i<nres; i++)
        {
          Vprintf("%2ld", i+1);
          for (j=0; j<4; j++)
            Vprintf("%14.3f", sres[i][j]);
          Vprintf("\n");
        }
      exit(EXIT_SUCCESS);
    }
  else
    {
      Vfprintf(stderr, "nres is out of range: nres = %5ld\n", nres);
      exit(EXIT_FAILURE);
    }
}
```

8.2. Program Data

```
g02fac Example Program Data
 24 11 10 .5798
      0.2660         0.5519
     -0.1387         0.9746
     -0.2971         0.6256
      0.5926         0.3144
     -0.4013         0.4106
      0.1396         0.6268
     -1.3173         0.5479
      1.1226         0.2325
      0.0321         0.4115
     -0.7111         0.3577
      0.3439         0.3342
     -0.4379         0.1673
      0.0633         0.3874
     -0.0936         0.1705
      0.9968         0.3466
      0.0209         0.3743
     -0.4056         0.7527
      0.1396         0.9069
      0.0327         0.2610
      0.2970         0.6256
     -0.2277         0.2485
      0.5180         0.3072
      0.5301         0.5848
     -1.0650         0.4794
```

8.3. Program Results

```
g02fac Example Program Results
```

Obs.	Internally studentized residuals	Externally studentized residuals	Cook's D	Atkinson's T
1	0.522	0.507	0.030	0.611
2	-1.143	-1.158	4.557	-7.797
3	-0.638	-0.622	0.062	-0.875
4	0.940	0.935	0.037	0.689
5	-0.686	-0.672	0.030	-0.610
6	0.300	0.289	0.014	0.408
7	-2.573	-3.529	0.729	-4.223
8	1.683	1.828	0.078	1.094
9	0.055	0.053	0.000	0.048
10	-1.165	-1.183	0.069	-0.960