

## A GENERAL SAS MACRO FOR PERFORMING WEIGHTED LEAST SQUARES

Wanda H. Burton, Medical College of Virginia

The SAS procedure GLM provides an excellent means for performing least squares regression analysis when the usual model assumptions can be made. The model referred to is

$$Y = X\beta + \epsilon,$$

where  $\epsilon$  is normally distributed with mean 0 and variance  $\sigma^2 I$ .

The case to be considered here is that in which the observations remain independent, but their variances are not all equal. The form of the variance-covariance matrix is  $\sigma^2 V$  where  $V$  is a diagonal matrix with unequal diagonal elements,

$$\sigma^2 V = \begin{bmatrix} \sigma_1^2 & & & 0 \\ & \sigma_2^2 & & \\ & & \ddots & \\ 0 & & & \sigma_n^2 \end{bmatrix}$$

A unique nonsingular symmetric matrix  $P$  can be found such that

$$P^2 = V.$$

A transformation can then be made on our original model by premultiplying by  $P^{-1}$ , obtaining

$$P^{-1}Y = P^{-1}X\beta + P^{-1}\epsilon$$

or

$$Z = Q\beta + f$$

with obvious substitutions. This model satisfies the necessary assumptions for carrying out the usual least squares regression analysis; that is,  $f \sim N(0, \sigma^2 I)$ .

The MACRO to be presented provides a thorough analysis for a simple linear regression model,

$$E(y_i) = \beta_0 + \beta_1 x_i.$$

Let us use the following notation for the variance of  $Y$ :

$$\text{Var}(Y) = \sigma^2 V = \sigma^2 \begin{bmatrix} w_1 & & & 0 \\ & w_2 & & \\ & & \ddots & \\ 0 & & & w_n \end{bmatrix}$$

where the  $w$ 's are known weights.

For this situation,

$$P^{-1} = \begin{bmatrix} 1/\sqrt{w_1} & & & 0 \\ & 1/\sqrt{w_2} & & \\ & & \ddots & \\ 0 & & & 1/\sqrt{w_n} \end{bmatrix}$$

and a simple transformation of the variables is appropriate. These calculations are carried out in the MACRO WT REGR in statements 4-10, creating the variables  $Z, Q_0$  and  $Q_1$ .

$$Y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}, \quad P^{-1}Y = Z = \begin{bmatrix} y_1/\sqrt{w_1} \\ y_2/\sqrt{w_2} \\ \vdots \\ y_n/\sqrt{w_n} \end{bmatrix}$$

$$X = \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ \vdots & \vdots \\ 1 & x_n \end{bmatrix}$$

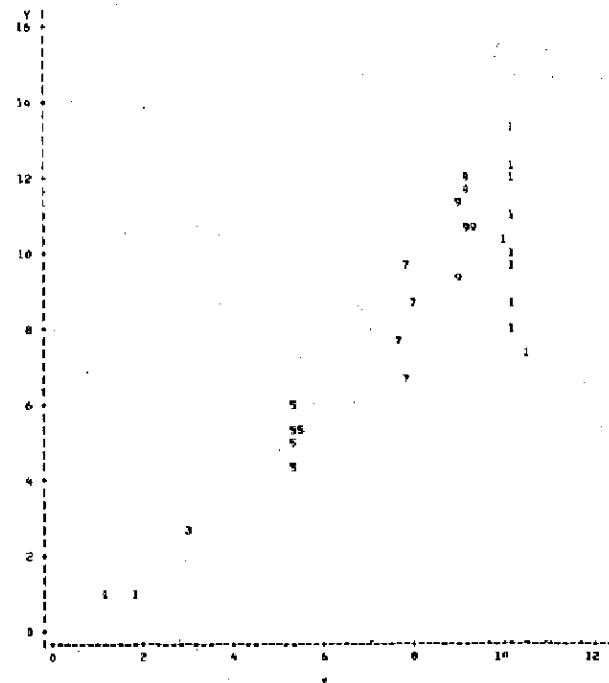
$$P^{-1}X = Q = (Q_0, Q_1) = \begin{bmatrix} 1/\sqrt{w_1} & x_1/\sqrt{w_1} \\ 1/\sqrt{w_2} & x_2/\sqrt{w_2} \\ \vdots & \vdots \\ 1/\sqrt{w_n} & x_n/\sqrt{w_n} \end{bmatrix}$$

These transformations are made after the regression variables,  $X$  and  $Y$ , and the weights,  $WT$ , are input separately and merged in statements 27-65. An identification variable,  $ID$ , is included in each of the input data sets and is necessary for the merger.

PROC GLM is used to compute the ordinary least squares regression of  $Z$  on  $Q_0$  and  $Q_1$ . Notice that the MODEL statement is written with NOINT option because the vector of ones normally present in a simple linear regression model has been transformed to a non-constant vector,  $Q_0$ . The OUTPUT facilities are used for access to the predicted values and the residuals from the fitted line.

