NAG Library Function Document

nag mv canon var (g03acc)

1 Purpose

nag mv canon var (g03acc) performs a canonical variate (canonical discrimination) analysis.

2 Specification

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, nag_mv_canon_var (g03acc) 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:

$$X = QDP^{\mathsf{T}}.$$

Let the SVD of V be:

$$V = U_x \Delta U_q^{\mathrm{T}}$$

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 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.

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

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

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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 (1979) The Advanced Theory of Statistics (3 Volumes) (4th Edition) Griffin

5 Arguments

1: **weight** – Nag Weightstype

Input

On entry: indicates the type of weights to be used in the analysis.

weight = Nag_NoWeights

No weights are used.

weight = Nag_Weightsfreq

The weights are treated as frequencies and the effective number of observations is the sum of the weights.

weight = Nag_Weightsvar

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 = Nag_NoWeights, Nag_Weightsfreq or Nag_Weightsvar.

2: **n** – Integer Input

On entry: the number of observations, n.

Constraint: $\mathbf{n} \geq \mathbf{n}\mathbf{x} + \mathbf{n}\mathbf{g}$.

3: \mathbf{m} - Integer Input

On entry: the total number of variables, m.

Constraint: $m \ge nx$.

4: $\mathbf{x}[\mathbf{n} \times \mathbf{tdx}]$ - const double Input

On entry: $\mathbf{x}[(i-1) \times \mathbf{tdx} + j - 1]$ must contain the *i*th observation for the *j*th variable, for i = 1, 2, ..., n and j = 1, 2, ..., m.

5: \mathbf{tdx} - Integer Input

On entry: the stride separating matrix column elements in the array \mathbf{x} .

Constraint: $\mathbf{tdx} \geq \mathbf{m}$.

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6: isx[m] – const Integer

Input

On entry: isx[j-1] indicates whether or not the jth variable is to be included in the analysis.

If $\mathbf{isx}[j-1] > 0$, then the variable contained in the *j*th column of \mathbf{x} is included in the canonical variate analysis, for j = 1, 2, ..., m.

Constraint: $\mathbf{isx}[j-1] > 0$ for \mathbf{nx} values of j.

7: \mathbf{nx} – Integer

Input

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

Constraint: $\mathbf{nx} \geq 1$.

8: ing[n] – const Integer

Input

On entry: ing[i-1] indicates which group the *i*th observation is in, for $i=1,2,\ldots,n$. The effective number of groups is the number of groups with nonzero membership.

Constraint: $1 \leq ing[i-1] \leq ng$, for i = 1, 2, ..., n.

9: **ng** – Integer

Input

On entry: the number of groups, n_g .

Constraint: $ng \ge 2$.

10: $\mathbf{wt}[\mathbf{n}]$ – const double

Input

On entry: if **weight** = Nag_Weightsfreq or Nag_Weightsvar then the elements of **wt** must contain the weights to be used in the analysis.

If $\mathbf{wt}[i-1] = 0.0$ then the *i*th observation is not included in the analysis.

Constraints:

 $\mathbf{wt}[i-1] \ge 0.0$, for $i = 1, 2, \dots, n$; $\sum_{i=1}^{n} \mathbf{wt}[i-1] \ge \mathbf{nx} + \text{ effective number of groups.}$

Note: if **weight** = Nag_NoWeights then **wt** is not referenced and may be **NULL**..

11: **nig**[**ng**] – Integer

Output

On exit: $\mathbf{nig}[j-1]$ gives the number of observations in group j, for $j=1,2,\ldots,n_q$.

12: $\operatorname{cvm}[\operatorname{ng} \times \operatorname{tdcvm}] - \operatorname{double}$

Output

On exit: $\mathbf{cvm}[(i-1) \times \mathbf{tdcvm} + j-1]$ 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: **tdcvm** – Integer

Input

On entry: the stride separating matrix column elements in the array cvm.

Constraint: $tdcvm \ge nx$.

14: $\mathbf{e}[\mathbf{min}(\mathbf{nx}, \mathbf{ng} - \mathbf{1}) \times \mathbf{tde}] - \text{double}$

Output

On exit: the statistics of the canonical variate analysis. $\mathbf{e}[(i-1) \times \mathbf{tde}]$, the canonical correlations, δ_i , for $i=1,2,\ldots,l$.

 $\mathbf{e}[(i-1) \times \mathbf{tde} + 1]$, the eigenvalues of the within-group sum of squares matrix, λ_i^2 , for i = 1, 2, ..., l.

 $\mathbf{e}[(i-1) \times \mathbf{tde} + 2]$, the proportion of variation explained by the *i*th canonical variate, for $i = 1, 2, \dots, l$.

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 $\mathbf{e}[(i-1) \times \mathbf{tde} + 3]$, the χ^2 statistic for the *i*th canonical variate, for $i = 1, 2, \dots, l$.

