NAG Library Routine Document

G03ACF

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, CVM, LDCVM, E, LDE, NCV, CVX, LDCVX, TOL, IRANKX, WK, IWK, IFAIL)

INTEGER

N, M, LDX, ISX(M), NX, ING(N), NG, NIG(NG), LDCVM, LDE, NCV, LDCVX, IRANKX, IWK, IFAIL

REAL (KIND=nag_wp) X(LDX,M), WT(*), CVM(LDCVM,NX), E(LDE,6), CVX(LDCVX,NG-1), TOL, WK(IWK)

CHARACTER(1)

WEIGHT
```

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 then 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_g-1) matrix V is given by:

$$V = Q_X^{\mathsf{T}} Q_q$$

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:

$$V = U_x \Delta U_g^{\mathsf{T}}$$

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

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.

Mark 25 G03ACF.1

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

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

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_x . 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

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 Hammarling S (1985) The singular value decomposition in multivariate statistics *SIGNUM Newsl.* **20(3)** 2–25

Kendall M G and Stuart A (1969) The Advanced Theory of Statistics (Volume 1) (3rd Edition) Griffin

5 Parameters

1: WEIGHT – CHARACTER(1)

Input

On entry: indicates if weights are to be used.

WEIGHT = 'U'

No weights are used.

WEIGHT = 'W' or 'V'

Weights are used and must be supplied in WT.

If 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 nonzero weights.

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

2: N – INTEGER Input

On entry: n, the number of observations.

Constraint: $N \ge NX + NG$.

3: M – INTEGER Input

On entry: m, the total number of variables.

Constraint: M > NX.

4: X(LDX, M) – REAL (KIND=nag wp) array Input

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

G03ACF.2 Mark 25

5: LDX – INTEGER

Input

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

Constraint: LDX \geq N.

6: ISX(M) – INTEGER array

Input

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

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

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

7: NX – INTEGER

Input

On entry: the number of variables in the analysis, n_x .

Constraint: $NX \ge 1$.

8: ING(N) - INTEGER array

Input

On entry: ING(i) indicates which group the ith observation is in, for i = 1, 2, ..., n. The effective number of groups is the number of groups with nonzero membership.

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

9: NG – INTEGER

Input

On entry: the number of groups, n_q .

Constraint: $NG \ge 2$.

10: WT(*) - REAL (KIND=nag_wp) array

Input

Note: the dimension of the array WT must be at least N if WEIGHT = 'W' or 'V', and at least 1 otherwise.

On entry: if WEIGHT = 'W' or 'V', the first n elements of WT must contain the weights to be used in the analysis.

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

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

Constraints:

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

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

11: NIG(NG) - INTEGER array

Output

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

12: CVM(LDCVM, NX) - REAL (KIND=nag_wp) array

Output

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

13: LDCVM – INTEGER

Input

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

Constraint: LDCVM \geq NG.

Mark 25

G03ACF.3

14: E(LDE, 6) - REAL (KIND=nag wp) array

Output

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

Input

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

Output

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

17: $CVX(LDCVX, NG - 1) - REAL (KIND=nag_wp)$ array

Output

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 $i = 1, 2, ..., n_x$ and j = 1, 2, ..., l; the remaining columns, if any, are used as workspace.

18: LDCVX – INTEGER

Input

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 (KIND=nag wp)

Input

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, the square root of *machine precision* is used instead.

Constraint: $TOL \ge 0.0$.

20: IRANKX - INTEGER

Output

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 $TOL \times (largest \ singular \ value)$.

G03ACF.4 Mark 25

21: WK(IWK) – REAL (KIND=nag_wp) array

Workspace

22: IWK – INTEGER

Input

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, IWK \ge N \times NX + max(5 \times (NX - 1) + (NX + 1) \times NX, N) + 1; if NX < NG - 1, IWK \ge N \times NX + max(5 \times (NX - 1) + (NG - 1) \times NX, N) + 1.
```

23: IFAIL - INTEGER

Input/Output

On entry: IFAIL must be set to 0, -1 or 1. If you are unfamiliar with this parameter you should refer to Section 3.3 in the Essential Introduction for details.

For environments where it might be inappropriate to halt program execution when an error is detected, the value -1 or 1 is recommended. If the output of error messages is undesirable, then the value 1 is recommended. Otherwise, if you are not familiar with this parameter, the recommended value is 0. When the value -1 or 1 is used it is essential to test the value of IFAIL on exit.

On exit: IFAIL = 0 unless the routine detects an error or a warning has been flagged (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 or warnings detected by the routine:

```
IFAIL = 1
```

```
On entry, NX < 1,
or
         NG < 2,
         M < NX
or
         N < NX + NG,
or
or
         LDX < N,
         LDCVX < NX,
or
         LDCVM < NG,
or
         LDE < min(NX, NG - 1),
or
         NX > NG - 1 and IWK < N \times NX + max(5 \times (NX - 1) + (NX + 1) \times NX, N),
or
         NX < NG - 1 and IWK < N \times NX + max(5 \times (NX - 1) + (NG - 1) \times NX, N),
or
          WEIGHT \neq 'U', 'W' or 'V',
or
         TOL < 0.0.
or
```

