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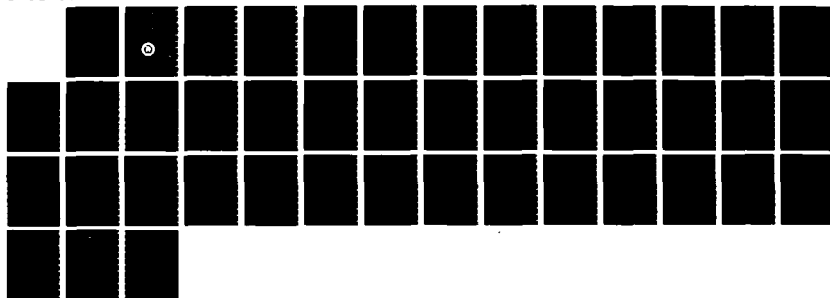
AN ALGORITHM FOR THE COMPUTATION OF GENERALIZED
LIKELIHOOD OR SELF-CRITIC. (U) RENSSELAER POLYTECHNIC
INST TROY NY SCHOOL OF MANAGEMENT T A DELAWANTY ET AL.
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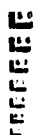
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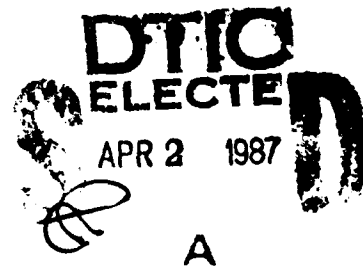
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**AN ALGORITHM FOR THE COMPUTATION OF GENERALIZED
LIKELIHOOD OR SELF-CRITICAL ESTIMATORS
FOR BINARY DATA**

by

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**An Algorithm for the Computation of Generalized Likelihood
or Self-Critical Estimators for Binary Data**

by

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and

A.S. Paulson*

Rensselaer Polytechnic Institute



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Documentation for Self-Critical Binary Program

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1. Purpose

Subroutine BINARY is an implementation of the self-critical estimation procedure of Paulson, Presser and Lawrence (1983). The logistic, Gaussian or Type I extreme value distribution may be selected as tolerance distribution. Estimates are expressed in location-scale form on entry and exit, but results may be printed out in regression form, location-scale form, or in both forms (see Section 3 for a description of the two parametrizations). It is possible to hold location parameters constant during the estimation procedure. Estimation is accomplished by a Newton-Raphson method.

Reference: F. T. O'Connell, "Stress models."

2. Specification (FORTRAN)

SUBROUTINE BINARY (N, IX, X, IA, NPAR, ISUB, ISTART, IDEP, IDIST, C,
REL TOL, ABSTOL, MAXIT, IPRINT, IFLAG, BETA, KLOGL, ICOV, COV, LMEM,
MEMORY, IFAULT)

3. Description

The routine is applicable to the following modeling situation: For $i = 1, \dots, n$ (n is the sample size), let v_i be the stress variable, a_i a zero-one indicator of withstand or failure, and $X_i = (x_{i1}, \dots, x_{ip})^T$ a column vector of covariates. A constant is incorporated as a covariate identically equal to unity. The case $p=0$ is possible, but should be rare.

The tolerance distribution and density, $F(v_i)$ and $f(v_i)$ respectively, depend on the covariates, a scale parameter σ , and a vector of location parameters $\beta = (\beta_1, \dots, \beta_p)^T$ as follows, where

$$u_1 = \frac{(v_1 - \beta' X_1)}{\sigma},$$

- logistic, $F(v_1) = \frac{\exp(u_1)}{(1 + \exp(u_1))}$,

$$f(v_1) = \frac{1}{|\sigma|} \frac{\exp(u_1)}{(1 + \exp(u_1))^2};$$

- Gaussian, $F(v_1) = \Phi(u_1)$

(Φ is the standard Normal distribution function),

$$f(v_1) = \frac{1}{|\sigma| \sqrt{2\pi}} \exp(-\frac{1}{2} u_1^2);$$

- extreme value, $F(v_1) = 1 - \exp(-\exp(u_1))$,

$$f(v_1) = \frac{1}{|\sigma|} \exp(u_1 - \exp(u_1)).$$

In principle, it is possible for σ to be negative, but this contradicts the physical notion of a stress variable.

In the method of maximum likelihood, parameters θ are estimated by maximizing the log likelihood,

$$L(\theta) = \sum_{i=1}^n [a_i \log F(v_i) + (1-a_i) \log S(v_i)],$$

where $S = 1-F$. The self-critical procedure depends on a user-specified quantity c , and estimates θ by solving the system

$$\sum_{i=1}^n f^c(v_i) \left[a_i \frac{\partial}{\partial \theta} \log F(v_i) + (1-a_i) \frac{\partial}{\partial \theta} \log S(v_i) \right] = 0.$$

When $c = 0$, maximum likelihood estimates are obtained. As c increases from zero, increasingly robust estimators result. The sensitivity of parameter estimates to departures from model assumptions can be examined by starting with $c = 0$ and refitting the model for increasing values of c . Negative values of c seem less useful.

4. Numerical Method

For the purpose of estimation, the routine (internally) reparameterizes the problem in a "regression" form. In this parameterization, $F(\cdot)$ depends on

$$w_1 = \alpha v_1 + \mu' X_1$$

instead of

$$u_1 = \frac{v_1 - \beta' X_1}{\sigma}.$$

The reparameterization offers two advantages:

- 1) Computation of the necessary partial derivatives is simple;
- 2) For maximum likelihood estimation with a logistic tolerance distribution, it can be shown that the regression parameterization results in a concave maximization problem. We anticipate that it will have fairly good properties in the more complicated cases.

It has a disadvantage in terms of potential ill-conditioning of the Hessian (or Jacobian) matrix, so that the input data should be sensibly scaled (see Section 11). Parameters are assumed to be expressed in the easier to understand location-scale form on entry, and are transformed back to location-scale form on exit, so the user need not worry about details of reparameterization.

Let

$$S_{\theta 1} = f^C(v_1) \left[a_1 \frac{\partial}{\partial \theta} \log F(v_1) + (1-a_1) \frac{\partial}{\partial \theta} \log S(v_1) \right].$$

The gradient used in Newton-Raphson iteration has θ^h component

$$g_{\theta} = n^{-1} \sum_{i=1}^n S_{\theta 1}, \text{ and the Hessian has } (\theta, \theta') \text{ component } H_{\theta\theta'} = n^{-1} \sum_{i=1}^n \frac{\partial}{\partial \theta}, S_{\theta 1}.$$

(the scaling by n^{-1} should be helpful when n is large).

When $c = 0$, the estimated asymptotic covariance matrix of the estimators is $n^{-1}(-H)^{-1}$, while when $c \neq 0$ it is $n^{-1}H^{-1}VH^{-1}$, where V has (θ, θ') component

$V_{\theta\theta'} = n^{-1} \sum_{i=1}^n S_{\theta i} S_{\theta' i}$. The estimated asymptotic covariance matrix is ex-

pressed in location-scale form by pre- and post-multiplying it by the Jacobian of the underlying parameter transformation.

Newton-Raphson iteration converges when

$$\max_1 \left| \theta^{(k)} - \theta_1^{(k-1)} \right| < \text{ABSTOL}$$

and

$$\max_1 \left| \frac{\theta_1^{(k)}}{\theta^{(k-1)}} \right| < \text{RELTOL},$$

where superscripts represent iteration numbers and ABSTOL and RELTOL are user-supplied. The user also specifies a maximum allowable number MAXIT of iterations.

Solution of linear equations and matrix inversion is accomplished by subroutines DECOMP and SOLVE, taken from Forsythe, Malcolm and Moler (1977). These routines are of high numerical quality, and provide an estimate of the condition number of the input matrix, which is often useful. In the interest of portability, iterative improvement is not used.

5. Parameters

5.1 Input Parameters

- N - INTEGER
Sample size. Unchanged on exit.
- IX - INTEGER
Row dimension of data matrix X. Unchanged on exit.

X - REAL array of DIMENSION (IX, q), where $q \geq N$. Data matrix containing the dependent (stress) variable and all covariates. Each column corresponds to one observation. If parameters are held fixed, values of the dependent variable will be changed, but their input values restored. Unchanged on exit.

IA - INTEGER array of DIMENSION N. Withstand or failure is indicated for observation I according as $IA(I) = 0$ or $IA(I) \neq 0$. Unchanged on exit.

NPAR - INTEGER. Number of parameters in the model. Unchanged on exit.

ISUB - INTEGER array of DIMENSION (NPAR). Indicates rows of data matrix X corresponding to parameters (the scale parameter is taken to correspond to the dependent variable). For instance, if $ISUB(3) = 5$, the third parameter corresponds to the fifth row of X. Unchanged on exit.

ISTART - INTEGER array of DIMENSION (NPAR). Indicates the status of parameters in the model, and whether a starting value is to be supplied. If $ISTART(I)$ is equal to
 0, BETA(I) is to be estimated, and its input value is to be disregarded;
 1, BETA(I) is to be estimated, and its input value is to be used as starting value;
 2, BETA(I) is to be held constant at its input value.
 (See Section 11.2 for comments on starting values; the scale of parameter is not allowed to be held constant.) Unchanged on exit.

IDEP - INTEGER. Row of dependent (stress) variable in X. Unchanged on exit.

IDIST - INTEGER. Indicates the tolerance distribution desired. If IDIST equals
 1, logistic distribution will be used;
 2, Gaussian distribution will be used;
 3, extreme value distribution will be used. Unchanged on exit.

C - DOUBLE PRECISION User-supplied constant for self-critical estimation. Unchanged on exit.

REL TOL - DOUBLE PRECISION.
Relative convergence tolerance for Newton-Raphson iteration
(see Section 4).
Unchanged on exit.

MAXIT - INTEGER.
Maximum allowable number of Newton-Raphson iteration.
Unchanged on exit.

IPRINT - INTEGER.
Output unit number. If IPRINT ≤ 0 , no output will be produced.
If IPRINT > 0 , standard output will be produced on logical
output unit IPRINT.
Unchanged on exit.

IFLAG - INTEGER.
Indicator for form of output.
If IFLAG < 0 , the standard output summary will be in regression
form. If IFLAG = 0, output summaries will be produced for
both regression and location-scale parameterizations. If
IFLAG > 0 , the standard output summary will be in location-
scale form. (See Section 11.3).
Unchanged on exit.

5.2 Input/Output (and associated dimension) Parameters.

BETA - DOUBLE PRECISION array of DIMENSION (NPAR).
On entry, contains starting values as specified by ISTART.
On exit, contains parameter estimates, in location scale form.

XLOGL - DOUBLE PRECISION.
On exit, contains the log likelihood if $c = 0$ (maximum
likelihood estimation), and zero otherwise.

ICOV - INTEGER.
Row dimension for COV.
Unchanged on exit.

COV - DOUBLE PRECISION array of DIMENSION (ICOV, q), where $q \geq \text{NPAR}$.
On exit, estimated asymptotic covariance information for the
parameters. The diagonal contains standard errors, the strict
lower triangle correlations, and the strict upper triangle
covariances. If a parameter is held constant, all correspond-
ing entries are zero. The covariance matrix will be for the
location-scale parameterization rulers IFLAG < 0 , when it will
be in regression form. See Section 11.3.

5.3 Workspace (and associated dimension) parameters

LMEM - INTEGER.

Length of work array MEMORY, as declared in calling program unit.
 $LMEM \geq NPAR + 4*NACT*(1+NACT) + 2*max(NFIX, 2*NACT)$, where
NACT is the number of parameters estimated and NFIX = NPAR-NACT
is the number of parameters held constant.
Unchanged on exit.

MEMORY - INTEGER array of DIMENSION (LMEM).
Used as workspace.

5.4 Diagnostic Parameter

IFAULT - INTEGER.

Unless the routine finds an error or gives a warning, IFAULT
will be 0 on exit. See Section 6.

6. Error Indications and Warnings

Errors or warnings specified by the routine:

IFAULT < 0 IF IFAULT = -I, I > 0, then an arithmetic exception was about to occur on observation I, while computing partial derivatives. This failure should be rare. If the data have been sensibly scaled, and the starting values are not too bad (see Sections 11.1-11.2), the probable cause is a data error. The routine stops as soon as the error is detected, and output parameter values are not of interest.