 $e[(i-1) \times tde + 4]$, the degrees of freedom for χ^2 statistic for the *i*th canonical variate, for i = 1, 2, ..., l.

 $\mathbf{e}[(i-1) \times \mathbf{tde} + 5]$, the significance level for the χ^2 statistic for the *i*th canonical variate, for $i = 1, 2, \dots, l$.

15: **tde** – Integer Input

On entry: the stride separating matrix column elements in the array e.

Constraint: $tde \geq 6$.

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 \mathbf{x} .

17: $\mathbf{cvx}[\mathbf{nx} \times \mathbf{tdcvx}] - \mathbf{double}$

Output

On exit: the canonical variate loadings. $\mathbf{cvx}[(i-1) \times \mathbf{tdcvx} + j - 1]$ contains the loading coefficient for the *i*th variable on the *j*th canonical variate, for $i = 1, 2, \dots, n_x$ and $j = 1, 2, \dots, l$; the remaining columns, if any, are used as workspace.

18: **tdcvx** – Integer Input

On entry: the stride separating matrix column elements in the array cvx.

Constraint: $tdcvx \ge ng - 1$.

19: **tol** – double *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, then 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**×(largest singular value).

21: **fail** – NagError *

Input/Output

The NAG error argument (see Section 3.6 in the Essential Introduction).

6 Error Indicators and Warnings

NE 2 INT ARG LT

On entry, $\mathbf{m} = \langle value \rangle$ while $\mathbf{n}\mathbf{x} = \langle value \rangle$. These arguments must satisfy $\mathbf{m} \geq \mathbf{n}\mathbf{x}$.

On entry, $tdcvm = \langle value \rangle$ while $nx = \langle value \rangle$. These arguments must satisfy $tdcvm \geq nx$.

On entry, $\mathbf{tdcvx} = \langle value \rangle$ while $\mathbf{ng} = \langle value \rangle$. These arguments must satisfy $\mathbf{tdcvx} \geq \mathbf{ng} - 1$.

On entry, $\mathbf{tdx} = \langle value \rangle$ while $\mathbf{m} = \langle value \rangle$. These arguments must satisfy $\mathbf{tdx} \geq \mathbf{m}$.

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NE 3 INT ARG CONS

On entry, $\mathbf{n} = \langle value \rangle$, $\mathbf{n}\mathbf{x} = \langle value \rangle$ and $\mathbf{n}\mathbf{g} = \langle value \rangle$. These arguments must satisfy $\mathbf{n} \geq \mathbf{n}\mathbf{x} + \mathbf{n}\mathbf{g}$.

NE ALLOC FAIL

Dynamic memory allocation failed.

NE BAD PARAM

On entry, argument weight had an illegal value.

NE_CANON_CORR_1

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

NE_GROUPS

Either the effective number of groups is less than two or the effective number of groups plus the number of variables, $\mathbf{n}\mathbf{x}$ is greater than the effective number of observations.

NE_INT_ARG_LT

```
On entry, \mathbf{ng} = \langle value \rangle.
Constraint: \mathbf{ng} \geq 2.
On entry, \mathbf{nx} = \langle value \rangle.
Constraint: \mathbf{nx} \geq 1.
On entry, \mathbf{tde} = \langle value \rangle.
Constraint: \mathbf{tde} > 6.
```

NE_INTARR_INT

```
On entry, \mathbf{ing}[\langle value \rangle] = \langle value \rangle, \mathbf{ng} = \langle value \rangle. Constraint: 1 \leq \mathbf{ing}[i-1] \leq \mathbf{ng}, for i = 1, 2, \dots, n.
```

NE_INTERNAL_ERROR

An internal error has occurred in this function. Check the function call and any array sizes. If the call is correct then please contact NAG for assistance.

NE NEG WEIGHT ELEMENT

```
On entry, \mathbf{wt}[\langle value \rangle] = \langle value \rangle.
Constraint: When referenced, all elements of \mathbf{wt} must be non-negative.
```

NE_RANK_ZERO

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

NE REAL ARG LT

On entry, **tol** must not be less than 0.0: **tol** = $\langle value \rangle$.

NE SVD NOT CONV

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

NE_VAR_INCL_INDICATED

The number of variables, **nx** in the analysis = $\langle value \rangle$, while number of variables included in the analysis via array **isx** = $\langle value \rangle$.

Constraint: these two numbers must be the same.

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NE WT ARGS

The wt array argument must not be NULL when the weight argument indicates weights.

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, nag my canon var (g03acc) should be less affected by ill conditioned problems.

8 Parallelism and Performance

Not applicable.

9 Further Comments

None.

10 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.

