

# NAG Fortran Library Routine Document

## G02FCF

**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

G02FCF calculates the Durbin–Watson statistic, for a set of residuals, and the upper and lower bounds for its significance.

### 2 Specification

```
SUBROUTINE G02FCF(N, IP, RES, D, PDL, PDU, WORK, IFAIL)
INTEGER          N, IP, IFAIL
real           RES(N), D, PDL, PDU, WORK(N)
```

### 3 Description

For the general linear regression model

$$y = X\beta + \epsilon,$$

where  $y$  is a vector of length  $n$  of the dependent variable,

$X$  is a  $n$  by  $p$  matrix of the independent variables,

$\beta$  is a vector of length  $p$  of unknown parameters,

and  $\epsilon$  is a vector of length  $n$  of unknown random errors.

The residuals are given by

$$r = y - \hat{y} = y - X\hat{\beta}$$

and the fitted values,  $\hat{y} = X\hat{\beta}$ , can be written as  $Hy$  for a  $n$  by  $n$  matrix  $H$ . Note that when a mean term is included in the model the sum of the residuals is zero. If the observations have been taken serially, that is  $y_1, y_2, \dots, y_n$  can be considered as a time series, the Durbin–Watson test can be used to test for serial correlation in the  $\epsilon_i$ , see Durbin and Watson (1950), Durbin and Watson (1951) and Durbin and Watson (1971).

The Durbin–Watson statistic is

$$d = \frac{\sum_{i=1}^{n-1} (r_{i+1} - r_i)^2}{\sum_{i=1}^n r_i^2}.$$

Positive serial correlation in the  $\epsilon_i$  will lead to a small value of  $d$  while for independent errors  $d$  will be close to 2. Durbin and Watson show that the exact distribution of  $d$  depends on the eigenvalues of the matrix  $HA$  where the matrix  $A$  is such that  $d$  can be written as

$$d = \frac{r^T Ar}{r^T r}$$

and the eigenvalues of the matrix  $A$  are  $\lambda_j = (1 - \cos(\pi j/n))$ , for  $j = 1, 2, \dots, n - 1$ .

However bounds on the distribution can be obtained, the lower bound being

$$d_1 = \frac{\sum_{i=1}^{n-p} \lambda_i u_i^2}{\sum_{i=1}^{n-p} u_i^2}$$

and the upper bound being

$$d_u = \frac{\sum_{i=1}^{n-p} \lambda_{i-1+p} u_i^2}{\sum_{i=1}^{n-p} u_i^2},$$

where the  $u_i$  are independent standard Normal variables. The lower tail probabilities associated with these bounds,  $p_l$  and  $p_u$ , are computed by G01EPF. The interpretation of the bounds is that, for a test of size (significance)  $\alpha$ , if  $p_l \leq \alpha$  the test is significant, if  $p_u > \alpha$  the test is not significant, while if  $p_l > \alpha$  and  $p_u \leq \alpha$  no conclusion can be reached.

The above probabilities are for the usual test of positive auto-correlation. If the alternative of negative auto-correlation is required, then a call to G01EPF should be made with the parameter D taking the value of  $4 - d$ ; see Newbold (1988).

## 4 References

- Durbin J and Watson G S (1950) Testing for serial correlation in least-squares regression. I *Biometrika* **37** 409–428
- Durbin J and Watson G S (1951) Testing for serial correlation in least-squares regression. II *Biometrika* **38** 159–178
- Durbin J and Watson G S (1971) Testing for serial correlation in least-squares regression. III *Biometrika* **58** 1–19
- Granger C W J and Newbold P (1986) *Forecasting Economic Time Series* (2nd Edition) Academic Press
- Newbold P (1988) *Statistics for Business and Economics* Prentice-Hall

## 5 Parameters

- |    |   |               |
|----|---|---------------|
| 1: | N – INTEGER   | <i>Input</i>  |
|    | <i>On entry:</i> the number of residuals, $n$ .   |               |
|    | <i>Constraint:</i> $N > IP$ .   |               |
| 2: | IP – INTEGER  | <i>Input</i>  |
|    | <i>On entry:</i> the number, $p$ , of independent variables in the regression model, including the mean.            |               |
|    | <i>Constraint:</i> $IP \geq 1$ .  |               |
| 3: | RES(N) – <i>real</i> array  | <i>Input</i>  |
|    | <i>On entry:</i> the residuals, $r_1, r_2, \dots, r_n$ .  |               |
|    | <i>Constraint:</i> the mean of the residuals $\leq \sqrt{\epsilon}$ , where $\epsilon =$ <i>machine precision</i> . |               |
| 4: | D – <i>real</i>   | <i>Output</i> |
|    | <i>On exit:</i> the Durbin–Watson statistic, $d$ .  |               |
| 5: | PDL – <i>real</i>   | <i>Output</i> |
|    | <i>On exit:</i> lower bound for the significance of the Durbin–Watson statistic, $p_l$ .                            |               |
| 6: | PDU – <i>real</i>   | <i>Output</i> |
|    | <i>On exit:</i> upper bound for the significance of the Durbin–Watson statistic, $p_u$ .                            |               |

7: WORK(N) – *real* array *Workspace*

8: 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, for users 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.**

## 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,  $N \leq IP$ ,  
or  $IP < 1$ .

IFAIL = 2

On entry, the mean of the residuals was  $> \sqrt{\epsilon}$ , where  $\epsilon = \textit{machine precision}$ .

IFAIL = 3

On entry, all residuals are identical.

## 7 Accuracy

The probabilities are computed to an accuracy of at least 4 decimal places.

## 8 Further Comments

If the exact probabilities are required, then the first  $n - p$  eigenvalues of  $HA$  can be computed and G01JDF used to compute the required probabilities with the parameter C set to 0.0 and the parameter D set to the Durbin–Watson statistic  $d$ .

## 9 Example

A set of 10 residuals are read in and the Durbin–Watson statistic along with the probability bounds are computed and printed.

### 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.

```
*      G02FCF Example Program Text.
*      Mark 15 Release. NAG Copyright 1991.
*      .. Parameters ..
      INTEGER          NIN, NOUT
      PARAMETER       (NIN=5, NOUT=6)
      INTEGER          N
      PARAMETER       (N=10)
*      .. Local Scalars ..
      real            D, PDL, PDU
```

```

      INTEGER          I, IFAIL, IP
*    .. Local Arrays ..
      real            RES(N), WORK(N)
*    .. External Subroutines ..
      EXTERNAL        G02FCF
*    .. Executable Statements ..
      WRITE (NOUT,*) 'G02FCF Example Program Results'
*    Skip heading in data file
      READ (NIN,*)
      READ (NIN,*) IP
      READ (NIN,*) (RES(I),I=1,N)
*
      IFAIL = 0
*
      CALL G02FCF(N,IP,RES,D,PDL,PDU,WORK,IFAIL)
*
      WRITE (NOUT,*)
      WRITE (NOUT,99999) ' Durbin-Watson statistic ', D
      WRITE (NOUT,*)
      WRITE (NOUT,99998) ' Lower and upper bound ', PDL, PDU
      STOP
*
99999 FORMAT (1X,A,F10.4)
99998 FORMAT (1X,A,2F10.4)
      END

```

## 9.2 Program Data

G02FCF Example Program Data

```

2
3.735719 0.912755 0.683626 0.416693 1.9902
-0.444816 -1.283088 -3.666035 -0.426357 -1.918697

```

## 9.3 Program Results

G02FCF Example Program Results

```

Durbin-Watson statistic      0.9238
Lower and upper bound       0.0610   0.0060

```

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