IFAULT = 1 Input parameter outside expected range. This failure will occur if, on entry, $N < 1$, $NPAR < 1$, $IX < NPAR$, $ICOV < NPAR$, $RELTOL \leq 0$, $ABSTOL < 0$, $MAXIT < 1$, $IDIST < 1$, $IDIST > 3$, $IPLAG \leq 0$ and $IPRINT \leq 0$ (see Section 11.3), $ISTART(I) < 0$ or $ISTART(I) > 2$ for some I, $ISTART(I) = 2$ for all I, $ISUB(I) \neq IDEP$ for all I, or $ISUB(I) = IDEP$ and $ISTART(I) = 2$ for some I. (The last restriction means that the scale parameter cannot be held constant.) The routine stops without doing any calculations.

IFAULT = 2 Insufficient workspace. This failure will occur if, on entry, $LMEM < NPAR + 4*NACT*(1+NACT) + 2*MAX(NFIX, 2*NACT)$, where NACT is the number of parameters estimated ($ISTART(I) < 2$), and NFIX = NPAR - NACT is the number of parameters held constant ($ISTART(I) = 2$). The routine stops without doing any calculations.

IFAULT = 3 The Hessian matrix has become numerically singular during Newton-Raphson iteration. The routine transforms estimates to location-scale form, restores values of the dependent variable if parameters were held fixed, and stops. Output parameter values are not generally of interest.

IPFAULT = 4 The Newton-Raphson iteration did not converge to within RELTOL and ABSTOL in the specified MAXIT iterations. The routine transforms estimates to location-scale form, restores values of the dependent variable if parameters were held fixed, and stops. Output parameter values are not generally of interest.

7. Auxiliary Routines

SUBROUTINE PREPAR(DX, N, X, NPAR, BETA, ISUB, ISTART, NACT, IACT, IMAP, PAR, NFIX, IPIXED, IDEP, WORK)

Prepares for iteration by setting up some indexing and working arrays, subtracting the effects of fixed parameters from the dependent variable, and setting initial active parameter values in regression form.

SUBROUTINE NEWTON(C, MAXIT, RELTOL, ABSTOL, DIFFER, XLOGL, NACT, IACT, PAR, IPIVOT, HESS, HESFAC, N, DX, X, IA, IPRINT, LWORK, WORK, LPAULT)

Carries out the Newton-Raphson iteration.

SUBROUTINE VARIAB(C, XLOGL, REGRES, DIFFER, VROUT, NACT, IACT, IMAP, PAR, IPIVOT, HESS, HESFAC, ICOV, NPAR, COV, DET, N, DX, X, IA, WORK)

Computes the estimated asymptotic covariance matrix, transforms it to location-scale form if requested, and computes correlations and standard errors.

SUBROUTINE EXPOST(REGRES, IDEP, NPAR, BETA, ISTART, NACT, PAR, IMAP, N, DX, X, NFIX, IPIXED, WORK)

Sets the output vector BETA, transforms to location-scale form if requested. If location-scale form is requested and parameters were held fixed, restore initial values of dependent variable.

SUBROUTINE RESULT (C, REGRES, IPRINT, IDIST, N, IDEP, NACT, NFIX, XLOGL, NPAR, BETA, ISUB, ISTART, ICOV, COV, DET)

Produces a standard output summary on logical output unit IPRINT.

The following routines are specific to particular tolerance distributions, and are declared EXTERNAL in BINARY. The first three, prefixed D, are passed to NEWTON and VARIAB. The next three, prefixed V, are passed to VARIAB.

```

SUBROUTINE DLOGBN(C, OBJECT, NACT, IACT, PAR, GRAD, HESS, N, IX, X,
IA,, IFAULT)
SUBROUTINE DGAUBN ( _____ " _____)
SUBROUTINE DEXVBN ( _____ " _____)

```

Compute gradient and Hessian for logistic, Gaussian and extreme value models, resp.

```

SUBROUTINE VLOGBN(C, NACT, IACT, PAR, V, N, IX, X, IA)
SUBROUTINE VGAUBN(_____ " _____)
SUBROUTINE VEXUBN(_____ " _____)

```

Compute V factor of estimated asymptotic covariance matrix for logistic, Gaussian and extreme value models, resp.

The following procedures perform general numerical tasks. They have been included in the interest of portability, although equivalents exist on many computer systems.

```

SUBROUTINE DECOMP(NDIM, N, A, UL, COND, IPUT, WORK)
SUBROUTINE SOLVE(NDIM, N, A, B, X, IPUT)

```

DECOMP decomposes a matrix into LU factors and estimates its condition. SOLVE solves a linear system, using the results of DECOMP. These routines are in Forsythe, Malcolm and Moler (1977), but the present versions include an extra argument so that A and B need not be overwritten.

```

SUBROUTINE DMXMLT(A, IA, N1, B, IB, N2, C, IC, N3, WORK, LWORK,
IFLAG, IFAULT)

```

Double precision matrix multiplication -
 $A(N1 \times N3) = B(N1 \times N2) * C(N2 \times N3)$, where B is overwritten if IFLAG < 0 and c is overwritten if IFLAG > 0.

```

DOUBLE PRECISION FUNCTION ALNORM (X, UPPER)
Algorithm as 66 (Hill, 1973) to compute tail areas of the standard Normal curve.

```

```

DOUBLE PRECISION FUNCTION RMILLS(X)
Computes  $Z(x)/Q(x)$ , the reciprocal of Mills' ratio, where x is a standard Normal variate. This procedure is based on procedures by Hill (1973) and Adams (1969).

```

8. References

Adams, A.G. (1969). Algorithm 39. Areas under the normal curve. Computer J., 12, 197-198.

Forsythe, G.E. Malcolm, M.A. and Moler, C.B. (1977). Computer Methods for Mathematical Computations. Englewood Cliffs, New Jersey: Prentice-Hall.

Hill, I.D. (1973). Algorithm AS 66. The normal integral. Applied Statistics, 22, 424-427.

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9. Storage

There are no internally declared arrays.

10. Precision, Machine Dependent Constants

The routine was developed on an IBM computers. The data matrix is single precision to conserve storage. To convert to single precision, take the following steps, with the exception noted below:

- 1) Change all DOUBLE PRECISION declarations to REAL;
- 2) Replace references to double precision FORTRAN library functions with single precision versions, e.g., EXP replaces DEXP;
- 3) Replace double precision constants by their single precision versions, e.g., 1.0 replaces 1.0D0.

Note: In routines PREPAR and EXPOST, the dependent variable is adjusted for the effect of fixed parameters. These adjustments must be calculated in double precision

to avoid a loss of significant digits in X which could affect subsequent computations.

Procedures DLOGBN, VLOGBN, DGAUBN, VGAUBN, DEXUBN, VEXUBN, DECOMP, ALNORM and RMILLS use machine-dependent constants whose values are set in DATA statements. These constants may have to be altered for some computers. They are pointed out by comments in the program units.

11. Further Comments

11.1 Scaling, conditioning of Hessian matrix

The user should be aware of potential problems of ill-conditioning. Because a regression-type parameterization is used, calculation of the Hessian matrix involves operations similar to the formation of $X^T X$, where X is the data matrix. Unfortunately, the resulting Hessian is often rather ill-conditioned. If IPRINT > 0, an estimate of the Hessian's condition number is printed out at each iteration. Roughly speaking, a condition number in excess of 10^7 is worrisome, although condition numbers in excess of 10^{10} can be tolerated when working in double precision on an IBM 3081. The user can and should avoid potentially excessive ill-conditioning by scaling the data matrix X before calling the routine.

Although it is not known how to optimally scale a problem, it seems that a nearly ideal scaling will be achieved if nonconstant variables are transformed to the range $[-1, 1]$, centered at zero. However, such precise scaling is often tedious. It seems most important to "equilibrate" the data matrix so that all variables have roughly the same (moderate) magnitude. Such equilibration is often simply accomplished by dividing by suitable powers of 10, e.g., expressing voltages in megavolts instead of kilovolts. Centering the variables will further reduce the condition number. One often centers covariates anyway, so that the "intercept"

parameter has a clear interpretation.

11.2. Starting values

Since Newton-Raphson procedure is employed, starting values are important. If no starting values are supplied by the user ($ISTART(I) = 0$ for all I), the routine will use unity for the scale parameter and zero for all location parameters. These starting values will not be acceptable unless the dependent variable has been centered and scaled. It is recommended that the user employ the mean and standard deviation of the dependent variable as starting values for the "intercept" parameter (if any) and scale parameter, respectively. If the model is sequentially refit with different values of c , it is recommended that estimates from the most recent call be used as starting values for the next call.

11.3 Output Flags

Because the routine will generally be called sequentially with different values of c , it is most convenient to always express entry and exit parameter values in the same form. The location-scale form is used, because that form affords the clearest interpretation. An inconsistency arises if $IFLAG < 0$, for then the output estimates are in location-scale form, but the covariance matrix is in regression form. Thus, if the user wants to make separate use of the output parameters of BINARY, the routine should be called with $IFLAG \geq 0$.

It is recommended that the routine be called with $IPRINT > 0$. If $IFLAG \leq 0$, the requirement $IPRINT > 0$ is enforced. (When $IFLAG < 0$, the requirement is in keeping with the inconsistency mentioned above.) The standard output summary should

be sufficient for most applications. For the maximum flexibility in output and interpretation of results, call BINARY with IPRINT > 0, IFLAG = 0.

12. Example

The following program illustrates the use of BINARY, and the standard output produced when IPRINT > 0 and IFLAG = 0.

```

C.....BINAO001
C (OPTIONALLY ROBUST) BINARY DATA ESTIMATION ROUTINE. BINAO002
C LOGISTIC, EXTREME VALUE, OR GAUSSIAN TOLERANCE DISTRIBUTION BINAO003
C CAN BE USED. PARAMETERS ON INPUT/OUTPUT ARE IN LOCATION- BINAO004
C SCALE FORM, BUT ESTIMATION IS CARRIED OUT USING A BINAO005
C REGRESSION-TYPE REPARAMETERIZATION. IF PRINTOUT IS DESIRED BINAO006
C (IPRINT GT 0), RESULTS CAN BE PRINTED OUT IN REGRESSION BINAO007
C FORM, LOCATION-SCALE FORM, OR BOTH. ARGUMENT IFLAG BINAO008
C SPECIFIES THE OUTPUT DESIRED. BINAO009
C FAILURE CODES - BINAO010
C IFALT = 1 - INPUT ERROR BINAO011
C IFALT = 2 - INSUFFICIENT WORKSPACE SUPPLIED BINAO012
C IFALT = 3 - HESSIAN MATRIX IS NUMERICALLY SINGULAR BINAO013
C IFALT = 4 - NO CONVERGENCE IN MAXIT ITERATIONS BINAO014
C IFALT = -1 - AN EXCEPTION WAS ABOUT TO OCCUR WHILE BINAO015
C PROCESSING THE I TH OBSERVATION BINAO016
C.....BINAO017
C SUBROUTINE BINARY(N, IX, X, IA, NPAR, ISUB, ISTART, IDEP, IDIST, BINAO018
C C, RELTOL, ABSTOL, MAXIT, IPRINT, IFLAG, BETA, XLOGL, BINAO019
C ICOV, COV, LMEM, MEMORY, IFAULT) BINAO020
C EXTERNAL PROCEDURES FOR PARTICULAR TOLERANCE DISTRIBUTIONS BINAO021
C EXTERNAL OLOGBN, DGAUBN, DEXVBN, VLOGBN, VGAUBN, VEXVBN BINAO022
C ARGUMENTS BINAO023
C INTEGER N, IX, IA(N), NPAR, ISUB(NPAR), ISTART(NPAR), IDEP, IDIST, BINAO024
C MAXIT, IPRINT, IFLAG, ICOV, LMEM, MEMORY(LMEM), IFAULT BINAO025
C REAL X(IX,N) BINAO026
C DOUBLE PRECISION C, RELTOL, ABSTOL, BETA(NPAR), XLOGL, BINAO027
C COV(ICOV,NPAR) BINAO028
C LOCAL SCALARS BINAO029
C LOGICAL REGRES BINAO030
C INTEGER NACT, NFIX, IND, IND1, MIACT, MIFIX, MIMAP, MIPVT, MPAR, BINAO031
C MHFESS, MHFAC, MWORK BINAO032
C DOUBLE PRECISION DET, ZERO BINAO033
C CONSTANTS BINAO034
C DATA ZERO /0.000/ BINAO035
C BINAO036
C CHECK FOR INPUT ERRORS BINAO037
C IFALT = 1 BINAO038
C IF (N LT 1 OR NPAR LT 1 OR IX LT NPAR OR ICOV LT BINAO039
C NPAR) RETURN BINAO040
C IF (RELTOL LE ZERO OR ABSTOL LE ZERO OR MAXIT LT 1) BINAO041
C RETURN BINAO042
C IF (IDIST LT 1 OR IDIST GT 3) RETURN BINAO043
C IF (IFLAG LE 0 AND IPRINT LE 0) RETURN BINAO044
C BINAO045
C MEMORY MANAGEMENT, POSSIBLE RELATED INPUT ERRORS. CHECK BINAO046
C ISUB(=), ISTART(=) VECTORS. COUNT ACTIVE PARAMETERS. BINAO047
C NOTE SCALE PARAMETER CANNOT BE FIXED. BINAO048
C NACT = 0 BINAO049
C IND1 = 0 BINAO050
C DO 10 I = 1, NPAR BINAO051
C IND = ISUB(I) BINAO052
C IF (IND LT 1 OR IND GT IX) RETURN BINAO053
C IF (IND EQ IDEP) IND1 = 1 BINAO054
C IND = ISTART(I) BINAO055
C IF (IND LT 0 OR IND GT 2) RETURN BINAO056
C IF (IND NE 2) NACT = NACT + 1 BINAO057
C BINAO058
C 10 CONTINUE BINAO059
C IF (NACT EQ 0 OR IND1 EQ 0) RETURN BINAO060
C IF (ISTART(IND1) EQ 2) RETURN

