G08RBF – NAG Fortran Library Routine Document

Note. Before using this routine, please read the Users' Note for your implementation to check the interpretation of bold italicised terms and other implementation-dependent details.

1 Purpose

G08RBF calculates the parameter estimates, score statistics and their variance-covariance matrices for the linear model using a likelihood based on the ranks of the observations when some of the observations may be right-censored.

2 Specification

SUBROUTINE	GO8RBF(NS, NV, NSUM, Y, IP, X, NX, ICEN, GAMMA, NMAX,
1	TOL, PARVAR, NPVAR, IRANK, ZIN, ETA, VAPVEC,
2	PAREST, WORK, LWORK, IWA, IFAIL)
INTEGER	NS, NV(NS), NSUM, IP, NX, ICEN(NSUM), NMAX,
1	NPVAR, IRANK(NMAX), LWORK, IWA(4*NMAX), IFAIL
real	Y(NSUM), X(NX,IP), GAMMA, TOL, PARVAR(NPVAR,IP),
1	ZIN(NMAX), ETA(NMAX), VAPVEC(NMAX*(NMAX+1)/2),
2	PAREST(4*IP+1), WORK(LWORK)

3 Description

Analysis of data can be made by replacing observations by their ranks. The analysis produces inference for the regression model where the location parameters of the observations, θ_i , $i = 1, 2, \ldots, n$, are related by $\theta = X\beta$. Here X is an n by p matrix of explanatory variables and β is a vector of p unknown regression parameters. The observations are replaced by their ranks and an approximation, based on Taylor's series expansion, made to the rank marginal likelihood. For details of the approximation see Pettitt [2].

An observation is said to be right-censored if we can only observe Y_j^* with $Y_j^* \leq Y_j$. We rank censored and uncensored observations as follows. Suppose we can observe Y_j , for j = 1, 2, ..., n, directly but Y_j^* , for j = n+1, n+2, ..., q; $n \leq q$, are censored on the right. We define the rank r_j of Y_j , for j = 1, 2, ..., n, in the usual way; r_j equals *i* if and only if Y_j is the *i*th smallest amongst the Y_1, Y_2, \ldots, Y_n . The right-censored Y_j^* , for $j = n + 1, n + 2, \ldots, q$ has rank r_j if and only if Y_j^* lies in the interval $[Y_{(r_j)}, Y_{(r_j+1)}]$, with $Y_0 = -\infty$, $Y_{(n+1)} = +\infty$ and $Y_{(1)} < \ldots < Y_{(n)}$ the ordered Y_j , for $j = 1, 2, \ldots, n$.

The distribution of the Y's is assumed to be of the following form. Let $F_L(y) = e^y/(1+e^y)$, the logistic distribution function, and consider the distribution function $F_{\gamma}(y)$ defined by $1 - F_{\gamma} = [1 - F_L(y)]^{1/\gamma}$. This distribution function can be thought of as either the distribution function of the minimum, $X_{1,\gamma}$, of a random sample of size γ^{-1} from the logistic distribution, or as the $F_{\gamma}(y - \log \gamma)$ being the distribution function of a random variable having the *F*-distribution with 2 and $2\gamma^{-1}$ degrees of freedom. This family of generalized logistic distribution functions $[F_{\gamma}(.); 0 \leq \gamma < \infty]$ naturally links the symmetric logistic distribution ($\gamma = 1$) with the skew extreme value distribution ($\lim \gamma \to 0$) and with the limiting negative exponential distribution (lim $\gamma \to \infty$). For this family explicit results are available for right-censored data. See Pettitt [3] for details.

Let l_R denote the logarithm of the rank marginal likelihood of the observations and define the $q \times 1$ vector a by $a = l'_{R_{\mu}}(\theta = 0)$, and let the q by q diagonal matrix B and q by q symmetric matrix A be given by $B - A = -l_R^{''}$ ($\theta = 0$). Then various statistics can be found from the analysis.

- (a) The score statistic $X^T a$. This statistic is used to test the hypothesis $H_0: \beta = 0$ (see (e)).
- (b) The estimated variance-covariance matrix of the score statistic in (a).
- (c) The estimate $\hat{\beta}_R = M X^T a$.
- (d) The estimated variance-covariance matrix $M = (X^T(B-A)X)^{-1}$ of the estimate $\hat{\beta}_R$. (e) The χ^2 statistic $Q = \hat{\beta}_R M^{-1} \hat{\beta}_r = a^T X (X^T(B-A)X)^{-1} X^T a$, used to test $H_0 : \beta = 0$. Under H_0 , Q has an approximate χ^2 distribution with p degrees of freedom.
- (f) The standard errors $M_{ii}^{\frac{1}{2}}$ of the estimates given in (c).

