

NAG Fortran Library Routine Document

G13FCF

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

G13FCF estimates the parameters of a univariate regression-type II AGARCH(p, q) process.

2 Specification

```

SUBROUTINE G13FCF(DIST, YT, X, LDX, NUM, IP, IQ, NREG, MN, NPAR, THETA,
1 SE, SC, COVAR, LDC, HP, ET, HT, LGF, COPTS, MAXIT,
2 TOL, WORK, LWORK, IFAIL)
INTEGER LDX, NUM, IP, IQ, NREG, MN, NPAR, LDC, MAXIT, LWORK,
1 IFAIL
real YT(NUM), X(LDX,*), THETA(NPAR), SE(NPAR), SC(NPAR),
1 COVAR(LDC,NPAR), HP, ET(NUM), HT(NUM), LGF, TOL,
2 WORK(LWORK)
LOGICAL COPTS(2)
CHARACTER*1 DIST

```

3 Description

A univariate regression-type II AGARCH(p, q) process, with q coefficients α_i , for $i = 1, \dots, q$, p coefficients, β_i , for $i = 1, \dots, p$ and k linear regression coefficients b_i , for $i = 1, \dots, k$, can be represented by:

$$y_t = b_o + x_t^T b + \epsilon_t \quad (1)$$

$$h_t = \alpha_0 + \sum_{i=1}^q \alpha_i (|\epsilon_{t-i}| + \gamma \epsilon_{t-i})^2 + \sum_{i=1}^p \beta_i h_{t-i}, \quad t = 1, \dots, T. \quad (2)$$

where $\epsilon_t | \psi_{t-1} = N(0, h_t)$ or $\epsilon_t | \psi_{t-1} = S_t(df, h_t)$. Here S_t is a standardised Student's t -distribution with df degrees of freedom and variance h_t , T is the number of terms in the sequence, y_t denotes the endogenous variables, x_t the exogenous variables, b_o the regression mean, b the regression coefficients, ϵ_t the residuals, h_t is the conditional variance, and ψ_t the set of all information up to time t .

G13FCF provides an estimate for the parameter vector $\theta = (b_o, b^T, \omega^T)$ where $b^T = (b_1, \dots, b_k)$, $\omega^T = (\alpha_0, \alpha_1, \dots, \alpha_q, \beta_1, \dots, \beta_p, \gamma)$ when DIST = 'N' and $\omega^T = (\alpha_0, \alpha_1, \dots, \alpha_q, \beta_1, \dots, \beta_p, \gamma, df)$ when DIST = 'T'.

MN and NREG can be used to simplify the GARCH(p, q) expression in (1) as follows:

No Regression and No Mean

$$y_t = \epsilon_t,$$

$$\text{MN} = 0,$$

$$\text{NREG} = 0 \text{ and}$$

θ is a $(p + q + 2)$ vector when DIST = 'N' and a $(p + q + 3)$ vector, when DIST = 'T'.

No Regression

$$y_t = b_o + \epsilon_t,$$

$$\text{MN} = 1,$$

$$\text{NREG} = 0 \text{ and}$$

θ is a $(p + q + 3)$ vector when DIST = 'N' and a $(p + q + 4) \times 1$ vector, when DIST = 'T'.

Note: if the $y_t = \mu + \epsilon_t$, where μ is known (not to be estimated by G13FCF) then (1) can be written as $y_t^\mu = \epsilon_t$, where $y_t^\mu = y_t - \mu$. This corresponds to the case **No Regression and No Mean**, with y_t replaced by $y_t - \mu$.

No Mean

$$y_t = x_t^T b + \epsilon_t,$$

$$MN = 0,$$

$$NREG = k \text{ and}$$

θ is a $(p + q + k + 2)$ vector when DIST = 'N' and a $(p + q + k + 3)$ vector, when DIST = 'T'.