The MACRO WT\_REGR includes a scatter of the original dependent variable (Y) against the independent variable (X). It provides plots which allow the user to examine the residuals from the analysis and thereby judge the effectiveness of applying the techniques of weighted regression to the data. A listing of the original variables with the scaled predicted values and residuals is included. This transformation back to the original scale of the data is carried out in statements 16-19.

#### STATISTICAL ANALYSIS S PLOT OF X<sup>W</sup>



#### Acknowledgment

The author would like to thank Maria Sauer for the use of laboratory data from the Pulmonary Division of the Department of Medicine, Medical College of Virginia. This work was supported in part by grant 1 R01 O1899-02 from the National Center for Health Services Research, HRA.

#### Reference

Draper, N. R. and Smith, H. Applied Regression Analysis, John Wiley & Sons, Inc., 1966, pp. 77-81.

#### STATISTICAL ANALYSIS S NOTE: THE JOB SORNTS HAS BEEN RUN UNDER RELEASE 76.2 OF SAS AT VIRGINIA COMMONWEALT

```

1  MACRO WT_REGR
2  PROC SORT DATA=ORIG1 BY ID;
3  PROC SORT DATA=WEIGHTS1 BY ID;
4  DATA TRANS;
5  MERGE ORIG1 WEIGHTS1;
6  BY ID;
7  WHERE ID>1;
8  X=Y/W;
9  O_1=Y/W;
10  O_1=Y/W;
11  PROC SCATTER;
12  PLOT X*Y*ID / NPOS=751;
13  PROC GLM;
14  MODEL Y=O_0 O_1 / NOINT SSL B;
15  OUTP=OUTPRED PREDICTED=Y_HAT RESIDUAL=Y_RESID;
16  DATA BACK1;
17  SET BACK1;
18  Y_HAT=Y**2_HAT;
19  Y_RESID=Y_HAT;
20  PROC PRINTN;
21  VAR ID Y X Y_HAT Y_RESID;
22  PROC SCATTER;
23  PLOT X*Y_RESID*ID / NPOS=751;
24  PROC SCATTER;
25  PLOT X*Y_RESID*ID / NPOS=751;
26  ;
27  DATA ORIG1 INPUT X 1-5 Y 6-10 ID 11-12;
28  CARDS;
NOTE: DATA SET WORK.ORIG1 HAS 35 OBSERVATIONS AND 3 VARIABLES.
NOTE: THE DATA STATEMENT USED 0.41 SECONDS AND 96K.
64  DATA WEIGHTS1 INPUT ID Y;
65  CARDS;
NOTE: DATA SET WORK.WEIGHTS1 HAS 6 OBSERVATIONS AND 2 VARIABLES.
NOTE: THE DATA STATEMENT USED 0.14 SECONDS AND 96K.
72  ;
73  WT_REGR
NOTE: DATA SET WORK.TRANS HAS 35 OBSERVATIONS AND 3 VARIABLES.
NOTE: THE PROCEDURE SORT USED 1.34 SECONDS AND 110K.
NOTE: DATA SET WORK.WEIGHTS HAS 6 OBSERVATIONS AND 2 VARIABLES.
NOTE: THE PROCEDURE SORT USED 1.38 SECONDS AND 110K.
NOTE: DATA SET WORK.TRANS HAS 35 OBSERVATIONS AND 3 VARIABLES.
NOTE: THE DATA STATEMENT USED 0.27 SECONDS AND 104K.

```

OBG	ID	X	Y	Y_HAT	Y_RESID
1	1	1	1.15	0.99	0.5738
2	1	1	1.98	0.98	1.4354
3	3	9	3.08	2.60	2.4990
4	3	9	3.03	2.67	2.4990
5	3	9	3.00	2.65	2.4990
6	3	9	3.00	2.74	2.4990
7	9	9	9.00	2.85	2.4990
8	5	25	5.24	5.97	5.3872
9	5	25	5.25	5.95	5.3872
10	5	25	5.42	5.33	5.4581
11	5	25	5.40	5.89	5.4581
12	6	25	5.45	5.21	5.5135
13	7	49	7.70	7.68	8.0963
14	7	49	7.88	9.81	8.2132
15	7	49	7.81	6.57	8.2846
16	7	49	7.85	9.71	8.2785
17	7	49	7.87	9.82	8.2938
18	7	49	7.91	9.81	8.3295
19	7	49	7.94	9.81	8.3405
20	9	81	9.03	9.41	9.4981
21	9	81	9.07	11.46	9.6721
22	9	81	9.11	12.14	9.7180
23	9	81	9.15	11.50	9.7525
24	9	81	9.18	10.65	9.7755
25	9	81	9.37	10.64	10.0167
26	10	100	10.17	9.74	10.9357
27	10	100	10.18	12.39	10.9478
28	10	100	10.23	11.03	10.9932
29	10	100	10.23	10.93	10.9932
30	10	100	10.22	11.90	10.9932
31	10	100	10.18	8.68	10.9478
32	10	100	10.50	7.25	11.3168
33	10	100	10.23	13.66	11.0647
34	10	100	10.63	10.19	10.7749
35	10	100	10.23	9.94	11.0047
36	10	100	10.23	9.94	11.0047