(g) Approximate z-statistics, i.e., $Z_i = \hat{\beta}_{R_i}/s.e.(\hat{\beta}_{R_i})$ for testing $H_0: \beta_i = 0$. For $i = 1, 2, ..., n, Z_i$ has an approximate N(0, 1) distribution.

In many situations, more than one sample of observations will be available. In this case we assume the model,

$$h_k(Y_k) = X_k^T \beta + e_k, \quad k = 1, 2, \dots, \text{NS},$$

where NS is the number of samples. In an obvious manner, Y_k and X_k are the vector of observations and the design matrix for the kth sample respectively. Note that the arbitrary transformation h_k can be assumed different for each sample since observations are ranked within the sample.

The earlier analysis can be extended to give a combined estimate of β as $\hat{\beta} = Dd$, where

$$D^{-1} = \sum_{k=1}^{\rm NS} X^T (B_k - A_k) X_k$$

and

$$d = \sum_{k=1}^{\rm NS} X_k^T a_k$$

with a_k , B_k and A_k defined as a, B and A above but for the kth sample.

The remaining statistics are calculated as for the one sample case.

4 References

- [1] Kalbfleisch J D and Prentice R L (1980) The Statistical Analysis of Failure Time Data Wiley
- [2] Pettitt A N (1982) Inference for the linear model using a likelihood based on ranks J. Roy. Statist. Soc. Ser. B 44 234–243
- [3] Pettitt A N (1983) Approximate methods using ranks for regression with censored data *Biometrika* 70 121–132

5 Parameters

1:	NS — INTEGER	Input
	On entry: the number of samples.	
	Constraint: $NS \ge 1$.	
2:	NV(NS) — INTEGER array	Input
	On entry: the number of observations in the <i>i</i> th sample, for $i = 1, 2, \ldots$, NS.	
	Constraint: $NV(i) \ge 1$, for $i = 1, 2, \dots, NS$.	
3:	NSUM — INTEGER	Input
	On entry: the total number of observations.	
	Constraint: NSUM = $\sum_{i=1}^{NS} NV(i)$.	
4:	Y(NSUM) - real array	Input
	On entry: the observations in each sample. Specifically, $Y\left(\sum_{k=1}^{i-1} NV(k) + j\right)$ must contain	the j th
	observation in the <i>i</i> th sample.	
5:	IP - INTEGER	Input

On entry: the number of parameters to be fitted. Constraint: IP ≥ 1 .

X(NX,IP) - real array

6:

On entry: the design matrices for each sample. Specifically, $X\left(\sum_{k=1}^{l} NV(k) + j, l\right)$ must contain the value of the lth explanatory variable for the jth observations in the ith sample.

Constraint: X must not contain a column with all elements equal.

7: NX — INTEGER

On entry: the first dimension of the array X as declared in the (sub)program from which G08RBF is called.

Constraint: $NX \ge NSUM$.

ICEN(NSUM) — INTEGER array 8:

On entry: defines the censoring variable for the observations in Y as follows:

ICEN(i) = 0 if Y(i) is uncensored.

ICEN(i) = 1 if Y(i) is censored.

Constraint: ICEN(i) = 0 or 1, for $i = 1, 2, \dots, NSUM$.

GAMMA - real9:

> On entry: the value of the parameter defining the generalized logistic distribution. For GAMMA \leq 0.0001, the limiting extreme value distribution is assumed.

Constraint: GAMMA > 0.0.

10: NMAX — INTEGER

On entry: the value of the largest sample size.

Constraint: NMAX = $\max_{1 \le i \le NS}(NV(i))$ and NMAX > IP.

11: TOL - real

On entry: the tolerance for judging whether two observations are tied. Thus, observations Y_i and Y_j are adjudged to be tied if $|Y_i - Y_j| < \text{TOL}$.

Constraint: TOL > 0.0.

12: PARVAR(NPVAR,IP) — *real* array

On exit: the variance-covariance matrices of the score statistics and the parameter estimates; the former being stored in the upper triangle and the latter in the lower triangle. Thus for $1 \le i \le j \le j$ IP, PARVAR(i, j) contains an estimate of the covariance between the *i*th and *j*th score statistics. For $1 \le j \le i \le IP-1$, PARVAR(i+1,j) contains an estimate of the covariance between the *i*th and jth parameter estimates.

13: NPVAR — INTEGER

On entry: the first dimension of the array PARVAR as declared in the (sub)program from which G08RBF is called.

Constraint: NPVAR \geq IP + 1.

14: IRANK(NMAX) — INTEGER array

On exit: for the one sample case, IRANK contains the ranks of the observations.

