

NAG Library Routine Document

G03GAF

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

G03GAF performs a mixture of Normals (Gaussians) for a given (co)variance structure.

2 Specification

```
SUBROUTINE G03GAF (N, M, X, LDX, ISX, NVAR, NG, POPT, PROB, LDPROB, NITER,      &
                   RITER, W, G, SOPT, S, LDS, SDS, F, TOL, LOGLIK, IFAIL)

INTEGER             N, M, LDX, ISX(M), NVAR, NG, POPT, LDPROB, NITER,      &
                   RITER, SOPT, LDS, SDS, IFAIL
REAL (KIND=nag_wp) X(LDX,M), PROB(LDPROB,NG), W(NG), G(NVAR,NG),      &
                   S(LDS,SDS,*), F(N,NG), TOL, LOGLIK
```

3 Description

A Normal (Gaussian) mixture model is a weighted sum of k group Normal densities given by,

$$p(x | w, \mu, \Sigma) = \sum_{j=1}^k w_j g(x | \mu_j, \Sigma_j), \quad x \in \mathbb{R}^p$$

where:

x is a p -dimensional object of interest;

w_j is the mixture weight for the j th group and $\sum_{j=1}^k w_j = 1$;

μ_j is p -dimensional vector of means for the j th group;

Σ_j is the covariance structure for the j th group;

$g(\cdot)$ is the p -variate Normal density:

$$g(x | \mu_j, \Sigma_j) = \frac{1}{(2\pi)^{p/2} |\Sigma_j|^{1/2}} \exp \left[-\frac{1}{2} (x - \mu_j) \Sigma_j^{-1} (x - \mu_j)^T \right].$$

Optionally, the covariance structure may be pooled (common to all groups) or calculated for each group, and may be full or diagonal.

4 References

Hartigan J A (1975) *Clustering Algorithms* Wiley

5 Parameters

1: N – INTEGER *Input*

On entry: n , the number of objects. There must be more objects than parameters in the model.

Constraints:

- if $SOPT = 1$, $N > NG \times (NVAR \times NVAR + NVAR)$;
- if $SOPT = 2$, $N > NVAR \times (NG + NVAR)$;

if SOPT = 3, N > 2 × NG × NVAR;
 if SOPT = 4, N > NVAR × (NG + 1);
 if SOPT = 5, N > NVAR × NG + 1.

2: M – INTEGER *Input*

On entry: the total number of variables in array X.

Constraint: M ≥ 1.

3: X(LDX,M) – REAL (KIND=nag_wp) array *Input*

On entry: X(i, j) must contain the value of the j th variable for the i th object, for $i = 1, 2, \dots, N$ and $j = 1, 2, \dots, M$.

4: LDX – INTEGER *Input*

On entry: the first dimension of the array X as declared in the (sub)program from which G03GAF is called.

Constraint: LDX ≥ N.

5: ISX(M) – INTEGER array *Input*

On entry: if NVAR = M all available variables are included in the model and ISX is not referenced; otherwise the j th variable will be included in the analysis if ISX(j) = 1 and excluded if ISX(j) = 0, for $j = 1, 2, \dots, M$.

Constraints:

if NVAR ≠ M, ISX(j) = 1 for NVAR values of j ;
 otherwise 0.

6: NVAR – INTEGER *Input*

On entry: p , the number of variables included in the calculations.

Constraint: 1 ≤ NVAR ≤ M.

7: NG – INTEGER *Input*

On entry: k , the number of groups in the mixture model.

Constraint: NG ≥ 1.

8: POPT – INTEGER *Input*

On entry: if POPT = 1, the initial membership probabilities in PROB are set internally; otherwise these probabilities must be supplied.

9: PROB(LDPROB,NG) – REAL (KIND=nag_wp) array *Input/Output*

On entry: if POPT ≠ 1, PROB(i, j) is the probability that the i th object belongs to the j th group. (These probabilities are normalised internally.)

On exit: PROB(i, j) is the probability of membership of the i th object to the j th group for the fitted model.

10: LDPROB – INTEGER *Input*

On entry: the first dimension of the array PROB as declared in the (sub)program from which G03GAF is called.

Constraint: LDPROB ≥ N.

11: NITER – INTEGER *Input/Output*

On entry: the maximum number of iterations.

Suggested value: 15

On exit: the number of completed iterations.