10.1 Program Text

```
/* nag_mv_canon_var (g03acc) Example Program.
 * Copyright 1998 Numerical Algorithms Group.
* Mark 5, 1998.
 * Mark 8 revised, 2004.
#include <nag.h>
#include <stdio.h>
#include <nag_stdlib.h>
#include <nagg03.h>
#define X(I, J)
                     x[(I) *tdx + J]
\#define E(I, J) e[(I) *tde + J]
#define CVM(I, J) cvm[(I) *tdcvm + J]
#define CVX(I, J) cvx[(I) *tdcvx + J]
int main(void)
                     exit_status = 0, i, irx, j, m, n, ncv, ng;
  Integer
                     nx, tdcvm, tdcvx, tde, tdx;
  Integer
                     *ing = 0, *isx = 0, *nig = 0;
*cvm = 0, *cvx = 0, *e = 0, tol, *wt = 0, *x = 0;
  Integer
  double
                     nag_enum_arg[40];
  char
  Nag_Weightstype weight;
  NagError
                     fail:
  INIT_FAIL(fail);
  printf("nag_mv_canon_var (g03acc) Example Program Results\n\n");
  /* Skip heading in data file */
  scanf("%*[^\n]");
scanf("%1d", &n);
scanf("%1d", &m);
  scanf("%ld", &nx);
  scanf("%ld", &ng);
scanf("%39s", nag_enum_arg);
```

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```
/* nag_enum_name_to_value (x04nac).
 * Converts NAG enum member name to value
weight = (Nag_Weightstype) nag_enum_name_to_value(nag_enum_arg);
if (n \ge nx + ng \&\& m \ge nx)
  {
    if (!(x = NAG_ALLOC(n*m, double)) ||
         !(wt = NAG_ALLOC(n, double)) ||
         !(ing = NAG_ALLOC(n, Integer)) ||
         !(e = NAG_ALLOC((MIN(nx, ng-1))*6, double)) ||
         !(cvm = NAG_ALLOC(ng*nx, double)) ||
         !(cvx = NAG_ALLOC(nx*(ng-1), double)) ||
         !(nig = NAG_ALLOC(ng, Integer)) ||
         !(isx = NAG_ALLOC(m, Integer))
      {
        printf("Allocation failure\n");
        exit_status = -1;
        goto END;
      }
    tdx = m;
    tde = 6;
    tdcvm = nx;
    tdcvx = ng-1;
else
  {
    printf("Invalid n or m.\n");
    exit_status = 1;
    return exit_status;
if (weight == Nag_Weightsfreq || weight == Nag_Weightsvar)
    for (i = 0; i < n; ++i)
        for (j = 0; j < m; ++j)
          scanf("%lf", &X(i, j));
        scanf("%lf", &wt[i]);
scanf("%ld", &ing[i]);
  }
else
    for (i = 0; i < n; ++i)
        for (j = 0; j < m; ++j)
  scanf("%lf", &X(i, j));
scanf("%ld", &ing[i]);</pre>
      }
for (j = 0; j < m; ++j)
  scanf("%ld", &isx[j]);</pre>
tol = 1e-6;
/* nag_mv_canon_var (g03acc).
 * Canonical variate analysis
 * /
nag_mv_canon_var(weight, n, m, x, tdx, isx, nx, ing, ng, wt, nig,
                  cvm, tdcvm, e, tde, &ncv, cvx, tdcvx, tol, &irx, &fail);
if (fail.code != NE_NOERROR)
  {
    printf("Error from nag_mv_canon_var (g03acc).\n%s\n",
             fail.message);
    exit_status = 1;
    goto END;
printf("%s%2ld\n\n", "Rank of x = ", irx);
printf("Canonical Eigenvalues Percentage CHISQ"
```

```
SIG \n");
                   DF
  printf("Correlations
                                         Variation\n");
  for (i = 0; i < ncv; ++i)
    {
      for (j = 0; j < 6; ++j)
       printf("%12.4f", E(i, j));
      printf("\n");
  printf("\nCanonical Coefficients for X\n");
  for (i = 0; i < nx; ++i)
      for (j = 0; j < ncv; ++j)
  printf("%9.4f", CVX(i, j));</pre>
      printf("\n");
  printf("\nCanonical variate means\n");
  for (i = 0; i < ng; ++i)
      for (j = 0; j < ncv; ++j)
        printf("%9.4f", CVM(i, j));
      printf("\n");
END:
 NAG_FREE(x);
  NAG_FREE(wt);
  NAG_FREE(ing);
 NAG_FREE(e);
  NAG_FREE(cvm);
  NAG_FREE(cvx);
  NAG_FREE(nig);
 NAG_FREE(isx);
 return exit_status;
}
```

10.2 Program Data

```
nag_mv_canon_var (g03acc) Example Program Data 9 3 3 3 Nag_NoWeights 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
```

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10.3 Program Results

 $nag_mv_canon_var\ (g03acc)\ Example\ Program\ Results$

Rank of x = 3

Canonical	Eigenvalues	Percentage	CHISQ	DF	SIG
Correlations		Variation			
0.8826	3.5238	0.9795	7.9032	6.0000	0.2453
0.2623	0.0739	0.0205	0.3564	2.0000	0.8368

Canonical Coefficients for ${\tt X}$

-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

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