4 References

Engle R (1982) Autoregressive conditional heteroskedasticity with estimates of the variance of United Kingdom inflation *Econometrica* **50** 987–1008

Bollerslev T (1986) Generalised autoregressive conditional heteroskedasticity *Journal of Econometrics* **31** 307–327

Engle R and Ng V (1993) Measuring and Testing the Impact of News on Volatility *Journal of Finance* **48** 1749–1777

Hamilton J (1994) *Time Series Analysis* Princeton University Press

5 Parameters

- 1: DIST – CHARACTER*1 *Input*
On entry: the type of distribution to use for e_t .
 If DIST = 'N', a Normal distribution is used.
 If DIST = 'T', a Student's t -distribution is used.
- 2: YT(NUM) – *real* array *Input*
On entry: the sequence of observations, y_t , for $t = 1, \dots, T$.
- 3: X(LDX,*) – *real* array *Input*
Note: the second dimension of the array X must be at least $\max(1, NREG + MN)$.
On entry: row t of X must contain the time dependent exogenous vector x_t , where $x_t^T = (x_t^1, \dots, x_t^k)$, for $t = 1, \dots, T$.
- 4: LDX – INTEGER *Input*
On entry: the first dimension of the array X as declared in the (sub)program from which G13FCF is called.
Constraint: $LDX \geq \text{NUM}$.
- 5: NUM – INTEGER *Input*
On entry: the number of terms in the sequence, T .
Constraint: $\text{NUM} \geq \max(\text{IP}, \text{IQ})$.
- 6: IP – INTEGER *Input*
On entry: the number of coefficients, β_i , for $i = 1, \dots, p$.
Constraint: $\text{IP} \geq 0$ (see also NPAR).

- 7: IQ – INTEGER Input
On entry: the number of coefficients, α_i , for $i = 1, \dots, q$.
Constraint: $IQ \geq 1$ (see also NPAR).
- 8: NREG – INTEGER Input
On entry: the number of regression coefficients, k .
Constraint: $NREG \geq 0$ (see also NPAR).
- 9: MN – INTEGER Input
On entry: if $MN = 1$, the mean term b_0 will be included in the model.
Constraint: $MN = 0$ or 1 .
- 10: NPAR – INTEGER Input
On entry: the number of parameters to be included in the model.
 $NPAR = 2 + IQ + IP + MN + NREG$ when $DIST = 'N'$, and
 $NPAR = 3 + IQ + IP + MN + NREG$ when $DIST = 'T'$.
Constraint: $NPAR < 20$.
- 11: THETA(NPAR) – *real* array Input/Output
On entry: the initial parameter estimates for the vector θ . The first element must contain the coefficient α_o and the next IQ elements must contain the coefficients α_i , for $i = 1, \dots, q$. The next IP elements must contain the coefficients β_j , for $j = 1, \dots, p$. The next element must contain the asymmetry parameter γ . If $DIST = 'T'$, then the next element must contain df , the number of degrees of freedom of the Student's t -distribution. If $MN = 1$ then the next element contains the mean term b_o . If $COPTS(2) = .FALSE.$, then the remaining NREG elements are taken as initial estimates of the linear regression coefficients b_i , for $i = 1, \dots, k$.
On exit: the estimated values $\hat{\theta}$ for the vector θ . The first element contains the coefficient α_o , the next IQ elements contain the coefficients α_i , for $i = 1, \dots, q$. The next IP elements are the coefficients β_j , for $j = 1, \dots, p$. The next element contains the estimate for the asymmetry parameter γ . If $DIST = 'T'$ then the next element contains an estimate for df , the number of degrees of freedom of the Student's t -distribution. If $MN = 1$ then the next element contains an estimate for the mean term b_o . The final NREG elements are the estimated linear regression coefficients b_i , for $i = 1, \dots, k$.
- 12: SE(NPAR) – *real* array Output
On exit: the standard errors for $\hat{\theta}$. The first element contains the standard error for α_o , the next IQ elements contain the standard errors for α_i , for $i = 1, \dots, q$, the next IP elements are the standard errors for β_j , for $j = 1, \dots, p$. The next element contains the standard error for γ . If $DIST = 'T'$ then the next element contains the standard error for df , the number of degrees of freedom of the Student's t -distribution. If $MN = 1$ then the next element contains the standard error for b_o . The final NREG elements are the standard errors for b_j , for $j = 1, \dots, k$.
- 13: SC(NPAR) – *real* array Output
On exit: the scores for $\hat{\theta}$. The first element contains the score for α_o , the next IQ elements contain the score for α_i , for $i = 1, \dots, q$, the next IP elements are the scores for β_j , for $j = 1, \dots, p$. The next element contains the score for γ . If $DIST = 'T'$ then the next element contains the score for df , the number of degrees of freedom of the Student's t -distribution. If $MN = 1$ then the next element contains the score for b_o . The final NREG elements are the scores for b_j , for $j = 1, \dots, k$.