```

```

C      NFIX = NPAR - NACT
C      CHECK IF WORKSPACE SIZE IS ADEQUATE, ALLOCATE IT
C      IF(AULT = 2)
C        IND = 2 * NACT
C        IND1 = IND * NACT
C        IF (LMEM .LT. NPAR + 2*NACT + IND + 2*IND1 + MAXO(2*NFIX, 2*IND))
C          RETURN
C        MIACI = 1
C        MIFX = MIACI + NACT
C        MIMAP = MIACI + NPAR
C        MIPVT = MIMAP + NACT
C        MPAR = MIPVT + NACT
C        MHES = MPAR + IND
C        MHFAC = MHES + IND1
C        MWORK = MHFAC + IND1
C
C      PREPARE FOR ESTIMATION - SET UP SUBSCRIPT ARRAYS AND
C      STARTING VALUES FOR ACTIVE PARAMETERS, SUBTRACT EFFECT OF
C      FIXED PARAMETERS FROM DEPENDENT VARIABLE
C      CALL PREPAR(X, N, X, NPAR, BETA, ISUB, ISTART, NACT,
C      1 MEMORY(MIACI), MEMORY(MIMAP), MEMORY(MPAR), NFIX,
C      2 MEMORY(MIFIX), IOEP, MEMORY(MWORK))
C
C      NEWTON-RAPHSON ITERATION WITH EXTERNAL ROUTINE PASSED FOR
C      FIRST AND SECOND PARTIALS
C      IF (IDIST.EQ. 1) CALL NEWTON(C, MAXIT, RELTOL, ABSTOL, DLOGBN,
C      1 XLOGL, NACT, MEMORY(MIACI), MEMORY(MPAR), MEMORY(MIPVT),
C      2 MEMORY(MHES), MEMORY(MHFAC), N, IX, X, IA, IPRINT, 2*NACT,
C      3 MEMORY(MWORK), IFAULT)
C      IF (IDIST.EQ. 2) CALL NEWTON(C, MAXIT, RELTOL, ABSTOL, DGAUBN,
C      1 XLOGL, NACT, MEMORY(MIACI), MEMORY(MPAR), MEMORY(MIPVT),
C      2 MEMORY(MHES), MEMORY(MHFAC), N, IX, X, IA, IPRINT, 2*NACT,
C      3 MEMORY(MWORK), IFAULT)
C      IF (IDIST.EQ. 3) CALL NEWTON(C, MAXIT, RELTOL, ABSTOL, DEXVBN,
C      1 XLOGL, NACT, MEMORY(MIACI), MEMORY(MPAR), MEMORY(MIPVT),
C      2 MEMORY(MHES), MEMORY(MHFAC), N, IX, X, IA, IPRINT, 2*NACT,
C      3 MEMORY(MWORK), IFAULT)
C
C      ERROR HANDLING IF NEWTON-RAPHSON PROCEDURE FAILS.
C      ON ERROR, PRINT OUT MESSAGE IF IPRINT.GT. O. SET UP
C      THE FINAL CALL TO EXPST()
C      IF (IFAULT.EQ. O) GO TO 30
C      REGRES = FALSE
C      IF (IPRINT.LE. O) GO TO 50
C      IF (IFAULT.GT. O) GO TO 20
C      IND = -IFAULT
C      WRITE (IPRINT,60) IND
C      GO TO 50
C
C      20 IFAULT = IFAULT + 1
C      IF (IFAULT.EQ. 3) WRITE (IPRINT,70)
C      IF (IFAULT.EQ. 4) WRITE (IPRINT,80) MAXIT
C      GO TO 50
C
C      COMPUTE APPROXIMATE ASYMPTOTIC COVARIANCE MATRIX.
C      REGRES IS A FLAG FOR WHETHER OR NOT TO SET UP OUTPUT IN
C      REGRESSION FORM. STATEMENT 30, THE FIRST BRANCH POINT,
C      SETS UP REGRES FOR FIRST OUTPUT (IF REGRESSION AND
C      LOCATION-SCALE BOTH REQUESTED, REGRESSION GOES FIRST).
C      STATEMENT 40, THE SECOND BRANCH POINT, IS FOR REPEAT
C      BINA0061
C      BINA0062
C      BINA0063
C      BINA0064
C      BINA0065
C      BINA0066
C      BINA0067
C      BINA0068
C      BINA0069
C      BINA0070
C      BINA0071
C      BINA0072
C      BINA0073
C      BINA0074
C      BINA0075
C      BINA0076
C      BINA0077
C      BINA0078
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C      BINA0093
C      BINA0094
C      BINA0095
C      BINA0096
C      BINA0097
C      BINA0098
C      BINA0099
C      BINA0100
C      BINA0101
C      BINA0102
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C      BINA0104
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C      BINA0109
C      BINA0110
C      BINA0111
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C      BINA0114
C      BINA0115
C      BINA0116
C      BINA0117
C      BINA0118
C      BINA0119
C      BINA0120

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C      CALL (ALWAYS LOCATION-SCALE)
30 REGRES = IFLAG LE O
40 IF (IDIST EQ 1) CALL VARIABLE(C, XLOGL, REGRES, DLOGBN, VLOGBN,
   1 NACT, MEMORY(MIACT), MEMORY(MINAP), MEMORY(MPAR)).
2     MEMORY(MIPVT), MEMORY(MHRESS), MEMORY(MHFAC), ICOV, NPAR, COV,
3     DET, N, IX, X, IA, MEMORY(MWORK))
   1 IF (IDIST EQ 2) CALL VARIABLE(C, XLOGL, REGRES, DGAUBN, VGAUBN,
   1 NACT, MEMORY(MIACT), MEMORY(MINAP), MEMORY(MPAR)).
2     MEMORY(MIPVT), MEMORY(MHRESS), MEMORY(MHFAC), ICOV, NPAR, COV,
3     DET, N, IX, X, IA, MEMORY(MWORK))
   1 IF (IDIST EQ 3) CALL VARIABLE(C, XLOGL, REGRES, DEXVBN, VEXVBN,
   1 NACT, MEMORY(MIACT), MEMORY(MINAP), MEMORY(MPAR)).
2     MEMORY(MIPVT), MEMORY(MHRESS), MEMORY(MHFAC), ICOV, NPAR, COV,
3     DET, N, IX, X, IA, MEMORY(MWORK))
C      C
C      ARRANGE ESTIMATED PARAMETERS IN BETA(*) FOR OUTPUT.
C      IF LOCATION-SCALE FORM, TRANSFORM PARAMETERS AND POSSIBLY
C      RESTORE DEPENDENT VARIABLE TO VALUE ON ENTRY.
C      50 CALL EXPOST(REGRES, IDEP, NPAR, BETA, ISTART, NACT, MEMORY(MPAR),
   1     MEMORY(MIMAP), N, IX, X, NFIX, MEMORY(MIFIX), MEMORY(MWORK))
C      C
C      IF NEWTON-RAPHSON FAILED, GOODBYE.
C      IF (IFAULT NE O) RETURN
C      C
C      PRODUCE STANDARD OUTPUT IF REQUESTED.
C      IF (IPRINT GT O) CALL RESULT(C, REGRES, IPRINT, IDIST, N, IDEP,
   1     NACT, NFIX, XLOGL, NPAR, BETA, ISUB, ISTART, ICOV, COV, DET)
C      C
C      IF OUTPUT DESIRED IN BOTH REGRESSION AND LOCATION-SCALE
C      FORMS, LOOP BACK, ELSE EXIT
REGRES = NOT REGRES
IF (NOT REGRES AND IFLAG EQ O) GO TO 40
C      C
C      IF ONLY REGRESSION FORM OUTPUT WAS REQUESTED, AN EXTRA
CALL TO EXPOST() IS NEEDED TO RESTORE LOCATION-SCALE FORM
IF (IFLAG LT O) CALL EXPOST(REGRES, IDEP, NPAR, BETA, ISTART,
   1 NACT, MEMORY(MPAR), MEMORY(MIMAP), N, IX, X, NFIX,
   2 MEMORY(MIFIX), MEMORY(MWORK))
C      C
60 FORMAT ('FAILURE - EXCEPTION ON OBSERVATION NO.', I5)
70 FORMAT ('FAILURE - HESSIAN MATRIX IS NUMERICALLY SINGULAR')
80 FORMAT ('WARNING - NO CONVERGENCE IN', I4, ' ITERATIONS - POSSIBLE
   1 FAILURE')
RETURN
END
C.....
C      PREPARE FOR ANALYSIS BY SETTING UP THE FOLLOWING ARRAYS -
PREPOO02    IACT(*) - X(*) ROWS FOR ACTIVE PARAMETERS
PREPOO03    IFIXED(*) - X(*) ROWS FOR FIXED PARAMETERS
PREPOO04    IMAP(*) - POSITIONS IN BETA(*) OF ACTIVE PARAMETERS
PREPOO05    PAR(*) - INITIAL VALUES OF ACTIVE PARAMETERS IN
PREPOO06    REGRESSION FORM
C      C
C      ALSO SUBTRACT EFFECTS OF FIXED PARAMETERS FROM DEPENDENT
VARIABLE, USING WORK(*)
C      C
C      NOTE - DEPENDENT VARIABLE (SCALE) PARAMETER IS ALWAYS
PLACED FIRST IN PAR(*), FOR CONVENIENCE LATER
C.....
SUBROUTINE PREPAR(IX, N, X, NPAR, BETA, ISUB, ISTART, NACT, IACT,
   1 IMAP, PAR, NFIX, IFIXED, IDEP, WORK)
C.....
C      ARGUMENTS
PREPOO15

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PREP0074
PREP0075

INTEGER IX, N, NPAR, ISUB(NPAR), ISTART(NPAR), NACT, IACT(NACT),
1   IMAP(NACT), NFIX, IFIXED(1), IDEP
REAL X(IX,N)
DOUBLE PRECISION BETA(NPAR), PAR(NACT), WORK(1)
LOCAL SCALARS
INTEGER IND, IND1, LACT, ISCALE
DOUBLE PRECISION TEMP, ZERO, ONE
DATA ZERO, ONE /0.000, 1.000/

C
C   SET UP IACT(*), IFIXED(*), IMAP(*)
LACT = 0
IND = 0
ISCALE = 0
DO 20 I = 1, NPAR
  IF (ISTART(I) EQ 2) GO TO 10
  LACT = LACT + 1
  IMAP(LACT) = I
  IND1 = ISUB(I)
  IACT(LACT) = IND1
  IF (IND1 EQ IDEP) ISCALE = I
  GO TO 20
10  IND = IND + 1
  IFIXED(IND) = ISUB(I)
  WORK(IND) = BETA(I)
20  CONTINUE

C
C   SWITCH PLACES SO SCALE IS FIRST ACTIVE PARAMETER
IND = IACT(1)
IACT(1) = IACT(ISCALE)
IACT(ISCALE) = IND
IND = IMAP(1)
IMAP(1) = IMAP(ISCALE)
IMAP(ISCALE) = IND

C
C   MAIN LOOP OVER SAMPLE IF PARAMETERS ARE FIXED, TO ADJUST
C   DEPENDENT VARIABLE (TO BE RESET ON EXIT FROM BINARY())
IF (NFIX EQ 0) GO TO 50
DO 40 I = 1, N
  THE FOLLOWING INNER PRODUCT MUST BE ACCUMULATED IN DOUBLE
  PRECISION
  TEMP = DBLE(X(IDEP,I))
  DO 30 J = 1, NFIX
    IND = IFIXED(J)
    TEMP = TEMP - WORK(J) * X(IND,I)
30  CONTINUE
  X(IDEP,I) = SNGL(TEMP)

C
C   40 CONTINUE

C
C   SET STARTING VALUES
50 DO 60 I = 1, NPAR
60 PAR(I) = ZERO
  TEMP = ONE
  IND = IMAP(1)
  IF (ISTART(IND) EQ 1 AND BETA(IND) NE ZERO) TEMP = BETA(IND)
  PAR(1) = ONE / TEMP
  IF (NACT EQ 1) RETURN
  DO 70 I = 2, NACT
    IND = IMAP(I)

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      IF (ISTART(IND) EQ. 1) PAR(1) = -BETA(IND) / TEMP
      70 CONTINUE
C
      RETURN
      END
C.....
C      NEWTON-RAPHSON ITERATION
C      FAILURE CODES -
C      IFALT = 1 - INSUFFICIENT WORKSPACE (LESS THAN 2 * NACT
C      LOCATIONS)
C      IFALT = 2 - THE HESSIAN MATRIX IS NUMERICALLY SINGULAR
C      IFALT = 3 - CONVERGENCE HAS NOT OCCURRED IN MAXIT ITNS
C      IFALT = -1 - AN EXCEPTION WAS ABOUT TO OCCUR WHILE
C      PROCESSING THE 1 TH OBSERVATION IN DIFFER()
C.....
C      SUBROUTINE NEWTON(C, MAXIT, RELTOL, ABSTOL, DIFFER, XLOGL, NACT,
1      IACT, PAR, IPIVOT, HESS, HESFAC, N, IX, X, IA, IPRINT,
2      LWORK, WORK, IFAULT)
C      ARGUMENTS - DIFFER IS EXTERNAL ROUTINE FOR PARTIALS
C      INTEGER MAXIT, NACT, IACT(NACT), IPIVOT(NACT), N, IX, IA(N),
1      IPRINT, LWORK, IFAULT
C      REAL X(IX,N)
C      DOUBLE PRECISION C, RELTOL, ABSTOL, XLOGL, PAR(NACT),
1      HESS(NACT,NACT), HESFAC(NACT,NACT), WORK(LWORK)
C      LOCAL SCALARS
C      INTEGER ITER, IND
C      DOUBLE PRECISION GNORM, COND, ABSERR, RELERR, TEMP, XNEW, ZERO,
1      ONE
C      DATA ZERO, ONE /0.000, 1.000/
C
      IFALT = 1
      IF (LWORK LT. 2*NACT) RETURN
      IFALT = 0
      IND = NACT + 1
      ITER = 0
      IF (IPRINT GT. 0) WRITE (IPRINT,50) C, MAXIT, RELTOL, ABSTOL
C
C      LOOPING POINT FOR ITERATION - THE FIRST NACT LOCATIONS OF
C      WORK(*) ARE USED FOR GRADIENT/INCREMENT, WHILE THE NEXT
C      NACT LOCATIONS ARE WORKSPACE FOR DECOMP
C
10  ITER = ITER + 1
      CALL DIFFER(C, XLOGL, NACT, IACT, PAR, WORK(1), HESS, N, IX, X,
1      IA, IFAULT)
      IF (IFALT LT. 0) RETURN
C
C      FACTOR THE HESSIAN INTO L * U, CHECK IF SINGULAR
C      CALL DECOMP(NACT, NACT, HESS, HESFAC, COND, IPIVOT, WORK(IND))
      IF (COND + ONE NE. COND) GO TO 20
      IFALT = 2
      RETURN
C
C      SOLVE THE SYSTEM HESS * INCREMENT = -GRADIENT,
C      OVERWRITING GRADIENT
C
20  GNORM = ZERO
      DO 30 I = 1, NACT
      TEMP = WORK(1)
      WORK(1) = -TEMP
      GNORM = GNORM + TEMP * TEMP
30  CONTINUE
      GNORM = DSQRT(GNORM)

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PREP0079
PREP0080

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C      CALL SOLVE(NACT, NACT, HESFAC, WORK, IPIVOT)
C
C      INCREMENT PARAMETERS, COMPUTE CONVERGENCE CRITERIA
      RELERR = ZERO
      ABSERR = ZERO
      DO 40 I = 1, NACT
        TEMP = WORK(I)
        ABSERR = DMAX1(ABSERR, DABS(TEMP))
        XNEW = PAR(I) + TEMP
        PAR(I) = XNEW
        IF (ONE + XNEW .NE. ONE) TEMP = TEMP / XNEW
        RELERR = DMAX1(RELERR, DABS(TEMP))
      40 CONTINUE
      IF (IPRINT .GT. 0) WRITE (IPRINT, 60) ITER, GNORM, RELERR, ABSERR,
      1 COND
C
C      CHECK CONVERGENCE, SET FAILURE CODE IF NONE
C      IF (RELERR .LT. RELTOL .AND. ABSERR .LT. ABSTOL) RETURN
      IF (ITER .LT. MAXIT) GO TO 10
      IF AULT = 3
C
C      50 FORMAT ('1 C      MAX ITNS REL TOLER ABS TOLER'/D11.3, I9,
      1 2D11.2)
      60 FORMAT ('0 ITN GRAD NORM REL CHANGE ABS CHANGE'/I5, 3D12.2, /,
      1 1 ESTIMATED CONDITION NUMBER OF HESSIAN IS', D10.2)
      RETURN
      END
C.....
C      COMPUTATIONS FOR VARIABILITY OF THE ESTIMATORS
C      ESTIMATE ASYMPTOTIC COVARIANCE MATRIX, TRANSFORM IT FROM
C      REGRESSION TO LOCATION-SCALE FORM, COMPUTE ASYMPTOTIC
C      CORRELATIONS AND STD ERRORS
C.....
C      SUBROUTINE VARIAB(C, XLDGL, REGRES, DIFFER, VROUT, NACT, IACT,
      1 IMAP, PAR, IPIVOT, HESS, HESFAC, ICOV, NPAR, COV, DET,
      2 N, IX, X, IA, WORK)
C      ARGUMENTS - DIFFER AND VROUT ARE SUBROUTINES
C      LOGICAL REGRES
C      INTEGER NACT, IACT(NACT), IMAP(NACT), IPIVOT(NACT), ICOV, NPAR, N,
      1 IX, IA(N)
      REAL X(IX,N)
      DOUBLE PRECISION C, XLDGL, PAR(NACT), HESS(NACT, NACT),
      1 HESFAC(NACT, NACT), COV(ICOV, NPAR), DET, WORK(NACT)
C      LOCAL SCALARS
C      LOGICAL FLAG
      INTEGER IND, IND1, IND2
      DOUBLE PRECISION ALPHA, ALPHA2, TEMP, TEMP1, ZERO, ONE, SMALL
C      MACHINE-DEPENDENT CONSTANT - SMALL SET SO THAT DEXP(X)
C      WILL CAUSE EXCEPTION IF X .LT. SMALL
      DATA ZERO, ONE, SMALL /0.000, 1.000, -180.000/
C
C      CALCULATE HESSIAN AT OPTIMAL POINT AND FACTOR IT
C      (THIS MAY BE WASTEFUL IN SOME CASES, BUT IT AVOIDS
C      SOME LOGICAL COMPLICATIONS.) WORK(*) USED AS SCRATCH
C      THE HESSIAN IS ASSUMED NONSINGULAR, AS IT SHOULD BE IF
C      THIS POINT IS REACHED FACTORIZATION TO HESFAC(*,*)
C      CALL DIFFER(C, XLDGL, NACT, IACT, PAR, WORK, HESS, N, IX, X, IA,
      1 IND)
      CALL DECOMP(NACT, NACT, HESS, HESFAC, TEMP, IPIVOT, WORK)
C

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C      INVERT NEGATIVE OF HESSIAN, USING IPIVOT(*) AND
C      FACTORIZATION IN HESFAC(*,*). PLACE INVERSE IN HESS(*,*)
DO 20 J = 1, NACT
  DO 10 I = 1, NACT
    HESS(I,J) = ZERO
  HESS(J,J) = -ONE
  CALL SOLVE(NACT, NACT, HESFAC, HESS(I,J), HESS(I,J), IPIVOT)
20 CONTINUE
  IF (C EQ ZERO) GO TO 30

C      SECTION FOR ROBUST ANALYSIS -
C      PLACE PRODUCT (H INVERSE) * V * (H INVERSE) IN HESS(*,*)
C      FIRST COMPUTE V(*,*), PLACE IN COV(*,*)
C      CALL VRUT(C, NACT, IACT, PAR, COV, N, IX, X, IA)

C      MULTIPLY HESS * COV, OVERWRITING COV(*,*)
C      CALL DMXMT(COV, NACT, NACT, HESS, NACT, NACT, COV, NACT, NACT,
1      WORK, NACT, 1, IND)

C      MULTIPLY COV * HESS, OVERWRITING HESS(*,*)
C      CALL DMXMT(HESS, NACT, NACT, COV, NACT, NACT, HESS, NACT, NACT,
1      WORK, NACT, 1, IND)

C      ALL VALUES OF C - IF LOCATION-SCALE FORM, TRANSFORM
C      THE COVARIANCE MATRIX.
30 IF (REGRES) GO TO 80
  ALPHA = PAR(1)
  ALPHA2 = ALPHA * ALPHA
  FLAG = NACT EQ 1

C      LEFT-MULTIPLY HESS(*,*) BY JACOBIAN, RESULT TO HESFAC(*,*)
DO 50 J = 1, NACT
  TEMP = HESS(1,J) / ALPHA2
  HESFAC(1,J) = -TEMP
  IF (FLAG) GO TO 50
  DO 40 I = 2, NACT
    HESFAC(I,J) = PAR(I) * TEMP - HESS(I,J) / ALPHA
  40 CONTINUE

C      RIGHT-MULTIPLY HESFAC(*,*) BY TRANSPOSE OF JACOBIAN,
C      RESULT TO HESS(*,*)
DO 70 I = 1, NACT
  TEMP = HESFAC(1,I) / ALPHA2
  HESS(1,I) = -TEMP
  IF (FLAG) GO TO 70
  DO 60 J = 2, NACT
    HESS(I,J) = PAR(J) * TEMP - HESFAC(I,J) / ALPHA
  60 CONTINUE

C      BOTH PARAMETERIZATIONS - DIVIDE COVARIANCE MATRIX BY
C      SAMPLE SIZE
C      TEMP = DBLE(FLOAT(N))
DO 90 J = 1, NACT
  DO 90 I = 1, J
    HESS(I,J) = HESS(I,J) / TEMP
    HESS(J,I) = HESS(I,J)
90 CONTINUE

C      FIND DETERMINANT OF COVARIANCE MATRIX - FACTOR IT,
C      RESETING IPIVOT(*) AND HESFAC(*,*)

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      CALL DECOMP(NACT, NACT, HESS, HESFAC, TEMP, IPIVOT, WORK)
      DET = ZERO
      IF (TEMP + ONE EQ TEMP) GO TO 110
      IND = IPIVOT(NACT)
      DO 100 I = 1, NACT
        TEMP1 = HESFAC(I,1)
        IF (TEMP1 LT ZERO) IND = -IND
        DET = DET + DLOG(DABS(TEMP1))
      100 CONTINUE
C
C      TAKE THE NACT ROOT OF DETERMINANT. SET DET TO ZERO
C      IF IT UNDERFLOWS OR IS NEGATIVE.
C      DET = DET / DBLE(FLOAT(NACT))
      DET = DET / LT SMALL OR IND LT 0) DET = ZERO
      IF (DET GE SMALL AND IND GT 0) DET = DEXP(DET)
C
C      MOVE CONTENTS OF HESS(*,*) TO UPPER TRIANGLE OF COV(*,*)
      DO 110 J = 1, NPAR
        DO 120 I = 1, NPAR
          DO 130 I = 1, NACT
            IND = IMAP(I)
            DO 140 J = 1, IND
              IND1 = IMAP(J)
              IND2 = MINO(IND,IND1)
              IND1 = MAXO(IND,IND1)
              COV(IND2,IND1) = HESS(I,J)
            140 CONTINUE
          150 CONTINUE
        160 CONTINUE
      170 CONTINUE
C
C      PUT STD ERRORS ON DIAG OF COV(*,*). CORRELATIONS BELOW
      DO 180 I = 1, NPAR
        TEMP = COV(I,1)
        IF (TEMP EQ ZERO) GO TO 150
        TEMP = DSORT(TEMP)
        COV(I,1) = TEMP
        IF (I EQ 1) GO TO 150
        IND = I - 1
        DO 190 J = 1, IND
          TEMP1 = COV(J,J)
          IF (TEMP1 NE ZERO) COV(I,J) = COV(J,I) / (TEMP*TEMP1)
        190 CONTINUE
      200 CONTINUE
C
      RETURN
      END
C.....EXP00001
C      EX POST ADJUSTMENTS BEFORE EXIT
C      PUT OPTIMAL PARAMETERS IN BETA(*). IF LOCATION-SCALE
C      FORM, TRANSFORM THE PARAMETERS AND RESTORE DEPENDENT
C      VARIABLE TO INITIAL VALUES IF PARAMETERS WERE FIXED.
C.....EXP00006
C      SUBROUTINE EXPST(REGRES, IDEP, NPAR, BETA, ISTART, NACT, PAR,
C      1 IMAP, N, IX, X, NFIX, IFIXED, WORK)
C      ARGUMENTS
C      LOGICAL REGRES
C      INTEGER IDEP, NPAR, ISTART(NPAR), NACT, IMAP(NACT), N, IX, NFIX,
C      1 IFIXED(1)
C      REAL X(IX,N)
C      DOUBLE PRECISION BETA(NPAR), PAR(NACT), WORK(1)
C      LOCAL SCALARS

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 VARIO131
 VARIO132
 VARIO133
 VARIO134
 VARIO135
 VARIO136
 VARIO137
 VARIO138
 EXP00001
 EXP00002
 EXP00003
 EXP00004
 EXP00005
 EXP00006
 EXP00007
 EXP00008
 EXP00009
 EXP00010
 EXP00011
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 EXP00013
 EXP00014
 EXP00015

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EXP00016
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EXP00020
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RESU0017
RESU0018
RESU0019

INTEGER IND
DOUBLE PRECISION TEMP, ONE
DATA ONE / 1.0D0/

C      INSERT WORKING PARAMETERS PAR(*) IN BETA(*), IN THE
C      SUITABLE FORM.
C      TEMP = PAR(1)
IND = IMAP(1)
IF (REGRES) BETA(IND) = TEMP
IF ( NOT REGRES) BETA(IND) = ONE / TEMP
IF (NACT.EQ.1) GO TO 20
DO 10 I = 2, NACT
  IND = IMAP(I)
  IF (REGRES) BETA(IND) = PAR(I)
  IF ( NOT REGRES) BETA(IND) = -PAR(I) / TEMP
10 CONTINUE

C      IF LOCATION-SCALE FORM AND PARAMETERS WERE HELD FIXED,
C      RESTORE DEPENDENT VARIABLE
C      20 IF (REGRES OR NFIX .EQ. 0) RETURN
IND = 0
DO 30 I = 1, NPAR
  IF (ISTART(I) NE 2) GO TO 30
  IND = IND + 1
  WORK(IND) = BETA(I)
30 CONTINUE
DO 50 I = 1, N

C      THE FOLLOWING INNER PRODUCT MUST BE ACCUMULATED IN DOUBLE
C      PRECISION
C      TEMP = DBLE(X(IDEP,I))
C      DO 40 J = 1, NFIX
C        IND = IFIXED(J)
C        TEMP = TEMP + WORK(J) * X(IND,I)
40 CONTINUE
X(IDEP,I) = SNGL(TEMP)

C      50 CONTINUE
C      RETURN
C      END
C.....
C      PRINT OUT TYPICAL OUTPUT SUMMARY ON OUTPUT UNIT IPRINT
C.....
C      SUBROUTINE RESULT(C, REGRES, IPRINT, IDIST, N, IDEP, NACT, NFIX,
1       XLOGL, NPAR, BETA, ISUB, ISTART, ICOV, COV, DET)
C      ARGUMENTS
C      LOGICAL REGRES
C      INTEGER IPRINT, IDIST, N, IDEP, NACT, NFIX, NPAR, ISUB(NPAR),
1       ISTART(NPAR), ICOV
C      DOUBLE PRECISION C, XLOGL, BETA(NPAR), COV(ICOV,NPAR), DET
C      LOCAL SCALARS
C      LOGICAL MLE
C      INTEGER IND
C      DOUBLE PRECISION TSTAT, ZERO
C      DATA ZERO / 0.0D0/

C      BASIC HEADINGS
C      MLE = C.EQ.ZERO
C      IF (MLE) WRITE (IPRINT,40)