15: ZIN(NMAX) — *real* array

On exit: for the one sample case, ZIN contains the expected values of the function g(.) of the order statistics.

Input

Input

Input

Input

Input

Output

Output

Input

Output

Input

G08RBF.3

Output

Output

16: ETA(NMAX) - real array

On exit: for the one sample case, ETA contains the expected values of the function g'(.) of the order statistics.

17: VAPVEC(NMAX*(NMAX+1)/2) - real array Output

On exit: for the one sample case, VAPVEC contains the upper triangle of the variance-covariance matrix of the function g(.) of the order statistics stored column-wise.

18: PAREST(4*IP+1) - real array

On exit: the statistics calculated by the routine as follows. The first IP components of PAREST contain the score statistics. The next IP elements contain the parameter estimates. PAREST($2 \times$ IP+1) contains the value of the χ^2 statistic. The next IP elements of PAREST contain the standard errors of the parameter estimates. Finally, the remaining IP elements of PAREST contain the z-statistics.

19: WORK(LWORK) — real array

20: LWORK — INTEGER

On entry: the dimension of the array WORK as declared in the (sub)program from which G08RBF is called.

Constraint: LWORK \geq NMAX \times (IP+1).

21: IWA(4*NMAX) — INTEGER array

22: IFAIL — INTEGER

On entry: IFAIL must be set to 0, -1 or 1. For users not familiar with this parameter (described in Chapter P01) the recommended value is 0.

On exit: IFAIL = 0 unless the routine detects an error (see Section 6).

6 Error Indicators and Warnings

If on entry IFAIL = 0 or -1, explanatory error messages are output on the current error message unit (as defined by X04AAF).

Errors detected by the routine:

```
\mathrm{IFAIL}=1
```

$$IFAIL = 2$$

On entry, ICEN $(i) \neq 0$ or 1 for some $1 \leq i \leq$ NSUM.

Workspace

Input/Output

Workspace

Input

IFAIL = 3

On entry, all the observations are adjudged to be tied. The user is advised to check the value supplied for TOL.

IFAIL = 4

The matrix $X^{T}(B-A)X$ is either ill-conditioned or not positive-definite. This error should only occur with extreme rankings of the data.

IFAIL = 5

On entry, at least one column of the matrix X has all its elements equal.

7 Accuracy

The computations are believed to be stable.

8 Further Comments

The time taken by the routine depends on the number of samples, the total number of observations and the number of parameters fitted.

In extreme cases the parameter estimates for certain models can be infinite, although this is unlikely to occur in practice. See Pettitt [2] for further details.

9 Example

A program to fit a regression model to a single sample of 40 observations using just one explanatory variable.