IFAIL = 2

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

 $\mathrm{IFAIL} = 3$

```
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.

Mark 25 G03ACF.5

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 nonzero 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.

IFAIL = -99

An unexpected error has been triggered by this routine. Please contact NAG.

See Section 3.8 in the Essential Introduction for further information.

IFAIL = -399

Your licence key may have expired or may not have been installed correctly.

See Section 3.7 in the Essential Introduction for further information.

IFAIL = -999

Dynamic memory allocation failed.

See Section 3.6 in the Essential Introduction for further information.

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 Parallelism and Performance

G03ACF is threaded by NAG for parallel execution in multithreaded implementations of the NAG Library.

G03ACF makes calls to BLAS and/or LAPACK routines, which may be threaded within the vendor library used by this implementation. Consult the documentation for the vendor library for further information.

Please consult the X06 Chapter Introduction for information on how to control and interrogate the OpenMP environment used within this routine. Please also consult the Users' Note for your implementation for any additional implementation-specific information.

9 Further Comments

None.

10 Example

This example uses a sample of nine observations, each consisting of three variables plus a group indicator. There are three groups. An unweighted canonical variate analysis is performed and the results printed.

G03ACF.6 Mark 25

10.1 Program Text

```
Program g03acfe
      GO3ACF Example Program Text
!
1
      Mark 25 Release. NAG Copyright 2014.
!
      .. Use Statements ..
      Use nag_library, Only: g03acf, nag_wp, x04caf
!
      .. Implicit None Statement ..
      Implicit None
!
      .. Parameters ..
                                        :: nin = 5, nout = 6
      Integer, Parameter
      .. Local Scalars ..
!
      Real (Kind=nag_wp)
                                        :: tol
                                        :: i, ifail, irankx, iwk, ldcvm, ldcvx, &
    lde, ldx, lwt, m, n, ncv, ng, nx
      Integer
      Character (1)
                                         :: weight
!
      .. Local Arrays ..
      Real (Kind=nag_wp), Allocatable :: cvm(:,:), cvx(:,:), e(:,:), wk(:),
                                            wt(:), x(:,:)
                                        :: ing(:), isx(:), nig(:)
      Integer, Allocatable
!
      .. Intrinsic Procedures ..
      Intrinsic
                                        :: max, min
      .. Executable Statements ..
      Write (nout,*) 'GO3ACF Example Program Results'
      Write (nout,*)
!
      Skip heading in data file
      Read (nin,*)
!
      Read in the problem size
      Read (nin,*) n, m, nx, ng, weight
      Select Case (weight)
      Case ('W','w','V','v')
       lwt = n
      Case Default
       lwt = 0
      End Select
      ldx = n
      ldcvm = nq
      lde = min(nx,ng-1)
      ldcvx = nx
      If (nx>=ng-1) Then
        iwk = n*nx + max(5*(nx-1)+(nx+1)*nx,n) + 1
      Else
        iwk = n*nx + max(5*(nx-1)+(ng-1)*nx,n) + 1
      End If
      Allocate (x(ldx,m),isx(m),ing(n),wt(lwt),nig(ng),cvm(ldcvm,nx),e(lde,6), &
        cvx(ldcvx,nx),wk(iwk))
      Read in data
!
      If (lwt>0) Then
       Read (nin,*)(x(i,1:m),wt(i),ing(i),i=1,n)
        Read (nin,*)(x(i,1:m),ing(i),i=1,n)
      End If
      Read in variable inclusion flags
      Read (nin,*) isx(1:m)
!
      Use default tolerance
      tol = 0.0E0_nag_wp
      Perform canonical variate analysis
      ifail = 0
      Call g03acf(weight,n,m,x,ldx,isx,nx,ing,ng,wt,nig,cvm,ldcvm,e,lde,ncv, &
        cvx,ldcvx,tol,irankx,wk,iwk,ifail)
```

Mark 25 G03ACF.7

```
Display results
      Write (nout, 99999) 'Rank of X = ', irankx
      Write (nout,*)
      Write (nout,*) &
      'Canonical Eigenvalues Percentage Write (nout,*) 'Correlations
                                                                        SIG'
                                                    CHISQ
                                                                DF
                                                   Variation'
      Write (nout, 99998)(e(i, 1:6), i=1, ncv)
      Write (nout,*)
      Flush (nout)
      ifail = 0
      Call x04caf('General',' ',nx,ncv,cvx,ldcvx, &
        'Canonical Coefficients for X', ifail)
      Write (nout,*)
      Flush (nout)
      ifail = 0
      Call x04caf('General',' ',ng,ncv,cvm,ldcvm,'Canonical variate means', &
        ifail)
99999 Format (1X,A,IO)
99998 Format (1X,2F12.4,F11.4,F10.4,F8.1,F8.4)
    End Program g03acfe
```

10.2 Program Data

```
G03ACF 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
```

10.3 Program Results

```
GO3ACF Example Program Results
Rank of X = 3
Canonical Eigenvalues Percentage
                                        CHISQ
                                                  DF
                                                          SIG
Correlations
                          Variation
                                       7.9032
      0.8826
                  3.5238
                            0.9795
                                                  6.0 0.2453
                                                  2.0 0.8368
      0.2623
                  0.0739
                             0.0205
                                       0.3564
Canonical Coefficients for X
            1
1
      -1.7070
                  0.7277
2
      -1.3481
                  0.3138
                  1.2199
      0.9327
3
Canonical variate means
          1
1
       0.9841
                 0.2797
      1.1805
                 -0.2632
2
3
      -2.1646
                 -0.0164
```

G03ACF.8 (last)

Mark 25