

## nag\_mv\_fac\_score (g03ccc)

### 1. Purpose

**nag\_mv\_fac\_score (g03ccc)** computes factor score coefficients from the result of fitting a factor analysis model by maximum likelihood as performed by **nag\_mv\_factor (g03cac)**.

### 2. Specification

```
#include <nag.h>
#include <nagg03.h>

void nag_mv_fac_score(Nag_FacScoreMethod method, Nag_FacRotation rotate,
                     Integer nvar, Integer nfac, double fl[], Integer tdf1, double psi[],
                     double e[], double r[], Integer tdr, double fs[], Integer tdfs,
                     NagError *fail)
```

### 3. Description

A factor analysis model aims to account for the covariances among  $p$  variables, observed on  $n$  individuals, in terms of a smaller number,  $k$ , of unobserved variables or factors. The values of the factors for an individual are known as factor scores. **nag\_mv\_factor (g03cac)** fits the factor analysis model by maximum likelihood and returns the estimated factor loading matrix,  $\Lambda$ , and the diagonal matrix of variances of the unique components,  $\Psi$ . To obtain estimates of the factors, a  $p$  by  $k$  matrix of factor score coefficients,  $\Phi$ , is formed. The estimated vector of factor scores,  $\hat{f}$ , is then given by:

$$\hat{f} = x^T \Phi,$$

where  $x$  is the vector of observed variables for an individual.

There are two commonly used methods of obtaining factor score coefficients.

The regression method:

$$\Phi = \Psi^{-1} \Lambda (I + \Lambda^T \Psi^{-1} \Lambda)^{-1},$$

and Bartlett's method:

$$\Phi = \Psi^{-1} \Lambda (\Lambda^T \Psi^{-1} \Lambda)^{-1}.$$

See Lawley and Maxwell (1971) for details of both methods. In the regression method as given above, it is assumed that the factors are not correlated and have unit variance; this is true for models fitted by **nag\_mv\_factor (g03cac)**. Further, for models fitted by **nag\_mv\_factor (g03cac)**,

$$\Lambda^T \Psi^{-1} \Lambda = \Theta - I,$$

where  $\Theta$  is the diagonal matrix of eigenvalues of the matrix  $S^*$ , as described in **nag\_mv\_factor (g03cac)**.

The factors may be orthogonally rotated using an orthogonal rotation matrix,  $R$ , as computed by **nag\_mv\_orthomax (g03bac)**. The factor scores for the rotated matrix are then given by  $\Lambda R$ .

### 4. Parameters

#### method

Input: indicates which method is to be used to compute the factor score coefficients.

If **method = Nag\_FacScoreRegsn**, then the regression method is used.

If **method = Nag\_FacScoreBart**, then Bartlett's method is used.

Constraint: **method = Nag\_FacScoreRegsn** or **Nag\_FacScoreBart**.

**rotate**

Input: indicates whether a rotation is to be applied.

If **rotate** = **Nag\_FacRotate**, then a rotation will be applied to the coefficients and the rotation matrix,  $R$ , must be given in **r**.

If **rotate** = **Nag\_FacNoRotate**, then no rotation is applied.

Constraint: **rotate** = **Nag\_FacRotate** or **Nag\_FacNoRotate**.

**nvar**

Input: the number of observed variables in the factor analysis,  $p$ .

Constraint: **nvar**  $\geq$  **nfac**.

**nfac**

Input: the number of factors in the factor analysis,  $k$ .

Constraint: **nfac**  $\geq$  1.

**fl[nvar][tdfl]**

Input: the matrix of unrotated factor loadings,  $\Lambda$ , as returned by nag\_mv\_factor (g03cac).

**tdfl**

Input: the last dimension of the array **fl** as declared in the calling program.

Constraint: **tdfl**  $\geq$  **nfac**.

**psi[nvar]**

Input: the diagonal elements of  $\Psi$ , as returned by nag\_mv\_factor (g03cac).

Constraint: **psi**[ $i - 1$ ]  $>$  0.0, for  $i = 1, 2, \dots, p$ .