9.1 Program Text

Note. The listing of the example program presented below uses bold italicised terms to denote precision-dependent details. Please read the Users' Note for your implementation to check the interpretation of these terms. As explained in the Essential Introduction to this manual, the results produced may not be identical for all implementations.

```
GO8RBF Example Program Text
*
     Mark 14 Revised. NAG Copyright 1989.
      .. Parameters ..
*
                       NIN, NOUT
      INTEGER
     PARAMETER
                       (NIN=5,NOUT=6)
     INTEGER
                       NSMAX, NX, NMXMAX, NSMMAX, IPMAX, NPVAR, LWORK
     PARAMETER
                       (NSMAX=5,NX=100,NMXMAX=NX,NSMMAX=NX,IPMAX=6,
                       NPVAR=IPMAX+1,LWORK=NMXMAX*(IPMAX+1))
     +
      .. Local Scalars ..
     real
                       GAMMA, TOL
      INTEGER
                       I, IFAIL, IP, J, NMAX, NS, NSUM
      .. Local Arrays ..
                       ETA(NMXMAX), PAREST(4*IPMAX+1),
     real
                       PARVAR(NPVAR, IPMAX), VAPVEC(NMXMAX*(NMXMAX+1)/2),
     +
     +
                       WORK(LWORK), X(NX, IPMAX), Y(NSMMAX), ZIN(NMXMAX)
     INTEGER
                       ICEN(NSMMAX), IRANK(NMXMAX), IWA(4*NMXMAX),
                       NV(NSMAX)
      .. External Subroutines ..
     EXTERNAL
                       GO8RBF
      .. Intrinsic Functions ..
     INTRINSIC
                       MAX
      .. Executable Statements ..
     WRITE (NOUT,*) 'GO8RBF Example Program Results'
```

```
Skip heading in data file
*
     READ (NIN,*)
     Read number of samples, number of parameters to be fitted,
     distribution power parameter and tolerance criterion for ties.
     READ (NIN,*) NS, IP, GAMMA, TOL
     WRITE (NOUT,*)
     IF (NS.GT.O .AND. NS.LE.NSMAX .AND. IP.GT.O .AND. IP.LE.IPMAX)
         THEN
     +
        WRITE (NOUT, 99999) 'Number of samples =', NS
        WRITE (NOUT, 99999) 'Number of parameters fitted =', IP
        WRITE (NOUT,99998) 'Distribution power parameter =', GAMMA
        WRITE (NOUT, 99998) 'Tolerance for ties =', TOL
        Read the number of observations in each sample
        READ (NIN,*) (NV(I),I=1,NS)
        NMAX = O
        NSUM = 0
        DO 20 I = 1, NS
            NSUM = NSUM + NV(I)
            NMAX = MAX(NMAX,NV(I))
  20
        CONTINUE
         IF (NMAX.GT.O .AND. NMAX.LE.NMXMAX .AND. NSUM.GT.O .AND.
     +
            NSUM.LE.NSMMAX) THEN
            Read in observations, design matrix and censoring variable
            READ (NIN,*) (Y(I),(X(I,J),J=1,IP),ICEN(I),I=1,NSUM)
            IFAIL = 0
            CALL GO8RBF(NS,NV,NSUM,Y,IP,X,NX,ICEN,GAMMA,NMAX,TOL,PARVAR,
     +
                        NPVAR, IRANK, ZIN, ETA, VAPVEC, PAREST, WORK, LWORK,
                        IWA, IFAIL)
     +
*
            WRITE (NOUT,*)
            WRITE (NOUT, *) 'Score statistic'
            WRITE (NOUT, 99997) (PAREST(I), I=1, IP)
            WRITE (NOUT, *)
            WRITE (NOUT,*) 'Covariance matrix of score statistic'
            DO 40 J = 1, IP
               WRITE (NOUT,99997) (PARVAR(I,J),I=1,J)
  40
            CONTINUE
            WRITE (NOUT,*)
            WRITE (NOUT, *) 'Parameter estimates'
            WRITE (NOUT, 99997) (PAREST(IP+I), I=1, IP)
            WRITE (NOUT,*)
            WRITE (NOUT,*) 'Covariance matrix of parameter estimates'
            DO 60 I = 1, IP
               WRITE (NOUT,99997) (PARVAR(I+1,J),J=1,I)
  60
            CONTINUE
            WRITE (NOUT,*)
            WRITE (NOUT, 99996) 'Chi-squared statistic =',
              PAREST(2*IP+1), ' with', IP, ' d.f.'
    +
            WRITE (NOUT,*)
            WRITE (NOUT,*) 'Standard errors of estimates and'
            WRITE (NOUT, *) 'approximate z-statistics'
            WRITE (NOUT, 99995) (PAREST(2*IP+1+I), PAREST(3*IP+1+I), I=1,
              TP)
     +
        END IF
     END IF
     STOP
```

99999 FORMAT (1X,A,I2) 99998 FORMAT (1X,A,F10.5) 99997 FORMAT (1X,F9.3) 99996 FORMAT (1X,A,F9.3,A,I2,A) 99995 FORMAT (1X,F9.3,F14.3) END

9.2 Program Data

```
      GO8RBF Example Program Data

      1
      1
      0.00001
      0.00001

      40

      143.0
      0.0
      0
      164.0
      0.0
      0
      188.0
      0.0
      0
      190.0
      0.0
      0

      192.0
      0.0
      0
      200.0
      0.0
      0
      213.0
      0.0
      0
      216.0
      0.0
      0

      220.0
      0.0
      0
      227.0
      0.0
      0
      230.0
      0.0
      0
      213.0
      0.0
      0
      216.0
      0.0
      0

      265.0
      0.0
      0
      227.0
      0.0
      0
      216.0
      0.0
      0
      234.0
      0.0
      0
      246.0
      0.0
      0

      265.0
      0.0
      0
      304.0
      0.0
      0
      216.0
      0.0
      1
      244.0
      0.0
      1
      142.0
      1.0
      0

      156.0
      1.0
      0
      136.0
      1.0
      0
      233.0
      1.0
      0
      233.0
      1.0
      0
      233.0
      1.0
      0
      233.0
      1.0
      0
      233.0
      1.0
      0
      233.0
```

9.3 Program Results

```
GO8RBF Example Program Results
```

```
Number of samples = 1
Number of parameters fitted = 1
                                0.00001
Distribution power parameter =
Tolerance for ties = 0.00001
Score statistic
    4.584
Covariance matrix of score statistic
    7.653
Parameter estimates
    0.599
Covariance matrix of parameter estimates
    0.131
                          2.746 with 1 d.f.
Chi-squared statistic =
Standard errors of estimates and
approximate z-statistics
                 1.657
    0.361
```