- 14: COVAR(LDC,NPAR) – *real* array *Output*
On exit: the covariance matrix of the parameter estimates $\hat{\theta}$, that is the inverse of the Fisher Information Matrix.
- 15: LDC – INTEGER *Input*
On entry: the first dimension of the array COVAR as declared in the (sub)program from which G13FCF is called.
Constraint: LDC \geq NPAR.
- 16: HP – *real* *Input/Output*
On entry: if COPTS(2) = .FALSE., then HP is the value to be used for the pre-observed conditional variance; otherwise HP is not referenced.
On exit: if COPTS(2) = .TRUE., then HP is the estimated value of the pre-observed of the conditional variance.
- 17: ET(NUM) – *real* array *Output*
On exit: the estimated residuals, ϵ_t , for $t = 1, \dots, T$.
- 18: HT(NUM) – *real* array *Output*
On exit: the estimated conditional variances, h_t , for $t = 1, \dots, T$.
- 19: LGF – *real* *Output*
On exit: the value of the log likelihood function at $\hat{\theta}$.
- 20: COPTS(2) – LOGICAL array *Input*
On entry: the options to be used by G13FCF.
 If COPTS(1) = .TRUE., stationary conditions are enforced, otherwise they are not.
 If COPTS(2) = .TRUE., the routine provides initial parameter estimates of the regression terms, otherwise these are provided by the user.
- 21: MAXIT – INTEGER *Input*
On entry: the maximum number of iterations to be used by the optimization routine when estimating the GARCH(p, q) parameters. If MAXIT is set to 0 then the standard errors, score vector and variance-covariance are calculated for the input value of θ in THETA; however the value of θ is not updated.
Constraint: MAXIT \geq 0.
- 22: TOL – *real* *Input*
On entry: the tolerance to be used by the optimization routine when estimating the GARCH(p, q) parameters.
- 23: WORK(LWORK) – *real* array *Workspace*
 24: LWORK – INTEGER *Input*
On entry: the first dimension of the array WORK as declared in the (sub)program from which G13FCF is called.
Constraint: LWORK \geq (NREG + 3) \times NUM + NPAR + 403.
- 25: IFAIL – INTEGER *Input/Output*
On entry: IFAIL must be set to 0, -1 or 1. Users who are unfamiliar with this parameter should refer to Chapter P01 for details.

On exit: IFAIL = 0 unless the routine detects an error (see Section 6).

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, because for this routine the values of the output parameters may be useful even if IFAIL $\neq 0$ on exit, the recommended value is -1 . **When the value -1 or 1 is used it is essential to test the value of IFAIL on exit.**

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, NREG < 0,
 or MN > 1,
 or MN < 0,
 or IQ < 1,
 or IP < 0,
 or NPAR > 20,
 or LDC < NPAR,
 or LDX < NUM,
 or DIST \neq 'N', and DIST \neq 'T',
 or MAXIT < 0.

IFAIL = 2

On entry, LWORK < (NREG + 3) \times NUM + 3.

IFAIL = 3

The matrix X is not full rank.

