G03ACF – 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

G03ACF performs a canonical variate (canonical discrimination) analysis.

2 Specification

```
SUBROUTINE GO3ACF(WEIGHT, N, M, X, LDX, ISX, NX, ING, NG, WT, NIG,1CVM, LDCVM, E, LDE, NCV, CVX, LDCVX, TOL,2IRANKX, WK, IWK, IFAIL)INTEGERN, M, LDX, ISX(M), NX, ING(N), NG, NIG(NG),1LDCVM, LDE, NCV, LDCVX, IRANKX, IWK, IFAILrealX(LDX,M), WT(*), CVM(LDCVM,NX), E(LDE,6),1CVX(LDCVX,NG-1), TOL, WK(IWK)CHARACTER*1WEIGHT
```

3 Description

Let a sample of n observations on n_x variables in a data matrix come from n_g groups with $n_1, n_2, \ldots, n_{n_g}$ observations in each group, $\sum n_i = n$. Canonical variate analysis finds the linear combination of the n_x variables that maximizes the ratio of between-group to within-group variation. The variables formed, the canonical variates, can be used to discriminate between groups.

The canonical variates can be calculated from the eigenvectors of the within-group sums of squares and cross-products matrix. However, G03ACF calculates the canonical variates by means of a singular value decomposition (SVD) of a matrix V. Let the data matrix with variable (column) means subtracted be X and let its rank be k; then the k by $(n_q - 1)$ matrix V is given by:

 $V = Q_X^T Q_g$, where Q_g is an n by $(n_g - 1)$ orthogonal matrix that defines the groups and Q_X is the first k rows of the orthogonal matrix Q either from the QR decomposition of X:

$$X = QR$$

if X is of full column rank, i.e., $k = n_x$, else from the SVD of X:

$$X = QDP^T$$
.

Let the SVD of V be:

$$V = U_x \Delta U_a^T$$

then the non-zero elements of the diagonal matrix Δ , δ_i , for i = 1, 2, ..., l, are the *l* canonical correlations associated with the *l* canonical variates, where $l = \min(k, n_q)$.

The eigenvalues, λ_i^2 , of the within-group sums of squares matrix are given by:

$$\lambda_i^2 = \frac{\delta_i^2}{1 - \delta_i^2}.$$

and the value of $\pi_i = \lambda_i^2 / \sum \lambda_i^2$ gives the proportion of variation explained by the *i*th canonical variate. The values of the π_i 's give an indication as to how many canonical variates are needed to adequately describe the data, i.e., the dimensionality of the problem.

To test for a significant dimensionality greater than i the χ^2 statistic:

$$(n-1-n_g - \frac{1}{2}(k-n_g)) \sum_{j=i+1}^l \log(1+\lambda_j^2)$$

[NP3390/19/pdf]

G03ACF.1

can be used. This is asymptotically distributed as a χ^2 distribution with $(k-i)(n_g-1-i)$ degrees of freedom. If the test for i = h is not significant, then the remaining tests for i > h should be ignored.

The loadings for the canonical variates are calculated from the matrix U_r . This matrix is scaled so that the canonical variates have unit within group variance.

In addition to the canonical variates loadings the means for each canonical variate are calculated for each group.

Weights can be used with the analysis, in which case the weighted means are subtracted from each column and then each row is scaled by an amount $\sqrt{w_i}$, where w_i is the weight for the *i*th observation (row).

4 References

- [1] Chatfield C and Collins A J (1980) Introduction to Multivariate Analysis Chapman and Hall
- Gnanadesikan R (1977) Methods for Statistical Data Analysis of Multivariate Observations Wiley $[\mathbf{2}]$
- [3] Hammarling S (1985) The singular value decomposition in multivariate statistics SIGNUM Newsl. **20 (3)** 2–25
- [4] Kendall M G and Stuart A (1969) The Advanced Theory of Statistics (Volume 1) Griffin (3rd Edition)

Parameters 5

WEIGHT — CHARACTER*1 1:

On entry: indicates if weights are to be used.