```

```

C          FAILURE CODE - IFAULT = -1 - AN EXCEPTION WAS ABOUT TO
C          OCCUR WHILE PROCESSING THE I TH OBSERVATION
C          .....
C          SUBROUTINE DLOGM(C, OBJECT, NACT, IACT, PAR, GRAD, HESS, N, IX,
1          X, IA, IFAULT)
C          ARGUMENTS
C          INTEGER NACT, IACT(NACT), N, IX, IA(N), IFAULT
C          REAL X(IX,N)
C          DOUBLE PRECISION C, OBJECT, PAR(NACT), GRAD(NACT), HESS(NACT,NACT)
C          LOCAL SCALARS
C          LOGICAL MLE
C          INTEGER IND
C          DOUBLE PRECISION EXTRA, F, S, G, FC, EITHER, DDENS, XJI, ZERO,
1          ONE, BIG
C          MACHINE-DEPENDENT CONSTANT - BIG ROUGHLY CHOSEN SO THAT
C          DEXP(X) WILL CAUSE EXCEPTION IF /X/ .GT. BIG
C          DATA ZERO, ONE, BIG /0.000, 1.000, 174.000/
C
C          IFAULT = 0
C          MLE = ONE + C .EQ. ONE
C          EXTRA = ONE / PAR(1)
C          OBJECT = ZERO
C          DO 10 J = 1, NACT
C            GRAD(J) = ZERO
C            DO 10 I = J, NACT
C              HESS(I,J) = ZERO
C            10 CONTINUE
C
C          MAIN LOOP OVER SAMPLE
C          DO 90 I = 1, N
C            FC = ZERO
C            DO 20 J = 1, NACT
C              IND = IACT(J)
C              FC = FC + PAR(J) * X(IND,I)
C            20 CONTINUE
C            IF (DABS(FC) .GT. BIG) GO TO 110
C            FC = DEXP(FC)
C            S = ONE + FC
C            F = FC / S
C            S = ONE / S
C            G = F * S
C            IF (IA(1) .EQ. 0) GO TO 30
C            EITHER = S
C            XJI = F
C            GO TO 40
C          30 EITHER = -F
C            XJI = S
C          40 IF (MLE) GO TO 50
C            FC = G ** C
C            GO TO 60
C          50 OBJECT = OBJECT + DLOG(XJI)
C            FC = ONE
C
C          LOOP TO INCREMENT GRADIENT AND HESSIAN UPPER TRIANGLE
C          60 EITHER = FC * EITHER
C            G = FC * G
C            FC = C * EITHER
C            DO 80 J = 1, NACT
C              IND = IACT(J)
C              XJI = DBLE(X(IND,I))

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DLOG0005
DLOG0006
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DLOG0059
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DLOG0062
DLOG0063
DLOG0064