IFAIL = 4

The information matrix is not positive definite.

IFAIL = 5

The maximum number of iterations has been reached.

IFAIL = 6

The log-likelihood cannot be optimised any further.

IFAIL = 7

No feasible model parameters could be found.

7 Accuracy

Not applicable.

8 Further Comments

None.

9 Example

This example program uses G05HLF to generate 1500 data points, with known process parameters θ for the following two time-series:

- (i) A GARCH(1,1) sequence with normally distributed residuals.
- (ii) A GARCH(1,1) sequence with Student's t -distributed residuals.

Here G05HLF is initially called, with the output discarded, to eliminate 'start-up effects' in these sequences. The process parameter estimates, $\hat{\theta}$, are then obtained using G13FCF, and compared with their true values, θ . Finally a four step ahead volatility estimate is computed using G13FDF.

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.

```
*      G13FCF Example Program Text
*      Mark 20 Release. NAG Copyright 2001.
*      .. Parameters ..
INTEGER          NOUT
PARAMETER        (NOUT=6)
INTEGER          NPARMX, NUM
real            ZERO
PARAMETER        (NPARMX=10,NUM=1500,ZERO=0.0e0)
INTEGER          NUM1, NREGMX
PARAMETER        (NUM1=3000,NREGMX=10)
*      .. Local Scalars ..
real            DF, FAC1, GAMMA, HP, LGF, MEAN, TOL, XTERM
INTEGER          I, IFLAG, IGEN, IP, IQ, K, LDX, LWK, MAXIT, MN,
+               NPAR, NPAR2, NREG, NT
LOGICAL          FCALL
CHARACTER        DIST
*      .. Local Arrays ..
real            BX(10), COVAR(NPARMX,NPARMX), CVAR(100),
+               ETM(NUM1), HT(NUM1+10), HTM(NUM1), PARAM(NPARMX),
+               RVEC(40), RWSAV(9), SC(NPARMX), SE(NPARMX),
+               THETA(NPARMX), WK(NUM1*3+NPARMX+NREGMX*NUM1+20*
+               20+1), X(NUM1,10), YT(NUM1+10)
INTEGER          ISEED(4)
LOGICAL          COPTS(2)
*      .. External Subroutines ..
EXTERNAL         G05HLF, G05KBF, G13FCF, G13FDF
*      .. Intrinsic Functions ..
INTRINSIC        real, SIN
*      .. Executable Statements ..

WRITE (NOUT,*) 'G13FCF Example Program Results'

ISEED(1) = 111
IGEN = 0

LDX = NUM1
BX(1) = 1.5e0
BX(2) = 2.5e0
BX(3) = 3.0e0
MEAN = 3.0e0

DO 20 I = 1, NUM
  FAC1 = real(I)*0.01e0
  X(I,1) = 0.01e0 + 0.7e0*SIN(FAC1)
  X(I,2) = 0.5e0 + FAC1*0.1e0
  X(I,3) = 1.0e0
20 CONTINUE

MN = 1
NREG = 2
GAMMA = -0.4e0
```

```

IP = 1
IQ = 1
NPAR = IQ + IP + 1
LWK = NREG*NUM + 3*NUM + NPAR + NREG + MN + 404

PARAM(1) = 0.08e0
PARAM(2) = 0.2e0
PARAM(3) = 0.7e0

FCALL = .TRUE.
DIST = 'N'
DF = 4.1e0

*      GOTO 88

ISEED(1) = 111
CALL G05KBF(IGEN, ISEED)
CALL G05HLF(DIST, 300, IP, IQ, PARAM, GAMMA, DF, HT, YT, FCALL, RVEC, IGEN,
+          ISEED, RWSAV, IFLAG)

FCALL = .FALSE.
CALL G05HLF(DIST, NUM, IP, IQ, PARAM, GAMMA, DF, HT, YT, FCALL, RVEC, IGEN,
+          ISEED, RWSAV, IFLAG)

DO 60 I = 1, NUM
  XTERM = ZERO
  DO 40 K = 1, NREG
    XTERM = XTERM + X(I, K)*BX(K)
40  CONTINUE
  IF (MN.EQ.1) THEN
    YT(I) = MEAN + XTERM + YT(I)
  ELSE
    YT(I) = XTERM + YT(I)
  END IF
60  CONTINUE

IFLAG = -1

DO 80 I = 1, NPAR
  THETA(I) = PARAM(I)*0.5e0
80  CONTINUE

THETA(NPAR+1) = GAMMA*0.5e0

IF (MN.EQ.1) THEN
  THETA(NPAR+1+MN) = MEAN*0.5e0
END IF

DO 100 I = 1, NREG
  THETA(NPAR+1+MN+I) = BX(I)*0.5e0
100 CONTINUE

MAXIT = 50
TOL = 1.0e-5

COPTS(1) = .TRUE.
COPTS(2) = .TRUE.
NPAR2 = 2 + IP + IQ + MN + NREG

CALL G13FCF(DIST, YT, X, LDX, NUM, IP, IQ, NREG, MN, NPAR2, THETA, SE, SC,
+          COVAR, NPARMX, HP, ETM, HTM, LGF, COPTS, MAXIT, TOL, WK, LWK,
+          IFLAG)

WRITE (NOUT, *)
WRITE (NOUT, *) 'Normal distribution'
WRITE (NOUT, *)
WRITE (NOUT, *) '          Parameter          Standard          Correct'
WRITE (NOUT, *) '          estimates          errors          values'

DO 120 I = 1, NPAR