If WEIGHT = 'U', no weights are used. If WEIGHT = W or 'V', weights are used and must be supplied in WT.

Constraint: WEIGHT = 'U', 'W' or 'V'.

In the case of WEIGHT = 'W', the weights are treated as frequencies and the effective number of observations is the sum of the weights. If WEIGHT = V', the weights are treated as being inversely proportional to the variance of the observations and the effective number of observations is the number of observations with non-zero weights.

N — INTEGER Input On entry: the number of observations, n. Constraint: $N \ge NX + NG$. M — INTEGER 3: On entry: the total number of variables, m. Constraint: M > NX. X(LDX,M) - real array 4: Input

On entry: X(i,j) must contain the *i*th observation for the *j*th variable, for i = 1, 2, ..., n; $j=1,2,\ldots,m.$

LDX — INTEGER 5:

> On entry: the first dimension of the array X as declared in the (sub)program from which G03ACF is called.

Constraint: $LDX \ge N$.

2:

Input

Input

Input

6: ISX(M) - INTEGER array

On entry: ISX(j) indicates whether or not the *j*th variable is to be included in the analysis.

If ISX(j) > 0, then the variables contained in the *j*th column of X is included in the canonical variate analysis, for $j = 1, 2, \ldots, m$.

Constraint: ISX(j) > 0 for NX values of j.

NX — INTEGER 7:

On entry: the number of variables in the analysis, n_{x} .

Constraint: $NX \ge 1$.

ING(N) - INTEGER array 8:

> On entry: ING(i) indicates which group the *i*th observation is in, for i = 1, 2, ..., n. The effective number of groups is the number of groups with non-zero membership.

Constraint: $1 \leq \text{ING}(i) \leq \text{NG}$, for $i = 1, 2, \dots, n$.

NG — INTEGER 9:

On entry: the number of groups, n_q .

Constraint: $NG \ge 2$.

10: WT(*) — *real* array

Note: the dimension of the array WT must be at least N if WEIGHT = 'W' or 'V' and 1 otherwise. On entry: if WEIGHT = 'W' or 'V', then the first n elements of WT must contain the weights to be used in the analysis.

If WT(i) = 0.0, then the *i*th observation is not included in the analysis.

If WEIGHT = 'U', then WT is not referenced.

Constraints:

WT(i)
$$\geq 0.0$$
, for $i = 1, 2, ..., n$,

$$\sum_{1}^{n} WT(i) \geq NX + \text{effective number of groups.}$$

11: NIG(NG) — INTEGER array

On exit: NIG(j) gives the number of observations in group j, for $j = 1, 2, ..., n_a$.

12: CVM(LDCVM,NX) — *real* array

On exit: CVM(i, j) contains the mean of the *j*th canonical variate for the *i*th group, for i = 0 $1, 2, \ldots, n_a; j = 1, 2, \ldots, l;$ the remaining columns, if any, are used as workspace.

13: LDCVM — INTEGER

On entry: the dimension of the array CVM as declared in the (sub)program from which G03ACF is called.

Constraint: LDCVM \geq NG.

Input

Input

Input

Input

Input

Output

Output

Input

On exit: the statistics of the canonical variate analysis.

- E(i,1), the canonical correlations, δ_i , for i = 1, 2, ..., l.
- E(i,2), the eigenvalues of the within-group sum of squares matrix, λ_i^2 , for i = 1, 2, ..., l.
- E(i,3), the proportion of variation explained by the *i*th canonical variate, for i = 1, 2, ..., l.
- E(i,4), the χ^2 statistic for the *i*th canonical variate, for i = 1, 2, ..., l.
- E(i,5), the degrees of freedom for χ^2 statistic for the *i*th canonical variate, for i = 1, 2, ..., l.