Constraint: NITER ≥ 1 .

12: RITER – INTEGER *Input*

On entry: if RITER > 0, membership probabilities are rounded to 0.0 or 1.0 after the completion of every RITER iterations.

Suggested value: 5

13: W(NG) – REAL (KIND=nag_wp) array *Output*

On exit: w_j , the mixing probability for the j th group.

14: G(NVAR,NG) – REAL (KIND=nag_wp) array *Output*

On exit: $G(i,j)$ gives the estimated mean of the i th variable in the j th group.

15: SOPT – INTEGER *Input*

On entry: determines the (co)variance structure:

SOPT = 1

Groupwise covariance matrices.

SOPT = 2

Pooled covariance matrix.

SOPT = 3

Groupwise variances.

SOPT = 4

Pooled variances.

SOPT = 5

Overall variance.

Constraint: SOPT = 1, 2, 3, 4 or 5.

16: S(LDS,SDS,*) – REAL (KIND=nag_wp) array *Output*

Note: the last dimension of the array S must be at least NG if SOPT = 1, and at least 1 otherwise.

On exit: if SOPT = 1, $S(i,j,k)$ gives the (i,j) th element of the k th group.

If SOPT = 2, $S(i,j,1)$ gives the (i,j) th element of the pooled covariance.

If SOPT = 3, $S(j,k,1)$ gives the j th variance in the k th group.

If SOPT = 4, $S(j,1,1)$ gives the j th pooled variance.

If SOPT = 5, $S(1,1,1)$ gives the overall variance.

17: LDS – INTEGER *Input*

On entry: the first dimension of the (co)variance structure S.

Constraints:

if SOPT = 5, LDS = 1;
otherwise LDS = NVAR.

18: SDS – INTEGER *Input*

On entry: the second dimension of the (co)variance structure S.

Constraints:

if SOPT = 1 or 2, SDS must be at least NVAR;
 if SOPT = 3, SDS must be at least NG;
 if SOPT = 4 or 5, SDS must be at least 1.

19: F(N,NG) – REAL (KIND=nag_wp) array *Output*

On exit: $F(i,j)$ gives the p -variate Normal (Gaussian) density of the i th object in the j th group.

20: TOL – REAL (KIND=nag_wp) *Input*

On entry: iterations cease the first time an improvement in log-likelihood is less than TOL. If $TOL \leq 0$ a value of 10^{-3} is used.

21: LOGLIK – REAL (KIND=nag_wp) *Output*

On exit: the log-likelihood for the fitted mixture model.

22: 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, $N = \langle value \rangle$ and $p = \langle value \rangle$.

Constraint: $N > p$, the number of parameters, i.e., too few objects have been supplied for the model.

IFAIL = 2

On entry, $M = \langle value \rangle$.

Constraint: $M \geq 1$.

IFAIL = 4

On entry, $LDX = \langle value \rangle$ and $N = \langle value \rangle$.

Constraint: $LDX \geq N$.

IFAIL = 5

On entry, $NVAR = \langle value \rangle$ and $M = \langle value \rangle$.

Constraint: $NVAR \geq 1$ and $NVAR \leq M$.

IFAIL = 6

On entry, $NVAR \neq M$ and ISX is invalid.

IFAIL = 7

On entry, NG = $\langle value \rangle$.
Constraint: NG ≥ 1 .

IFAIL = 8

On entry, POPT is neither 1 or 2.

IFAIL = 9

On entry, row k of supplied PROB does not sum to 1: $k = \langle value \rangle$.

IFAIL = 10

On entry, LDPROB = $\langle value \rangle$ and N = $\langle value \rangle$.
Constraint: LDPROB $\geq N$.

IFAIL = 11

On entry, NITER = $\langle value \rangle$.
Constraint: NITER ≥ 1 .

IFAIL = 16

On entry, SOPT < 1 or SOPT > 5.

IFAIL = 18

On entry, LDS = $\langle value \rangle$ was invalid.

IFAIL = 19

On entry, SDS = $\langle value \rangle$ was invalid.

IFAIL = 44

A covariance matrix is not positive definite, try a different initial allocation.

IFAIL = 45

An iteration cannot continue due to an empty group, try a different initial allocation.

IFAIL = -999

Dynamic memory allocation failed.

7 Accuracy

Not applicable.

8 Further Comments

None.