```

IF ( NOT MLE) WRITE (IPRINT,50) C
IF (IDIST EQ 1) WRITE (IPRINT,60)
IF (IDIST EQ 2) WRITE (IPRINT,70)
IF (IDIST EQ 3) WRITE (IPRINT,80)
WRITE (IPRINT,90) N, NPAR, NACT, NFIX
IF (REGRES) WRITE (IPRINT,100)
IF ( NOT REGRES) WRITE (IPRINT,110)

C
C
      PARAMETER ESTIMATES
      WRITE (IPRINT,120)
      DO 20 I = 1, NPAR
        IND = ISUB(I)
        IF (ISTART(I) LT 2) GO TO 10
        WRITE (IPRINT,130) I, IND, BETA(I)
      GO TO 20
10    TSTAT = BETA(I) / COV(I,I)
      IF (IND EQ IDEP) WRITE (IPRINT,140) I, IND, BETA(I), COV(I,I), RESU0036
1    TSTAT
      IF (IND NE IDEP) WRITE (IPRINT,150) I, IND, BETA(I), COV(I,I), RESU0038
1    TSTAT
20 CONTINUE

C
C
      ASYMPTOTIC QUANTITIES
      IF (MLE) WRITE (IPRINT,160) XLOGL
      IF (DET LE ZERO) WRITE (IPRINT,170)
      IF (DET GT ZERO) WRITE (IPRINT,180) NACT, DET
      WRITE (IPRINT,190)
      DO 30 I = 1, NPAR
30    WRITE (IPRINT,200) (COV(I,J),J=1,I)

C
C
      40 FORMAT ('MAXIMUM LIKELIHOOD BINARY ANALYSIS')
      50 FORMAT ('SELF-CRITICAL BINARY ANALYSIS, C = ', D11.3)
      60 FORMAT ('LOGISTIC TOLERANCE DISTRIBUTION')
      70 FORMAT ('GAUSSIAN TOLERANCE DISTRIBUTION')
      80 FORMAT ('EXTREME VALUE TOLERANCE DISTRIBUTION')
      90 FORMAT ('ANALYSIS OF ', I6, ' CASES AND ', I4, ' VARIABLES'/'O',
     1   '14', ' PARAMETERS WERE ESTIMATED AND ', I4,
     2   ' WERE HELD CONSTANT', 2('','O'))
100 FORMAT ('PARAMETERS ARE EXPRESSED IN REGRESSION FORM'/'O')
110 FORMAT ('PARAMETERS ARE EXPRESSED IN LOCATION-SCALE FORM'/'O')
120 FORMAT ('1 VAR', 8X, 'BETA(1)', 6X, 'STD ERR', 9X,
     1   'T STAT', 5X, 'STATUS')
130 FORMAT (2I5, D15.7, 35X, 'FIXED LOCATION')
140 FORMAT (2I5, 3D15.7, 5X, 'ESTIMATED SCALE')
150 FORMAT (2I5, 3D15.7, 5X, 'ESTIMATED LOCATION')
160 FORMAT (2I/, 'O'), '/', 'THE LOG LIKELIHOOD IS', D16.8)
170 FORMAT (2I/, 'O'), '/', 'THE ESTIMATED ASYMPTOTIC COVARIANCE',
     1   'MATRIX IS NOT POSITIVE DEFINITE')
180 FORMAT (2I/, 'O'), '/', 'THE', I4, ' ROOT OF THE DETERMINANT',
     1   'OF THE ESTIMATED'/'ASYMPTOTIC COVARIANCE MATRIX IS',
     2   D16.8)
190 FORMAT ('CORRELATED ASYMPTOTIC STD ERRORS (ON DIAGONAL)',
     1   'CORRELATIONS (BELOW DIAGONAL)'/' ')
200 FORMAT (10D12.4)
      RETURN
      END

C.....
C FIRST AND SECOND PARTIAL DERIVATIVES FOR LOGISTIC BINARY,
C REGRESSION PARAMETERIZATION PARTIALS ARE SCALED BY
C SAMPLE SIZE
DLOG0001
DLOG0002
DLOG0003
DLOG0004

```

```

DDENS = XJI * EITHER
GRAD(J) = GRAD(J) + DDENS
DDENS = XJI * (S - F)
IF (J.EQ.1) DDENS = DDENS + EXTRA
DDENS = DDENS + FC - G * XJI
DO 70 K = J, NACT
  IND = IACT(K)
  HESS(K,J) = HESS(K,J) + X(IND,I) * DDENS
70 CONTINUE
80 CONTINUE
90 CONTINUE

