

nag_simple_linear_regression (g02cac)

1. Purpose

nag_simple_linear_regression (g02cac) performs a simple linear regression with or without a constant term. The data is optionally weighted.

2. Specification

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

void nag_simple_linear_regression(Nag_SumSquare mean, Integer n,
    double x[], double y[], double wt[], double *a, double *b, double *a_serr,
    double *b_serr, double *rsq, double *rss, double *df, NagError *fail)
```

3. Description

This function fits a straight line model of the form,

$$E(y) = a + bx,$$

where $E(y)$ is the expected value of the variable y , to the data points

$$(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n),$$

such that

$$y_i = a + bx_i + e_i, i = 1, 2, \dots, n (n > 2).$$

where the e_i values are independent random errors. The i th data point may have an associated weight w_i , these may be used either in the situation when $\text{var}(\varepsilon_i) = \sigma^2/w_i$ or if observations have to be removed from the regression by having zero weight or have been observed with frequency w_i .

The regression coefficient, b , and the regression constant, a are estimated by minimizing

$$\sum_{i=1}^n w_i e_i^2,$$

if the weights option is not selected then $w_i = 1.0$.

The following statistics are computed:

the estimate of regression constant $\hat{a} = \bar{y} - \hat{b}\bar{x}$,

the estimate of regression coefficient $\hat{b} = \frac{\sum w_i (x_i - \bar{x})(y_i - \bar{y})}{\sum w_i (x_i - \bar{x})^2}$,

the residual sum of squares $rss = \sum w_i (y_i - \hat{y}_i)^2$,

where the weighted means \bar{x} and \bar{y} are

$$\bar{x} = \frac{\sum w_i x_i}{\sum w_i} \quad \text{and} \quad \bar{y} = \frac{\sum w_i y_i}{\sum w_i}.$$

The number of degrees of freedom associated with rss is

$$\begin{aligned} df &= \sum w_i - 2 \quad \text{where } \mathbf{mean} = \mathbf{Nag_AboutMean} \\ df &= \sum w_i - 1 \quad \text{where } \mathbf{mean} = \mathbf{Nag_AboutZero}. \end{aligned}$$

Note: the weights should be scaled to give the correct degrees of freedom in the case $\text{var}(\varepsilon_i) = \sigma^2/w_i$.

The R^2 value or coefficient of determination

$$R^2 = \frac{\sum w_i (\hat{y}_i - \bar{y})^2}{\sum w_i (y_i - \bar{y})^2} = \frac{\sum w_i (y_i - \bar{y})^2 - rss}{\sum w_i (y_i - \bar{y})^2}.$$

This measures the proportion of the total variation about the mean \bar{y} that can be explained by the regression.

The standard error for the regression constant \hat{a}

$$a_serr = \sqrt{\frac{rss}{df} \left(\frac{1}{\sum w_i} + \frac{(\bar{x})^2}{\sum w_i(x_i - \bar{x})^2} \right)} = \sqrt{\frac{rss}{df} \frac{1}{\sum w_i} \frac{\sum w_i x_i^2}{\sum w_i(x_i - \bar{x})^2}}.$$

The standard error for the regression coefficient \hat{b}

$$b_serr = \sqrt{\frac{rss}{df \sum w_i(x_i - \bar{x})^2}}.$$

Similar formulae can be derived for the case when the line goes through the origin, that is $a = 0$.

4. Parameters

mean

Input: indicates whether nag_simple_linear_regression is to include a constant term in the regression.

If **mean** = **Nag_AboutMean**, the regression constant a is included.

If **mean** = **Nag_AboutZero**, the regression constant a is not included, i.e., $a = 0$

Constraint: **mean** = **Nag_AboutMean** or **Nag_AboutZero**.

n

Input: the number of observations, n .

Constraint: if **mean** = **Nag_AboutMean** $n \geq 2$. If **mean** = **Nag_AboutZero** $n \geq 1$.

x[n]

Input: the values of the independent variable with the i th value stored in $x[i - 1]$ for $i = 1, \dots, n$.

Constraint: all the values of x must not be identical.

y[n]

Input: the values of the dependent variable with the i th value stored in $y[i - 1]$ for $i = 1, \dots, n$.

Constraint: all the values of y must not be identical.

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.

Usually **wt**[$i - 1$] will be an integral value corresponding to the number of observations associated with the i th data point, or zero if the i th data point is to be ignored. The sum of the weights therefore represents the effective total number of observations used to create the regression line.

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

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

a

Output: If **mean** = **Nag_AboutMean** then **a** is the regression constant \hat{a} , otherwise **a** is set to zero.

b

Output: the regression coefficient \hat{b} .

a_serr

Output: the standard error of the regression constant \hat{a} .

b_serr

Output: the standard error of the regression coefficient \hat{b} .

rsq

Output: the coefficient of determination, R^2 .

rss

Output: the sum of squares of the residuals about the regression.

df

Output: the degrees of freedom associated with the residual sum of squares.

fail

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

5. Error Indications and Warnings

NE_BAD_PARAM

On entry, parameter **mean** had an illegal value.

NE_INT_ARG_LT

On entry, **n** must not be less than 1: **n** = $\langle value \rangle$
if **mean** = **Nag_AboutZero**.On entry, **n** must not be less than 2: **n** = $\langle value \rangle$
if **mean** = **Nag_AboutMean**.

NE_NEG_WEIGHT

On entry, at least one of the weights is negative.

NE_WT_LOW

On entry, **wt** must contain at least 1 positive element if **mean** = **Nag_AboutZero** or at least 2 positive elements if **mean** = **Nag_AboutMean**.

NE_X_OR_Y_IDEN

On entry, all elements of **x** and/or **y** are equal.

NE_SW_LOW

On entry, the sum of elements of **wt** must be greater than 1.0 if **mean** = **Nag_AboutZero** or greater than 2.0 if **mean** = **Nag_AboutMean**.

NE_ZERO_DOF_RESID

On entry, the degrees of freedom for the residual are zero, i.e., the designated number of parameters = the effective number of observations.

NW_RSS_EQ_ZERO

Residual sum of squares is zero, i.e., a perfect fit was obtained.

6. Further Comments

The time taken by the function depends on n .

The function uses a two-pass algorithm.

6.1. Accuracy

The computations are believed to be stable.

6.2. References

Draper N R and Smith H (1981) *Applied Regression Analysis*. (2nd Edn) Wiley.

7. See Also

nag_regress_confid_interval (g02cbc)

8. Example

A program to calculate regression constants, \hat{a} and \hat{b} , the standard error of the regression constants, the regression coefficient of determination and the degrees of freedom about the regression.

8.1. Program Text