```

```

      WRITE (NOUT,99999) THETA(I), SE(I), PARAM(I)
120 CONTINUE

      WRITE (NOUT,99999) THETA(NPAR+1), SE(NPAR+1), GAMMA

      IF (MN.EQ.1) THEN
        WRITE (NOUT,99999) THETA(NPAR+2), SE(NPAR+2), MEAN
      END IF

      DO 140 I = 1, NREG
        WRITE (NOUT,99999) THETA(NPAR+MN+1+I), SE(NPAR+MN+1+I), BX(I)
140 CONTINUE

      NT = 4
      CALL G13FDF(NUM,NT,IP,IQ,THETA,GAMMA,CVAR,HTM,ETM,IFLAG)

      WRITE (NOUT,*)
      WRITE (NOUT,99998) 'Volatility forecast = ', CVAR(NT)
      WRITE (NOUT,*)

      LWK = NUM1*3 + NPARMX + NREGMX*NUM1 + 1
      LDX = NUM1

      BX(1) = 1.5e0
      BX(2) = 2.5e0
      BX(3) = 3.0e0
      MEAN = 3.0e0

      DO 180 I = 1, NUM
        FAC1 = real(I)*0.01e0
        X(I,1) = 0.01e0 + 0.7e0*SIN(FAC1)
        X(I,2) = 0.5e0 + FAC1*0.1e0
        X(I,3) = 1.0e0
180 CONTINUE

      MN = 1
      NREG = 2
      GAMMA = -0.4e0
      IP = 1
      IQ = 1
      NPAR = IQ + IP + 1
      LWK = NREG*NUM + 3*NUM + NPAR + NREG + MN + 405

      PARAM(1) = 0.1e0
      PARAM(2) = 0.1e0
      PARAM(3) = 0.8e0

      FCALL = .TRUE.
      DIST = 'T'

      ISEED(1) = 111
      CALL G05KBF(IGEN,ISEED)

      CALL G05HLF(DIST,300,IP,IQ,PARAM,GAMMA,DF,HT,YT,FCALL,RVEC,IGEN,
+             ISEED,RWSAV,IFLAG)

      FCALL = .FALSE.
      CALL G05HLF(DIST,NUM,IP,IQ,PARAM,GAMMA,DF,HT,YT,FCALL,RVEC,IGEN,
+             ISEED,RWSAV,IFLAG)

      FCALL = .FALSE.
      CALL G05HLF(DIST,NUM,IP,IQ,PARAM,GAMMA,DF,HT,YT,FCALL,RVEC,IGEN,
+             ISEED,RWSAV,IFLAG)

      DO 220 I = 1, NUM
        XTERM = ZERO
        DO 200 K = 1, NREG
          XTERM = XTERM + X(I,K)*BX(K)
200 CONTINUE