9 Example

This example fits a Gaussian mixture model with pooled covariance structure to New Haven schools test data, see Table 5.1 (p. 118) in Hartigan (1975).

9.1 Program Text

```

Program g03gafe

!     Mark 24 Release. NAG Copyright 2012.

!     .. Use Statements ..
Use nag_library, Only: g03gaf, nag_wp, x04caf
!     .. Implicit None Statement ..
Implicit None
!     .. Parameters ..
Integer, Parameter :: nin = 5, nout = 6
!     .. Local Scalars ..
Real (Kind=nag_wp) :: loglik, tol
Integer :: i, ifail, lds, ldx, m, n, &
           ng, niter, nvar, popt, riter, sds, &
           sopt
!     .. Local Arrays ..
Real (Kind=nag_wp), Allocatable :: f(:,:,:), g(:,:,:),
                                   prob(:,:,:), s(:,:,:,:),
                                   w(:), x(:,:)
Integer, Allocatable :: isx(:)
!     .. Executable Statements ..
Write (nout,*) 'G03GAF Example Program Results'
Write (nout,*)

!     Skip heading in data file
Read (nin,*)

!     Problem size
Read (nin,*) n, m, nvar

!     Number of groups
Read (nin,*) ng

!     Scaling option
Read (nin,*) sopt

!     Initial probabilities option
Read (nin,*) popt

!     Maximum number of iterations
Read (nin,*) niter

!     Leading dimensions
ldx = n
ldprob = n

Select Case (sopt)
Case (1)
    Allocate (s(nvar,nvar,ng))
    lds = nvar
    sds = nvar
Case (2)
    Allocate (s(nvar,nvar,1))
    lds = nvar
    sds = nvar
Case (3)
    Allocate (s(nvar,ng,1))
    lds = nvar
    sds = ng
Case (4)
    Allocate (s(nvar,1,1))
    lds = nvar
    sds = 1
Case Default
    Allocate (s(1,1,1))
    lds = 1
    sds = 1
End Select

Allocate (x(ldx,m),prob(ldprob,ng),g(nvar,ng),w(ng),isx(m),f(n,ng))

```

```

!
! Data matrix X
Read (nin,*)(x(i,1:m),i=1,n)

!
! Included variables
If (nvar/=m) Then
  Read (nin,*) isx(1:m)
End If

!
! Optionally read initial probabilties of group membership
If (popt==2) Then
  Read (nin,*)(prob(i,1:ng),i=1,n)
End If

tol = 0.0E0_nag_wp
riter = 5

ifail = 0
Call g03gaf(n,m,x,ldx,isx,nvar,ng,popt,prob,ldprob,niter,riter,w,g,sopt, &
  s,lds,sds,f,tol,loglik,ifail)

!
! Results
Write (nout,*)
ifail = 0
Call x04caf('g','n',1,ng,w,1,'Mixing proportions',ifail)

Write (nout,*)
ifail = 0
Call x04caf('g','n',nvar,ng,g,nvar,'Group means',ifail)

Write (nout,*)
Select Case (sopt)
Case (1)
  Do i = 1, ng
    ifail = 0
    Call x04caf('g','n',nvar,nvar,s(1,1,i),lds, &
      'Variance-covariance matrix',ifail)
  End Do
Case (2)
  ifail = 0
  Call x04caf('g','n',nvar,nvar,s,lds, &
    'Pooled Variance-covariance matrix',ifail)
Case (3)
  ifail = 0
  Call x04caf('g','n',nvar,ng,s,lds,'Groupwise Variance',ifail)
Case (4)
  ifail = 0
  Call x04caf('g','n',nvar,1,s,lds,'Pooled Variance',ifail)
Case (5)
  ifail = 0
  Call x04caf('g','n',1,1,s,lds,'Overall Variance',ifail)
End Select

Write (nout,*)
ifail = 0
Call x04caf('g','n',n,ng,f,n,'Densities',ifail)

Write (nout,*)
ifail = 0
Call x04caf('g','n',n,ng,prob,n,'Membership probabilities',ifail)

Write (nout,*)
Write (nout,'(1X,A,1X,I16)') 'No. iterations:', niter

Write (nout,'(1X,A,1X,F16.4)') 'Log-likelihood:', loglik

Deallocate (x,prob,g,s,w,isx,f)
End Program g03gafe