C
C
C SCALE GRADIENT AND HESSIAN BY SAMPLE SIZE
EXTRA = DBLE(FLOAT(N))
DO 100 J = 1, NACT
  GRAD(J) = GRAD(J) / EXTRA
  DO 100 I = J, NACT
    HESS(I,J) = HESS(I,J) / EXTRA
    HESS(J,I) = HESS(I,J)
100 CONTINUE
IFAUULT = 0
RETURN

C
C
C ERROR EXIT
110 IFAULT = -1
RETURN

C
C
C
C.....
C COMPUTE V(*,*) FACTOR FOR ASYMPTOTIC COVARIANCE MATRIX.
C LOGISTIC BINARY. CALLED ONLY WHEN C.NE. 0
C.....
C SUBROUTINE VLOGBN(C, NACT, IACT, PAR, V, N, IX, X, IA)
C ARGUMENTS
C INTEGER NACT, IACT(NACT), N, IX, IA(N)
C REAL X(IX,N)
C DOUBLE PRECISION C, PAR(NACT), V(NACT,NACT)
C LOCAL SCALARS
C INTEGER IND
C DOUBLE PRECISION FC, S, F, EITHER, TEMP, ZERO, ONE, BIG
C MACHINE-DEPENDENT CONSTANT - BIG ROUGHLY CHOSEN SO THAT
C DEXP(X) WILL CAUSE EXCEPTION IF /X/ .GT. BIG
C DATA ZERO, ONE, BIG / 0.000, 1.000, 174.000/

C
C
C SET UPPER TRIANGLE OF V(*,*) TO 0
DO 10 J = 1, NACT
  DO 10 I = J, NACT
    V(I,J) = ZERO
10 CONTINUE

C
C
C MAIN LOOP OVER SAMPLE
DO 50 I = 1, N
  FC = ZERO
  DO 20 J = 1, NACT
    IND = IACT(J)
    FC = FC + PAR(J) * X(IND,I)
20 CONTINUE
IF (DABS(FC) .GE. BIG) GO TO 50
FC = DEXP(FC)
S = ONE + FC
F = FC / S

```

```

S = ONE / S
FC = (F*S) ** C
EITHER = S
IF (IA(1) .EQ. 0) EITHER = F
EITHER = (EITHER*FC) ** 2

C
C
C INCREMENT TERM OF V(*,*)
DO 40 J = 1, NACT
  IND = IACT(J)
  TEMP = EITHER * X(IND,1)
  DO 30 K = J, NACT
    IND = IACT(K)
    V(K,J) = V(K,J) + TEMP * X(IND,1)
  30 CONTINUE
  40 CONTINUE
  50 CONTINUE

C
C
C DIVIDE V(*,*) BY SAMPLE SIZE, FILL OUT
TEMP = DBLE(FLOAT(N))
DO 60 J = 1, NACT
  DO 60 I = J, NACT
    V(I,J) = V(I,J) / TEMP
    V(J,1) = V(I,J)
  60 CONTINUE

C
C
C RETURN
END

C .....
C FIRST AND SECOND PARTIAL DERIVATIVES FOR BINARY GAUSSIAN.
C REGRESSION PARAMETERIZATION PARTIALS ARE SCALED BY
C SAMPLE SIZE
C FAILURE CODE - IFAULT = -1 - AN EXCEPTION WAS ABOUT TO
C OCCUR WHILE PROCESSING THE I TH OBSERVATION
C .....
SUBROUTINE DGAUBN(C, OBJECT, NACT, IACT, PAR, GRAD, HESS, N, IX,
1 X, IA, IFAULT)
C
C ARGUMENTS
C INTEGER NACT, IACT(NACT), N, IX, IA(N), IFAULT
C REAL X(IX,N)
C DOUBLE PRECISION C, OBJECT, PAR(NACT), GRAD(NACT), HESS(NACT,NACT)
C LOCAL SCALARS
C LOGICAL MLE
C INTEGER IND
C DOUBLE PRECISION EXTRA, DOT, DOTMIN, FC, CBY2, RATIO, EITHER,
1 HTERM, XJ1, DDENS, ZERO, ONE, TWO, BIG
C
C FUNCTIONS CALLED
C DOUBLE PRECISION ALNORM, RMILLS
C
C MACHINE-DEPENDENT CONSTANT - BIG ROUGHLY CHOSEN SO THAT
C DEXP(X) WILL CAUSE EXCEPTION IF /X/ .GT. BIG
C DATA ZERO, ONE, TWO, BIG / 0.000, 1.000, 2.000, 174.000/

C
C IFAULT = 0
C MLE = ONE + C .EQ. ONE
C EXTRA = ONE / PAR(1)
C CBY2 = C / TWO
C OBJECT = ZERO
C DO 10 J = 1, NACT
  GRAD(J) = ZERO
  DO 10 I = J, NACT
    HESS(I,J) = ZERO
  10 CONTINUE

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VLOG0033

VLOG0034

VLOG0035

VLOG0036

VLOG0037

VLOG0038

VLOG0039

VLOG0040

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VLOG0044

VLOG0045

VLOG0046

VLOG0047

VLOG0048

VLOG0049

VLOG0050

VLOG0051

VLOG0052

VLOG0053

VLOG0054

VLOG0055

VLOG0056

VLOG0057

VLOG0058

VLOG0059

DGAU0001

DGAU0002

DGAU0003

DGAU0004

DGAU0005

DGAU0006

DGAU0007

DGAU0008

DGAU0009

DGAU0010

DGAU0011

DGAU0012

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DGAU0014

DGAU0015

DGAU0016

DGAU0017

DGAU0018

DGAU0019

DGAU0020

DGAU0021

DGAU0022

DGAU0023

DGAU0024

DGAU0025

DGAU0026

DGAU0027

DGAU0028

DGAU0029

DGAU0030

DGAU0031

DGAU0032

DGAU0033

```

10 CONTINUE
C
C   MAIN LOOP OVER SAMPLE
DO 90 I = 1, N
  DOT = ZERO
  DO 20 J = 1, NACT
    IND = IACT(J)
    DOT = DOT + PAR(J) * X(IND,I)
  20 CONTINUE
  DOTMIN = -DOT
  IF (IA(I) .EQ. 0) GO TO 30
  RATIO = RMILLS(DOTMIN)
  EITHER = RATIO
  GO TO 40
30  RATIO = RMILLS(DOT)
  DOT = DOTMIN
  EITHER = -RATIO
  IF (MLE) GO TO 50
  XJI = CBY2 * DOT * DOT
  IF (DABS(XJI) .GT. BIG) GO TO 110
  FC = DEXP(-XJI)
  GO TO 60
50  FC = ONE
  XJI = ALNORM(DOT, .FALSE.)
  IF (XJI .EQ. ZERO) GO TO 110
  OBJECT = OBJECT + DLOG(XJI)
C
C   INCREMENT GRADIENT AND HESSIAN
  EITHER = FC * EITHER
  HTERM = -FC * RATIO * (DOT + RATIO)
  FC = C * EITHER
  DO 80 J = 1, NACT
    IND = IACT(J)
    XJI = DBLE(X(IND,I))
    GRAD(J) = GRAD(J) + XJI * EITHER
    DDENS = XJI * DOTMIN
    IF (J .EQ. 1) DDENS = DDENS + EXTRA
    DDENS = DDENS * FC + XJI * HTERM
    DO 70 K = J, NACT
      IND = IACT(K)
      HESS(K,J) = HESS(K,J) + X(IND,I) * DDENS
  70 CONTINUE
  80 CONTINUE
  90 CONTINUE
C
C   SCALE GRADIENT AND HESSIAN BY SAMPLE SIZE
  EXTRA = DBLE(FLOAT(N))
  DO 100 J = 1, NACT
    GRAD(J) = GRAD(J) / EXTRA
    DO 100 I = J, NACT
      HESS(I,J) = HESS(I,J) / EXTRA
      HESS(J,I) = HESS(I,J)
  100 CONTINUE
  IF ALT = 0
    RETURN
  C
  C   ERROR EXIT
  110 IF ALT = -1
    C
    RETURN