```



```

      IF (MN.EQ.1) THEN
        YT(I) = MEAN + XTERM + YT(I)
      ELSE
        YT(I) = XTERM + YT(I)
      END IF
220 CONTINUE

      IFLAG = -1

      DO 240 I = 1, NPAR
        THETA(I) = PARAM(I)*0.5e0
240 CONTINUE

      THETA(NPAR+1) = GAMMA*0.5e0
      THETA(NPAR+2) = DF*0.5e0

      IF (MN.EQ.1) THEN
        THETA(NPAR+2+MN) = MEAN*0.5e0
      END IF
      DO 260 I = 1, NREG
        THETA(NPAR+MN+2+I) = BX(I)*0.5e0
260 CONTINUE

      MAXIT = 100
      TOL = 1.0e-5

      COPTS(1) = .TRUE.
      COPTS(2) = .TRUE.
      NPAR2 = 3 + IP + IQ + MN + NREG

      CALL G13FCF(DIST,YT,X,LDX,NUM,IP,IQ,NREG,MN,NPAR2,THETA,SE,SC,
+              COVAR,NPARAMX,HP,ETM,HTM,LGF,COPTS,MAXIT,TOL,WK,LWK,
+              IFLAG)

      WRITE (NOUT,*)
      WRITE (NOUT,*) 'Student t-distribution'
      WRITE (NOUT,*)
      WRITE (NOUT,*) '          Parameter          Standard          Correct'
      WRITE (NOUT,*) '          estimates          errors          values'

      DO 280 I = 1, NPAR
        WRITE (NOUT,99999) THETA(I), SE(I), PARAM(I)
280 CONTINUE

      WRITE (NOUT,99999) THETA(NPAR+1), SE(NPAR+1), GAMMA

      WRITE (NOUT,99999) THETA(NPAR+2), SE(NPAR+2), DF

      IF (MN.EQ.1) THEN
        WRITE (NOUT,99999) THETA(NPAR+2+MN), SE(NPAR+2+MN), MEAN
      END IF

      DO 300 I = 1, NREG
        WRITE (NOUT,99999) THETA(NPAR+2+MN+I), SE(NPAR+2+MN+I), BX(I)
300 CONTINUE

      NT = 4
      CALL G13FDF(NUM,NT,IP,IQ,THETA,GAMMA,CVAR,HTM,ETM,IFLAG)

      WRITE (NOUT,*)
      WRITE (NOUT,99998) 'Volatility forecast = ', CVAR(NT)
      WRITE (NOUT,*)
      STOP
*
99999 FORMAT (1X,3F16.4)
99998 FORMAT (1X,A,F12.4)
END

```

9.2 Program Data

None.

9.3 Program Results

G13FCF Example Program Results

Normal distribution

Parameter estimates	Standard errors	Correct values
0.0836	0.0154	0.0800
0.2150	0.0312	0.2000
0.6895	0.0324	0.7000
-0.3758	0.0655	-0.4000
3.0457	0.0591	3.0000
1.4562	0.0389	1.5000
2.4570	0.0445	2.5000

Volatility forecast = 3.0370

Student t-distribution

Parameter estimates	Standard errors	Correct values
0.0945	0.0320	0.1000
0.0801	0.0231	0.1000
0.8196	0.0478	0.8000
-0.5139	0.1362	-0.4000
3.7504	0.3690	4.1000
3.0046	0.0647	3.0000
1.5321	0.0379	1.5000
2.4798	0.0479	2.5000

Volatility forecast = 2.3713