E(i,6), the significance level for the χ^2 statistic for the *i*th canonical variate, for i = 1, 2, ..., l.

15: LDE — INTEGER

On entry: the first dimension of the array E as declared in the (sub)program from which G03ACF is called.

Constraint: LDE $\geq \min(NX, NG-1)$.

16: NCV — INTEGER

On exit: the number of canonical variates, l. This will be the minimum of $n_q - 1$ and the rank of X.

17: CVX(LDCVX,NG-1) — *real* array

On exit: the canonical variate loadings. CVX(i, j) contains the loading coefficient for the *i*th variable on the *j*th canonical variate, for $i = 1, 2, ..., n_x$; j = 1, 2, ..., l; the remaining columns, if any, are used as workspace.

18: LDCVX — INTEGER

On entry: the first dimension of the array CVX as declared in the (sub)program from which G03ACF is called.

Constraint: LDCVX \geq NX.

19: TOL – *real*

On entry: the value of TOL is used to decide if the variables are of full rank and, if not, what is the rank of the variables. The smaller the value of TOL the stricter the criterion for selecting the singular value decomposition. If a non-negative value of TOL less than *machine precision* is entered, then the square root of *machine precision* is used instead.

Constraint: TOL ≥ 0.0 .

20: IRANKX — INTEGER

On exit: the rank of the dependent variables.

If the variables are of full rank then IRANKX = NX.

If the variables are not of full rank then IRANKX is an estimate of the rank of the dependent variables. IRANKX is calculated as the number of singular values greater than TOL \times (largest singular value).

21: WK(IWK) — real array

22: IWK — INTEGER

 $On\ entry:$ the dimension of the array WK as declared in the (sub)program from which G03ACF is called.

Constraints:

if $NX \ge NG-1$, then $IWK \ge N \times NX + \max(5 \times (NX-1) + (NX+1) \times NX, N)$, if NX < NG-1, then $IWK \ge N \times NX + \max(5 \times (NX-1) + (NG-1) \times NX, N)$. Output

Input

Output

Output

Input

Input

Work space

Input

Output

23: IFAIL — INTEGER

Input/Output

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:

IFAIL
$$= 1$$

IFAIL
$$= 2$$

On entry, WEIGHT = 'W' or 'V' and a value of WT < 0.0.

IFAIL = 3

On entry, a value of ING < 1, or a value of ING > NG.

IFAIL = 4

On entry, the number of variables to be included in the analysis as indicated by ISX is not equal to NX.

IFAIL = 5

A singular value decomposition has failed to converge. This is an unlikely error exit.

IFAIL = 6

A canonical correlation is equal to 1. This will happen if the variables provide an exact indication as to which group every observation is allocated.

IFAIL = 7

- On entry, less than two groups have non-zero membership, i.e., the effective number of groups is less than 2,
 - or the effective number of groups plus the number of variables, NX, is greater than the effective number of observations.

IFAIL = 8

The rank of the variables is 0. This will happen if all the variables are constants.

7 Accuracy

As the computation involves the use of orthogonal matrices and a singular value decomposition rather than the traditional computing of a sum of squares matrix and the use of an eigenvalue decomposition, G03ACF should be less affected by ill conditioned problems.

8 Further Comments

None.

9 Example

A sample of nine observations, each consisting of three variables plus group indicator, is read in. There are three groups. An unweighted canonical variate analysis is performed and the results printed.