**e[nvar]**

Input: the eigenvalues of the matrix  $S^*$ , as returned by nag\_mv\_factor (g03cac).

Constraint: **e**[ $i - 1$ ]  $>$  1.0, for  $i = 1, 2, \dots, p$ .

**r[nfac][tdr]**

Input: if **rotate** = **Nag\_FacRotate**, then **r** must contain the orthogonal rotation matrix,  $R$ , as returned by nag\_mv\_orthomax (g03bac). If **rotate** = **Nag\_FacNoRotate** then **r** need not be set.

**tdr**

Input: the last dimension of the array **r** as declared in the calling program.

Constraint: if **rotate** = **Nag\_FacRotate** then **tdr**  $\geq$  **nfac**.

**fs[nvar][tdfs]**

Output: the matrix of factor score coefficients,  $\Phi$ . **fs**[ $i - 1$ ][ $j - 1$ ] contains the factor score coefficient for the  $j$ th factor and the  $i$ th observed variable, for  $i = 1, 2, \dots, p$ ;  $j = 1, 2, \dots, k$ .

**tdfs**

Input: the last dimension of the array **fs** as declared in the calling program.

Constraint: **tdfs**  $\geq$  **nfac**.

**fail**

The NAG error parameter, see the Essential Introduction to the NAG C Library.

## 5. Error Indications and Warnings

**NE\_BAD\_PARAM**

On entry, parameter **method** had an illegal value.

On entry, parameter **rotate** had an illegal value.

**NE\_INT\_ARG\_LT**

On entry, **nfac** must not be less than 1: **nfac** =  $\langle value \rangle$ .

**NE\_2.INT\_ARG\_LT**

On entry, **nvar** =  $\langle value \rangle$  while **nfac** =  $\langle value \rangle$ .

These parameters must satisfy **nvar**  $\geq$  **nfac**.

On entry, **tdfl** =  $\langle value \rangle$  while **nfac** =  $\langle value \rangle$ .

These parameters must satisfy **tdfl**  $\geq$  **nfac**.

On entry, **tdfs** =  $\langle value \rangle$  while **nfac** =  $\langle value \rangle$ .

These parameters must satisfy **tdfs**  $\geq$  **nfac**.

**NE\_2.INT\_ARG\_ENUM\_CONS**

On entry, **tdr** =  $\langle value \rangle$  while **nfac** =  $\langle value \rangle$  and **rotate** = **Nag\_FacRotate**.

These parameters must satisfy **tdr**  $\geq$  **nfac** when **rotate** = **Nag\_FacRotate**.

**NE\_REAL\_ARRAY\_INPUT**

On entry, **e**[ $\langle value \rangle$ ] =  $\langle value \rangle$ .

Constraint: **e**[ $\langle value \rangle$ ]  $>$  1.0.

On entry, **psi**[ $\langle value \rangle$ ] =  $\langle value \rangle$ .

Constraint: **psi**[ $\langle value \rangle$ ]  $>$  0.0.

**NE\_ALLOC\_FAIL**

Memory allocation failed.

**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 consult NAG for assistance.

**6. Further Comments**

To compute the factor scores using the factor score coefficients, the values for the observed variables first need to be standardized by subtracting the sample means and, if the factor analysis is based upon a correlation matrix, dividing by the sample standard deviations. This may be performed using `nag_mv_z_scores` (g03zac). The standardized variables are then post-multiplied by the factor score coefficients.

If principal component analysis is required, the routine `nag_mv_prin_comp` (g03aac) computes the principal component scores directly. Hence, the factor score coefficients are not needed.

**6.1. Accuracy**

Accuracy will depend on the accuracy requested when computing the estimated factor loadings using `nag_mv_factor` (g03cac).

**6.2. References**

Lawley D N and Maxwell A E (1971) *Factor Analysis as a Statistical Method* Butterworths (2nd Edition).

**7. See Also**

`nag_mv_canon_corr` (g03adc)

`nag_mv_factor` (g03cac)

`nag_mv_z_scores` (g03zac)

`nag_mv_orthomax` (g03bac)