```

9.2 Program Data

```
G03GAF Example Program Data
25 4 4      : N M IP
2          : NG
2          : SOPT
2          : POPT
15         : NITER
2.7 3.2 4.5 4.8
3.9 3.8 5.9 6.2
4.8 4.1 6.8 5.5
3.1 3.5 4.3 4.6
3.4 3.7 5.1 5.6
3.1 3.4 4.1 4.7
4.6 4.4 6.6 6.1
3.1 3.3 4.0 4.9
3.8 3.7 4.7 4.9
5.2 4.9 8.2 6.9
3.9 3.8 5.2 5.4
4.1 4.0 5.6 5.6
5.7 5.1 7.0 6.3
3.0 3.2 4.5 5.0
2.9 3.3 4.5 5.1
3.4 3.3 4.4 5.0
4.0 4.2 5.2 5.4
3.0 3.0 4.6 5.0
4.0 4.1 5.9 5.8
3.0 3.2 4.4 5.1
3.6 3.6 5.3 5.4
3.1 3.2 4.6 5.0
3.2 3.3 5.4 5.3
3.0 3.4 4.2 4.7
3.8 4.0 6.9 6.7 : X
1.0 0.0
1.0 0.0
1.0 0.0
1.0 0.0
1.0 0.0
1.0 0.0
1.0 0.0
1.0 0.0
1.0 0.0
1.0 0.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0
0.0 1.0 : P
```

9.3 Program Results

G03GAF Example Program Results

Mixing proportions		
	1	2
1	0.4798	0.5202
Group means	1	2

1	4.0041	3.3350
2	3.9949	3.4434
3	5.5894	4.9870
4	5.4432	5.3602

Pooled Variance-covariance matrix

	1	2	3	4
1	0.4539	0.2891	0.6075	0.3413
2	0.2891	0.2048	0.4101	0.2490
3	0.6075	0.4101	1.0648	0.6011
4	0.3413	0.2490	0.6011	0.3759

Densities

	1	2
1	2.5836E-01	1.1853E-02
2	3.7065E-07	1.1241E-01
3	5.3069E-03	1.8080E-06
4	4.2461E-01	2.8584E-05
5	5.0387E-02	1.1544E+00
6	1.1260E+00	7.2224E-02
7	2.0911E+00	2.1224E-02
8	5.7856E-03	1.3227E+00
9	1.1609E+00	2.9411E-02
10	8.9826E-02	2.4260E-05
11	3.0170E-01	1.0106E+00
12	1.2930E+00	3.5422E-01
13	2.8644E-02	6.7851E-07
14	2.0759E-02	3.1690E+00
15	7.6461E-02	1.5231E+00
16	3.0279E-04	8.4017E-01
17	5.6101E-01	4.6699E-05
18	2.6573E-05	6.4442E-01
19	2.1250E+02	5.1006E-02
20	8.6822E-04	2.7626E+00
21	1.9223E-01	2.3971E+00
22	1.2469E-02	2.8179E+00
23	1.8389E-02	5.3572E-01
24	1.2409E+00	9.6489E-03
25	2.1037E-05	4.8674E-02

Membership probabilities

	1	2
1	9.5018E-01	4.9823E-02
2	3.3259E-06	1.0000E+00
3	9.9961E-01	3.8664E-04
4	9.9992E-01	7.9913E-05
5	3.8999E-02	9.6100E-01
6	9.3270E-01	6.7295E-02
7	9.8881E-01	1.1190E-02
8	4.1252E-03	9.9587E-01
9	9.7252E-01	2.7479E-02
10	9.9969E-01	3.0805E-04
11	2.1722E-01	7.8278E-01
12	7.6938E-01	2.3062E-01
13	9.9997E-01	2.6937E-05
14	6.1133E-03	9.9389E-01
15	4.4189E-02	9.5581E-01
16	3.5006E-04	9.9965E-01
17	9.9990E-01	9.7029E-05
18	4.0270E-05	9.9996E-01
19	9.7380E-01	2.6202E-02
20	3.0204E-04	9.9970E-01
21	6.9471E-02	9.3053E-01
22	4.1603E-03	9.9584E-01
23	3.0839E-02	9.6916E-01

24 9.9116E-01 8.8421E-03
25 4.1534E-04 9.9958E-01

No. iterations: 14
Log-likelihood: -29.6831