```

/* nag_simple_linear_regression(g02cac) Example Program
 *
 * Copyright 1994 Numerical Algorithms Group.
 *
 * Mark 3, 1994.
 */

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

#define NMAX 10
main()
{
    Nag_SumSquare mean;
    char m, w;
    Integer i, n;
    double x[NMAX], y[NMAX], wt[NMAX];
    double a, b, err_a, err_b, rsq, rss, df;
    double *wtptr;

    Vprintf("g02cac Example Program Results\n");
    /* Skip heading in data file */
    Vscanf("%*[^\\n]");
    Vscanf(" %c %c",&m, &w);
    Vscanf("%ld", &n);
    if (n>=1 && n<=NMAX)
    {
        if (m == 'M' || m == 'm')
            mean = Nag_AboutMean;
        else
            mean = Nag_AboutZero;
        if (w == 'W' || w == 'w')
        {
            wtptr = wt;
            for(i = 0; i < n; ++i)
                Vscanf("%lf%lf%lf", &x[i], &y[i], &wt[i]);
        }
        else
        {
            wtptr = (double *)0;
            for(i = 0; i < n; ++i)
                Vscanf("%lf%lf", &x[i], &y[i]);
        }

        g02cac(mean, n, x, y, wtptr, &a, &b, &err_a, &err_b, &rsq, &rss,
                &df, NAGERR_DEFAULT);

        if (mean == Nag_AboutMean)
        {
            Vprintf("\nRegression constant a = %6.4f\n\n", a);
            Vprintf("Standard error of the regression constant a = %6.4f\n\n",
                    err_a);
        }

        Vprintf("Regression coefficient b = %6.4f\n\n", b);
        Vprintf("Standard error of the regression coefficient b = %6.4f\n\n",
                err_b);

        Vprintf("The regression coefficient of determination = %6.4f\n\n", rsq);
        Vprintf("The sum of squares of the residuals about the \
regression = %6.4f\n\n", rss);
        Vprintf("Number of degrees of freedom about the \
regression = %6.4f\n\n",df);
    }
    else

```

```
    {  
        Vfprintf(stderr, "n is out of range:\n  
n = %-3ld\n",n);  
        exit(EXIT_FAILURE);  
    }  
    exit(EXIT_SUCCESS);  
}
```

8.2. Program Data

```
g02cac Example Program Data  
m w  
8  
1.0 20.0 1.0  
0.0 15.5 1.0  
4.0 28.3 1.0  
7.5 45.0 1.0  
2.5 24.5 1.0  
0.0 10.0 1.0  
10.0 99.0 1.0  
5.0 31.2 1.0
```

8.3. Program Results

```
g02cac Example Program Results
```

```
Regression constant a = 7.5982
```

```
Standard error of the regression constant a = 6.6858
```

```
Regression coefficient b = 7.0905
```

```
Standard error of the regression coefficient b = 1.3224
```

```
The regression coefficient of determination = 0.8273
```

```
The sum of squares of the residuals about the regression = 965.2454
```

```
Number of degrees of freedom about the regression = 6.0000
```