```

DGAU0034
 DGAU0035
 DGAU0036
 DGAU0037
 DGAU0038
 DGAU0039
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 DGAU0092
 DGAU0093

```

C.....
C      COMPUTE V(*.*) FACTOR FOR ASYMPTOTIC COVARIANCE MATRIX,
C      GAUSSIAN BINARY. CALLED ONLY WHEN C.NE.O
C.....
C      SUBROUTINE VGAUBN(C, NACT, IACT, PAR, V, N, IX, X, IA)
C.....
C      ARGUMENTS
C      INTEGER NACT, IACT(NACT), N, IX, IA(N)
C      REAL X(IX,N)
C      DOUBLE PRECISION C, PAR(NACT), V(NACT,NACT)
C      LOCAL SCALARS
C.....
C      INTEGER IND
C      DOUBLE PRECISION F2C, TEMP, EITHER, ZERO, BIG
C      FUNCTION CALLED
C.....
C      DOUBLE PRECISION RMILLS
C      MACHINE-DEPENDENT CONSTANT - BIG ROUGHLY CHOSEN SO THAT
C      DEFP(X) WILL CAUSE EXCEPTION IF /X/ .GT. BIG
C      DATA ZERO, BIG /0.000, 174.000/
C.....
C      SET UPPER TRIANGLE OF V(*.*) TO ZERO
C      DO 10 J = 1, NACT
C        DO 10 I = J, NACT
C          10 V(I,J) = ZERO
C.....
C      MAIN LOOP OVER SAMPLE
C      DO 70 I = 1, N
C        F2C = ZERO
C        DO 20 J = 1, NACT
C          IND = IACT(J)
C          F2C = F2C + PAR(J) * X(IND,I)
C          20 CONTINUE
C          TEMP = C * F2C * F2C
C          IF (DABS(TEMP) GE. BIG) GO TO 70
C          IF (IA(I) EQ. 0) GO TO 30
C          EITHER = RMILLS(-F2C)
C          GO TO 40
C          30 EITHER = RMILLS(F2C)
C          40 F2C = DEFP(-TEMP) * EITHER * EITHER
C          DO 60 J = 1, NACT
C            IND = IACT(J)
C            TEMP = F2C * X(IND,I)
C            DO 50 K = J, NACT
C              IND = IACT(K)
C              V(K,I) = V(K,J) + TEMP * X(IND,I)
C              50 CONTINUE
C              60 CONTINUE
C              70 CONTINUE
C.....
C      DIVIDE V(*.*) BY SAMPLE SIZE, FILL OUT
C      TEMP = DBLE(FLOAT(N))
C      DO 80 J = 1, NACT
C        DO 80 I = J, NACT
C          V(I,J) = V(I,J) / TEMP
C          V(J,I) = V(I,J)
C          80 CONTINUE
C.....
C      RETURN
C      END
C.....
C      FIRST AND SECOND PARTIAL DERIVATIVES FOR BINARY EXTREME
C.....
DGAU0009
VGAU0001
VGAU0002
VGAU0003
VGAU0004
VGAU0005
VGAU0006
VGAU0007
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VGAU0050
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VGAU0052
VGAU0053
VGAU0054
VGAU0055
VGAU0056
VGAU0057
DEXV0001
DEXV0002

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```

C VALUE, REGRESSION PARAMETERIZATION. PARTIALS ARE SCALED DEXV0003
C BY SAMPLE SIZE. DEXV0004
C FAILURE CODE - IFAULT = -1 - AN EXCEPTION WOULD HAVE DEXV0005
C OCCURRED WHEN PROCESSING THE I TH OBSERVATION DEXV0006
C..... DEXV0007
C SUBROUTINE DEXVBN(C, OBJECT, NACT, IACT, PAR, GRAD, HESS, N, IX, DEXV0008
C X, IA, IFAULT) DEXV0009
C ARGUMENTS DEXV0010
C INTEGER NACT, IACT(NACT), N, IX, IA(N), IFAULT DEXV0011
C REAL X(IX,N) DEXV0012
C DOUBLE PRECISION C, OBJECT, PAR(NACT), GRAD(NACT), HESS(NACT,NACT) DEXV0013
C LOCAL SCALARS DEXV0014
C LOGICAL MLE DEXV0015
C INTEGER IND DEXV0016
C DOUBLE PRECISION EXTRA, DOT, EXPON, DENFAC, EXPEXP, FC, EITHER, DEXV0017
C HTERM, ZERO, ONE, BIG, BIG1 DEXV0018
C MACHINE-DEPENDENT CONSTANTS - BIG ROUGHLY CHOSEN SO THAT DEXV0019
C DEXP(X) WILL CAUSE EXCEPTION IF /X/ .GT. BIG, BIG1 SUCH DEXV0020
C THAT 1.000 / DEXP(DEXP(X)) NE. 1.000 IF X .GT. BIG1 DEXV0021
C DATA ZERO, ONE, BIG, BIG1 / 0.000, 1.000, 174.000, -36.200/ DEXV0022
C IFAULT = 0 DEXV0023
C MLE = ONE + C .EQ. ONE DEXV0024
C EXTRA = ONE / PAR(1) DEXV0025
C OBJECT = ZERO DEXV0026
C DO 10 J = 1, NACT DEXV0027
C GRAD(J) = ZERO DEXV0028
C DO 10 I = J, NACT DEXV0029
C HESS(I,J) = ZERO DEXV0030
C 10 CONTINUE DEXV0031
C
C MAIN LOOP OVER SAMPLE DEXV0032
C DO 90 I = 1, N DEXV0033
C DOT = ZERO DEXV0034
C DO 20 J = 1, NACT DEXV0035
C IND = IACT(J) DEXV0036
C DOT = DOT + PAR(J) * X(IND,I) DEXV0037
C 20 CONTINUE DEXV0038
C IF (DABS(DOT) GE BIG) GO TO 110 DEXV0039
C EXPON = DEXP(DOT) DEXV0040
C DENFAC = ONE - EXPON DEXV0041
C IF (IA(I) EQ. 0) GO TO 40 DEXV0042
C
C FAILURE TERM, IA(I) = 1 DEXV0043
C IF (EXPON GE BIG OR DOT LE BIG1) GO TO 110 DEXV0044
C EXPEXP = DEXP(EXPON) DEXV0045
C DOT = ONE - ONE / EXPEXP DEXV0046
C FC = EXPON / EXPEXP DEXV0047
C EITHER = FC / DOT DEXV0048
C HTERM = EITHER * (DENFAC - EITHER) DEXV0049
C IF (MLE) GO TO 30 DEXV0050
C FC = FC ** C DEXV0051
C GO TO 60 DEXV0052
C FC = ONE DEXV0053
C OBJECT = OBJECT + DLOG(DOT) DEXV0054
C GO TO 60 DEXV0055
C
C NON FAILURE TERM, IA(I) = 0 DEXV0056
C EITHER = EXPON DEXV0057
C HTERM = EITHER DEXV0058
C 40 DEXV0059
C DEXV0060
C DEXV0061
C DEXV0062

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DEXV0065 IF (MLE) GO TO 50
DEXV0066 FC = C * (DOT - EXPON)
DEXV0067 IF (DABS(FC) GE BIG) GO TO 110
DEXV0068 FC = DEXP(FC)
DEXV0069 GO TO 60
DEXV0070 FC = ONE
DEXV0071 OBJECT = OBJECT + EITHER
DEXV0072
DEXV0073 LOOP TO INCREMENT THE DERIVATIVES
DEXV0074 EITHER = FC * EITHER
DEXV0075 MTERM = FC * MTERM
DEXV0076 FC = C * EITHER
DEXV0077 DO 80 J = 1, NACT
DEXV0078 IND = IACT(J)
DEXV0079 DOT = DBLE(X(IND,I))
DEXV0080 GRAD(J) = GRAD(J) + DOT * EITHER
DEXV0081 EXPON = DOT * DENFAC
DEXV0082 IF (J EQ 1) EXPON = EXPON + EXTRA
DEXV0083 EXPON = FC * EXPON + DOT * MTERM
DEXV0084 DO 20 K = J, NACT
DEXV0085 IND = IACT(K)
DEXV0086 HESS(K,J) = HESS(K,J) + X(IND,I) * EXPON
DEXV0087
DEXV0088 CONTINUE
DEXV0089
DEXV0090
DEXV0091 EXTRA = DBLE(FLOAT(N))
DEXV0092 DO 100 J = 1, NACT
DEXV0093 GRAD(J) = GRAD(J) / EXTRA
DEXV0094 DO 100 I = J, NACT
DEXV0095 HESS(I,J) = HESS(I,J) / EXTRA
DEXV0096 HESS(J,I) = HESS(I,J)
DEXV0097
DEXV0098 CONTINUE
DEXV0099 RETURN
DEXV0100
DEXV0101 ERROR EXIT - EXCEPTIONS
DEXV0102 110 IF(AULT = 1)
DEXV0103
DEXV0104 RETURN
DEXV0105 END
DEXV0001 .....
DEXV0002 .....
DEXV0003 .....
DEXV0004 .....
DEXV0005 .....
DEXV0006 .....
DEXV0007 .....
DEXV0008 .....
DEXV0009 .....
DEXV0010 .....
DEXV0011 .....
DEXV0012 .....
DEXV0013 .....
DEXV0014 .....
DEXV0015 .....
DEXV0016 .....
DEXV0017 .....
DEXV0018 .....
DEXV0019 .....

```

```

C
C
DO 10 I = 1, NACT
10 V(I,J) = ZERO
C
C
C MAIN LOOP OVER SAMPLE
DO 70 I = 1, N
DOT = ZERO
DO 20 J = 1, NACT
IND = 1ACT(J)
DOT = DOT + PAR(J) * X(IND,I)
20 CONTINUE
IF (DABS(DOT) GE BIG) GO TO 70
EXPON = DEXP(DOT)
IF (IA(I) EQ 0) GO TO 30
IF (DOT LT BIG) GO TO 70
FC = DEXP(EXPON)
DOT = ONE - ONE / FC
FC = EXPON / FC
EXPON = FC / DOT
FC = EXPON * FC ** C
GO TO 40
30 FC = C * (DOT - EXPON)
IF (DABS(FC) GT BIG) GO TO 70
FC = EXPON * DEXP(FC)
C
C
C SUMMANTS FOR V(*,*)
FC = FC * FC
DO 60 J = 1, NACT
IND = 1ACT(J)
DOT = FC * X(IND,I)
DO 50 K = J, NACT
IND = 1ACT(K)
V(K,J) = V(K,J) + DOT * X(IND,I)
50 CONTINUE
60 CONTINUE
70 CONTINUE
C
C
C DIVIDE V(*,*) BY SAMPLE SIZE, FILL OUT
DOT = DBLE(FLOAT(N))
DO 80 J = 1, NACT
DO 80 I = J, NACT
V(I,J) = V(I,J) / DOT
V(J,I) = V(I,J)
80 CONTINUE
C
C
C RETURN
END
C.....
C DOUBLE PRECISION MATRIX MULTIPLICATION
C X(N1 BY N3) = Y(N1 BY N2) * Z(N2 BY N3)
C THREE OPTIONS -
C IFLAG = 0 - X, Y, AND Z ARE DISTINCT
C IFLAG LT 0 - X, Y OVERLAP, CORNER OF Y OVERWRITTEN
C IFLAG GT 0 - X, Z OVERLAP, CORNER OF Z OVERWRITTEN
C.....
C SUBROUTINE DMXMLT(X, IX, N1, Y, IY, N2, Z, IZ, N3, WORK, LWORK,
1 IFLAG, IFAULT)
C
C ARGUMENTS
C INTEGER IX, N1, IY, N2, IZ, N3, LWORK, IFLAG, IFAULT
C DOUBLE PRECISION X(IX,N3), Y(IY,N2), Z(IZ,N3), WORK(LWORK)
C LOCAL SCALARS

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VEXV0020
 VEXV0021
 VEXV0022
 VEXV0023
 VEXV0024
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 VEXV0032
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 VEXV0063
 VEXV0064
 VEXV0065
 DMXV0001
 DMXV0002
 DMXV0003
 DMXV0004
 DMXV0005
 DMXV0006
 DMXV0007
 DMXV0008
 DMXV0009
 DMXV0010
 DMXV0011
 DMXV0012
 DMXV0013
 DMXV0014

```

C      DOUBLE PRECISION TEMP, ZERO
C      DATA ZERO /0.000/
C
C      ERROR EXITS
C      IF AULT = 1
C        IF (MINO(N1,N2,N3) LT 1 OR MINO(IX,IY) LT N1 OR IZ LT
1          N2) RETURN
C      IF AULT = 2
C        IF ((IFLAG LT 0 AND (N3 GT N2 OR LWORK LT N3))
1          .OR.
C        IF AULT = 3
C        IF ((IFLAG GT 0 AND (N1 GT IZ OR LWORK LT N1))
1          .OR.
C        IF AULT = 0
C        IF ((IFLAG NE 0) GO TO 30
C
C      STRAIGHTFORWARD, NO OVERWRITING
C      DO 20 I = 1, N1
C        DO 20 J = 1, N3
C          TEMP = ZERO
C          DO 10 K = 1, N2
C            TEMP = TEMP + Y(I,K) * Z(K,J)
10           X(I,J) = TEMP
C          CONTINUE
20          CONTINUE
C        RETURN
C
C      CORNER OF MATRIX Z IS OVERWRITTEN
C      30 IF ((IFLAG LT 0) GO TO 80
C        DO 70 J = 1, N3
C          DO 50 I = 1, N1
C            TEMP = ZERO
C            DO 40 K = 1, N2
C              TEMP = TEMP + Y(I,K) * Z(K,J)
40             WORK(I) = TEMP
C            CONTINUE
50             DO 60 I = 1, N1
C              X(I,J) = WORK(I)
60             CONTINUE
70             CONTINUE
C          RETURN
C
C      CORNER OF MATRIX Y IS OVERWRITTEN
C      80 DO 120 I = 1, N1
C        DO 100 J = 1, N3
C          TEMP = ZERO
C          DO 90 K = 1, N2
C            TEMP = TEMP + Y(I,K) * Z(K,J)
90             WORK(J) = TEMP
C          CONTINUE
100            DO 110 J = 1, N3
C              X(I,J) = WORK(J)
110            CONTINUE
120            CONTINUE
C
C      RETURN
C      END
C.....ALND0001
C      DOUBLE PRECISION VERSION OF ALGORITHM AS 66, APPLIED
C      ALND0002
C      STATISTICS (1973), VOL 22, NO 3
C      EVALUATES THE TAIL AREA OF THE STANDARD NORMAL CURVE
C      FROM X TO INFINITY IF UPPER IS TRUE OR FROM
C      MINUS INFINITY TO X IF UPPER IS FALSE
C      ALND0003
C      ALND0004
C      ALND0005
C      ALND0006

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C.....ALND0007
C DOUBLE PRECISION FUNCTION ALNORM(X,UPPER)
C ARGUMENTS
C LOGICAL UPPER
C DOUBLE PRECISION X
C LOCAL SCALARS
C LOGICAL UP
C DOUBLE PRECISION LTONE, UTZERO, ZERO, HALF, ONE, CON, Z, Y
C MACHINE-DEPENDENT CONSTANTS - LTONE = (N + 9) / 3, WHERE
C N IS NO. OF DECIMAL DIGITS IN DOUBLE PRECISION NUMBER,
C UTZERO SUCH THAT DEXP(-X*X/2.000) WILL CAUSE AN EXCEPTION
C IF X .GT. UTZERO
C CONSTANTS IN EXPRESSIONS ARE AS IN AS 66
C DATA LTONE, UTZERO, ZERO, HALF, ONE, CON /8.000, 18.6500, 0.000,
C 1 0.500, 1.000, 1.2800/
C
C UP = UPPER
C Z = X
C IF (Z .GE. ZERO) GO TO 10
C UP = NOT UP
C Z = -Z
C 10 IF (Z .LE. LTONE .OR. UP .AND. Z .LE. UTZERO) GO TO 20
C ALNORM = ZERO
C GO TO 40
C 20 Y = HALF * Z * Z
C IF (Z .GT. CON) GO TO 30
C
C ALNORM = HALF - Z * (0.3989422804400 - 0.39990343850400*Y/(Y + 5.
C 17588548045800 - 29.821355780800/(Y + 2.62433312167900+48.
C 269593069200/(Y + 5.9288572443800)))
C GO TO 40
C
C 30 ALNORM = 0.39894228038500 * DEXP(-Y) / (Z - 3.80520 - 8 + 1.
C 10000061530200/(Z + 3.980647940 - 4 + 1.9861538136400/(Z - 0.
C 215167911663500+5.2933032492600/(Z + 4.838591280800 - 15.
C 3150897245100/(Z + 0.74238092402700+30.78993303400/(Z + 3.
C 49901941701100))))))
C
C 40 IF (NOT UP) ALNORM = ONE - ALNORM
C
C RETURN
C END
C.....RMIL0001
C RECIPROCAL OF MILLS RATIO, Z(X) / Q(X), X A STANDARD NORMAL
C VARIATE, BASED ON FUNCTION ALNORM(), ALGORITHM AS 66.
C.....RMIL0003
C DOUBLE PRECISION FUNCTION RMILLS(X)
C ARGUMENTS
C DOUBLE PRECISION X
C LOCAL SCALARS
C LOGICAL UP
C DOUBLE PRECISION Z, Y, LTONE, UTZERO, ZERO, HALF, ONE, CON, FPI1,
C 1 FPI2
C MACHINE-DEPENDENT CONSTANTS - LTONE = (N + 9) / 3, WHERE
C N IS NO. OF DECIMAL DIGITS IN DOUBLE PRECISION NUMBER,
C UTZERO SUCH THAT DEXP(-X*X/2.000) WILL CAUSE AN EXCEPTION
C IF X .LT. UTZERO
C FPI1 = SQRT(2/PI), FPI2 = 1/SQRT(2*PI)
C CONSTANTS IN EXPRESSIONS ARE AS IN AS 66
C DATA LTONE, UTZERO, ZERO, HALF, ONE, CON, FPI1, FPI2 /8.000,