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.

```
GO3ACF Example Program Text
*
*
     Mark 18 Revised. NAG Copyright 1997.
      .. Parameters ..
*
     INTEGER
                       NMAX, MMAX, IWKMAX
     PARAMETER
                       (NMAX=9,MMAX=3,IWKMAX=50)
                       NIN, NOUT
     INTEGER
     PARAMETER
                       (NIN=5,NOUT=6)
*
      .. Local Scalars .
     real
                       TOL
      INTEGER
                       I, IFAIL, IRX, J, M, N, NCV, NG, NX
                       WEIGHT
     CHARACTER
      .. Local Arrays ..
     real
                       CVM(MMAX,MMAX), CVX(MMAX,MMAX), E(MMAX,6),
                       WK(IWKMAX), WT(NMAX), X(NMAX,MMAX)
                       ING(NMAX), ISX(2*MMAX), NIG(MMAX)
     INTEGER
      .. External Subroutines ..
     EXTERNAL
                       GO3ACF
      .. Executable Statements ..
     WRITE (NOUT,*) 'GO3ACF Example Program Results'
     Skip heading in data file
     READ (NIN,*)
     READ (NIN,*) N, M, NX, NG, WEIGHT
     IF (N.LE.NMAX .AND. M.LE.MMAX) THEN
         IF (WEIGHT.EQ.'W' .OR. WEIGHT.EQ.'w' .OR. WEIGHT.EQ.'V' .OR.
             WEIGHT.EQ.'v') THEN
     +
            DO 20 I = 1, N
               READ (NIN,*) (X(I,J),J=1,M), WT(I), ING(I)
  20
            CONTINUE
        FLSE
            DO 40 I = 1, N
               READ (NIN, *) (X(I,J), J=1,M), ING(I)
  40
            CONTINUE
        END IF
        READ (5,*) (ISX(J), J=1, M)
        TOL = 0.000001e0
         IFAIL = 0
```

```
CALL GO3ACF(WEIGHT, N, M, X, NMAX, ISX, NX, ING, NG, WT, NIG, CVM, MMAX, E,
     +
                      MMAX, NCV, CVX, MMAX, TOL, IRX, WK, IWKMAX, IFAIL)
*
         WRITE (NOUT, *)
         WRITE (NOUT,99999) 'Rank of X = ', IRX
         WRITE (NOUT,*)
         WRITE (NOUT, *)
                      Eigenvalues Percentage
                                                    CHISQ
                                                                DF
                                                                       SIG'
         'Canonical
     +
         WRITE (NOUT,*) 'Correlations
                                                     Variation'
         DO 60 I = 1, NCV
            WRITE (NOUT,99998) (E(I,J),J=1,6)
   60
         CONTINUE
         WRITE (NOUT,*)
         WRITE (NOUT,*) 'Canonical Coefficients for X'
         DO 80 I = 1, NX
            WRITE (NOUT, 99997) (CVX(I,J), J=1, NCV)
   80
         CONTINUE
         WRITE (NOUT, *)
         WRITE (NOUT, *) 'Canonical variate means'
         DO 100 I = 1, NG
            WRITE (NOUT, 99997) (CVM(I,J), J=1, NCV)
  100
         CONTINUE
      END IF
      STOP
99999 FORMAT (1X,A,I2)
99998 FORMAT (1X,2F12.4,F11.4,F10.4,F8.1,F8.4)
99997 FORMAT (1X,5F9.4)
      END
```

9.2 Program Data

GO3ACF Example Program Data 9 3 3 3 'U' 13.3 10.6 21.2 1 13.6 10.2 21.0 2 14.2 10.7 21.1 3 13.4 9.4 21.0 1 13.2 9.6 20.1 2 13.9 10.4 19.8 3 12.9 10.0 20.5 1 12.2 9.9 20.7 2 13.9 11.0 19.1 3 1 1 1 1

9.3 Program Results

GO3ACF Example Program Results

Rank of X = 3

Canonical	Eigenvalues	Percentage	CHISQ	DF	SIG
Correlations		Variation			
0.8826	3.5238	0.9795	7.9032	6.0	0.2453
0.2623	0.0739	0.0205	0.3564	2.0	0.8368

Canonical Coefficients for X -1.7070 0.7277 -1.3481 0.3138 0.9327 1.2199 Canonical variate means 0.9841 0.2797 1.1805 -0.2632 -2.1646 -0.0164