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1      -18 6500, 0 000, 0 500, 1 000, 1 2800., 7 9788456080286536D-1, RMIL0019
2      3 9894228040143268D-1/ RMIL0020
C      RMIL0021
C      TRIVIAL CASES
C      IF (X NE ZERO) GO TO 10
C      RMILLS = FPI1
C      RETURN
C      10 IF (X GT UTZERO) GO TO 20
C      RMILLS = ZERO
C      RETURN
C      USUAL SITUATION
C      20 UP = TRUE.
C      Z = X
C      IF (Z GE ZERO) GO TO 30
C      UP = FALSE
C      Z = -Z
C      30 Y = HALF * Z * Z
C      IF (Z GT CON) GO TO 50
C      CENTRAL PORTION - O LT ABS(X) LT 1 28
C      RMILLS = Z * (O 39894228044DO-O 399903438504DO*(Y/(Y + 5
1/5885480458DO-29 8213557808DO/(Y + 2 62433121679DO+48
2695930692DO/(Y + 5 92885724438DO))))
C      Y = FPI2 * DEXP(-Y)
C      IF (UP) GO TO 40
C      RMILLS = Y / (HALF + RMILLS)
C      RETURN
C      40 RMILLS = Y / (HALF - RMILLS)
C      RETURN
C      OUTER PORTION - ABS(X) GE 1 28
C      50 IF (UP OR Z LE LTONE) GO TO 60
C      SPECIAL CASE - LOWER TAIL AND Q ESSENTIALLY 1
C      RMILLS = FPI2 * DEXP(-Y)
C      RETURN
C      USUAL SITUATION
C      60 RMILLS = (Z - 3 8052D 8 + 1 00000615302DO/(Z + 3 98064794D-4 + 1
198615381364DO/(Z - O 151679116635DO*5 29330324926DO/(Z + 4
28385912808DO 15 1508972451DO/(Z + O 742380924027DO+30 789933034DO/
3(Z + 3 99019417011DO))))))
C      IF (UP) RETURN
C      Y = FPI2 * DEXP( Y )
C      RMILLS = Y / (ONE - Y/RMILLS)
C      RETURN
C      ENO

```

```

C.....MAIN0001
C   SAMPLE MAIN PROGRAM AND INPUT SUBROUTINE FOR SELF-CRITICAL
C   BINARY ESTIMATION. IT IS RECOMMENDED THAT BINARY() AND
C   ITS AUXILIARY PROCEDURES BE COMPILED AND PLACED IN AN
C   OBJECT CODE LIBRARY, SO THE USER CAN CALL BINARY() FROM
C   ARBITRARY PROGRAMS.
C.....MAIN0002
C   INTEGER N, IX, IA(750), NPAR, ISUB(10), ISTART(10), IDEP, IDIST,
C   1   MAXIT, IPRINT, IFLAG, ICOV, LMEM, MEMORY(490), IFAULT, NC,
C   2   NMAX, NCMAX, IREAD
C   REAL X(10,750)
C   DOUBLE PRECISION C(5), RELTOL, ABSTOL, BETA(10), XLOGL, COV(10,10)
C.....MAIN0003
C   DIMENSION SPECIFICATIONS FOR ARRAYS.
C   THIS MAIN PROGRAM CAN HANDLE UP TO 750 OBSERVATIONS,
C   WITH UP TO 10 PARAMETERS AND 5 DIFFERENT VALUES OF
C   C FOR ESTIMATION.
C   DATA IX, ICOV, LMEM, NMAX, NCMAX /10, 10, 490, 750, 5/
C.....MAIN0004
C   BASIC INFORMATION TO PASS TO BINARY() - IPRINT STANDS
C   FOR LOGICAL OUTPUT UNIT 6
C   DATA MAXIT, RELTOL, ABSTOL, IPRINT, IFLAG /15, 1.0D-7, 1.0D-7, 6,
C   1   0/
C   IREAD IS LOGICAL INPUT UNIT 5
C   DATA IREAD /5/
C.....MAIN0005
C   SPECIFY EVERYTHING BY HAND IN SUBROUTINE INPUT - ARRAY
C   MEMORY(*) PASSED AS WORKSPACE
C   CALL INPUT(IX, N, NMAX, X, IA, NPAR, BETA, ISUB, ISTART, NC,
C   1   NCMAX, C, IDEP, IDIST, MEMORY, IREAD, IPRINT, IFAULT)
C   IF (IFAUULT EQ 0) GO TO 10
C   IF (IFAUULT EQ 1) WRITE (IPRINT,50) N, NMAX
C   IF (IFAUULT EQ 2) WRITE (IPRINT,60) NPAR, IX
C   IF (IFAUULT EQ 3) WRITE (IPRINT,70) NC, NCMAX
C   WRITE (IPRINT,80)
C   STOP
C.....MAIN0006
C   LOOP OVER THE VALUES OF C REQUESTED FOR ESTIMATION
C   10 DO 40 I = 1, NC
C       CALL BINARY(N, IX, X, IA, NPAR, ISUB, ISTART, IDEP, IDIST, C(I),
C   1   RELTOL, ABSTOL, MAXIT, IPRINT, IFLAG, BETA, XLOGL, ICOV,
C   2   COV, LMEM, MEMORY, IFAULT)
C       IF (IFAUULT EQ 0) GO TO 20
C       WRITE (IPRINT,90) IFAULT
C       STOP
C   20 IF (I GT. 1) GO TO 40
C.....MAIN0007
C   AFTER THE FIRST ESTIMATION, SET ISTART(*) TO 1, SO
C   LATEST ESTIMATES CAN BE USED AS STARTING VALUES
C   DO 30 J = 1, NPAR
C   30 ISTART(J) = 1
C   40 CONTINUE
C.....MAIN0008
C   50 FORMAT ('OERROR - SAMPLE SIZE OF ', I10, ' EXCEEDS DIMENSION OF ',
C   1   I10)
C   60 FORMAT ('OERROR - ', I3, ' PARAMETERS EXCEEDS DIMENSION OF ', I3)
C   70 FORMAT ('OERROR - ', I3, ' C VALUES EXCEEDS DIMENSION OF ', I3)
C   80 FORMAT ('ORECOMPIL MAIN PROGRAM WITH NEW DIMENSIONS')

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90 FORMAT ('ESTIMATION PROCEDURE HAS FAILED - IFAULT =', I6)
STOP
END
C.....
C CRUDE INPUT ROUTINE SETS ESTIMATION CONTROL PARAMETERS
C EXPLICITLY, INSTEAD OF READING THEM IN
C.....
C SUBROUTINE INPUT(IX, N, NMAX, X, IA, NPAR, BETA, ISUB, ISTART, NC, INPU0005
1 NCMAX, C, IDEP, IDIST, WORK, IREAD, IPRINT, IFAULT)
C
C ARGUMENTS
C INTEGER IX, N, NMAX, IA(1), NPAR, ISUB(1), ISTART(1), NC, NCMAX,
1 IDEP, IDIST, IPRINT, IFAULT
C REAL X(IX,1), WORK(1)
C DOUBLE PRECISION BETA(1), C(1)
C LOCAL TYPE DECLARATION
C LOGICAL FLAG
C
C SET DETAILS OF PROBLEM SIZE
C N = 680
C NPAR = 5
C NC = 4
C
C CHECK FOR SIZE ERRORS
C IFAULT = 1
C IF (N GT NMAX) RETURN
C IFAULT = 2
C IF (NPAR GT IX) RETURN
C IFAULT = 3
C IF (NC GT NCMAX) RETURN
C IFAULT = 0
C
C MODELING DETAILS
C DEPENDENT VARIABLE IN FIRST ROW
C IDEP = 1
C CONSTANT IN SECOND ROW
C ICONST = 2
C GAUSSIAN TOLERANCE DISTRIBUTION
C IDIST = 2
C VALUES OF C FOR SELF-CRITICAL
C C(1) = 0.000
C C(2) = 0.100
C C(3) = 0.200
C C(4) = 0.300
C
C INITIALIZE AVERAGES TO 0
C YBAR = 0.0
C YSE = 0.0
C DO 10 I = 1, NPAR
C WORK(I) = 0.0
C
C LOOP OVER SAMPLE TO READ IN DATA
C FLAG = NPAR LE 2
C DO 40 I = 1, N
C READ (IREAD,20) VC, IA(I), ALPHA, S
C FORMAT (F10.0, 12, 2F8.0)
C
C EXPRESS S IN METERS, SCALE VC
C VC = ALOG(VC/1000.0)
C S = 0.3048 * EXP(S)
C

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C      C      FILL THE DATA MATRIX - CONSTANT IN SECOND ROW
      X(IDEP,1) = VC
      X(ICONST,1) = 1.0
      X(3,1) = ALPHA
      X(4,1) = S
      X(5,1) = ALPHA * S

C      C      UPDATE MEAN AND STD ERROR OF DEPENDENT VARIABLE IN A
C      C      WAY WHICH DOESN'T LOSE SIGNIFICANT FIGURES
      TEMP = VC - YBAR
      YBAR = YBAR + TEMP / FLOAT(1)
      YSE = YSE + TEMP * (VC - YBAR)

C      C      INCREMENT COVARIATE AVERAGES (3D ROW AND UP)
      IF (FLAG) GO TO 40
      DO 30 J = 3, NPAR
      30  WORK(J) = WORK(J) + X(J,1)
C      C      40 CONTINUE

C      C      SET ISUB(*) AND ISTART(*), PROVIDING STARTING VALUES
C      C      FOR INTERCEPT AND SCALE
      DO 50 I = 1, NPAR
      50  ISUB(I) = 1
      50  ISTART(I) = 0
      50  CONTINUE
      TEMP = FLOAT(N)
      YSE = SORT(YSE/TEMP)
      ISTART(IDEP) = 1
      BETA(IDEP) = DBLE(YSE)
      ISTART(ICONST) = 1
      BETA(ICONST) = DBLE(YBAR)

C      C      WRITE OUT INFORMATION ON STD. ERROR AND AVERAGES
      WRITE (IPRINT,60) YBAR, YSE
      60  FORMAT ('DEPENDENT (STRESS) VARIABLE'/'MEAN =', E14.6,
      1  ', STANDARD DEVIATION =', E14.6)
      IF (FLAG) RETURN
      WRITE (IPRINT,70)
      70  FORMAT ('COVARIATES HAVE BEEN CENTERED BY THEIR MEANS'/'
      1  'OVARIBLE', 10X, 'MEAN')
      DO 90 I = 3, NPAR
      90  WORK(I) = WORK(I) / TEMP
      WRITE (IPRINT,80) ISUB(1), WORK(1)
      80  FORMAT ('O', 18, E14.6)
      90  CONTINUE

C      C      SUBTRACT MEANS FROM COVARIATES
      DO 110 I = 1, N
      110  X(J,I) = X(J,I) - WORK(J)
      100  X(J,I) = X(J,I) - WORK(J)
      110  CONTINUE
C      C      RETURN
      END

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INPU0058
 INPU0059
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20. ABSTRACT (Continue on reverse side if necessary and identify by block number) This paper describes the computational algorithms for computing generalized likelihood estimators for parametric proportional hazards models.		

END

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DTIC