## THE DISTRIBUTION OF THE RATIOS OF CERTAIN QUADRATIC FORMS IN TIME SERIES<sup>1</sup>

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- 1. Introduction. In testing the hypothesis that successive members of a series of observations are serially correlated a number of statistics have been proposed. Durbin and Watson [4] gave the exact distribution of several of these statistics when they are slightly modified. We shall extend the work of Durbin and Watson for a non-null case of two of their modified statistics and also find a simple expression for the moments of another of their statistics.
- 2. The Double root result. Assume that  $X' = (x_1, x_2, \dots, x_n)$  has probability density

(2.1) 
$$f(X) = |\Lambda|^{1/2} (2\pi)^{-n/2} \exp\left[-X'\Lambda X/2\right],$$

where  $\Lambda$  is a positive definite matrix and n = 2m. Let

(2.2) 
$$A = \begin{pmatrix} A_1 & 0 \\ 0 & A_1 \end{pmatrix}; \quad B = \begin{pmatrix} B_1 & 0 \\ 0 & B_1 \end{pmatrix}; \quad \Lambda = \begin{pmatrix} \Lambda_1 & 0 \\ 0 & \Lambda_1 \end{pmatrix};$$

where  $B_1$  is positive definite or positive semi-definite and of rank m-q which is  $\geq$  the rank of  $A_1$ , a real symmetric matrix. Further assume that A, B, and  $\Lambda$  commute pairwise, and that the characteristic roots  $a_j$  of A and the characteristic roots  $b_j$  of B are so numbered that if  $a_j \geq 0$ ,  $b_j > 0$  and  $a_j/b_j \geq a_{j+1}/b_{j+1}$  for all  $a_j$  and  $a_{j+1}$  which are  $\neq 0$ .

Now

(2.3) 
$$G(z) = P\left[\frac{X'AX}{X'BX} \le z\right] = P[X'(A - zB)X \le 0],$$

where X is  $N(0, \Lambda^{-1})$ . Making an orthogonal transformation X = PY where  $P'AP = D_a$ ,  $P'BP = D_b$ ,  $P'\Lambda P = D_{\lambda}$  are diagonal matrices with elements  $a_j = a_{m+j}$ ,  $b_j = b_{m+j}$  and  $\lambda_j = \lambda_{m+j}$ , we get

(2.4) 
$$G(z) = P[Y'(D_a - zD_b)Y \le 0],$$

where Y is  $N(0, D_{\lambda}^{-1})$ . Now let  $Y = D_{\lambda}^{-1/2}W$  so that

(2.5) 
$$G(z) = P[W'(D_a - zD_b)D_{\lambda}^{-1}W \leq 0],$$

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where W is N(0, I). Hence by the duplication of roots  $a_i$ ,  $b_i$  and  $\lambda_i$  we get

(2.6) 
$$G(z) = P\left[\sum_{j=1}^{m-q} \lambda_j^{-1} (a_j - zb_j) w_j^2 \le 0\right]$$

where  $w_i^2$  are independent and each is a  $\chi^2$  variable with 2 degrees of freedom.

Using a result by R. L. Anderson [1] which in general terms states that if the  $c_i$  are all different then

(2.7) 
$$P\left[\sum_{j=1}^{p} c_{j} w_{j}^{2} \leq 0\right] = 1 - \sum_{j \in s} c_{j}^{p-1} \prod_{\substack{j \geq k \\ j \geq 1}}^{p} (c_{k} - c_{j})^{-1}$$

where  $S = \{j/c_j > 0\}$ , we find that

(2.8) 
$$G(z) = 1 - \prod_{j=1}^{m-q} \lambda_j \sum_{k=1}^{L} (a_k - b_k z)^{m-q-1} \lambda_k^{-1} \cdot \prod_{\substack{j=1 \ j \neq k}}^{m-q} [\lambda_j (a_k - b_k z) - \lambda_k (a_j - b_j z)]^{-1},$$

where

$$\frac{a_{L+1}}{b_{L+1}} \le z \le \frac{a_L}{b_L}$$

for  $L=1, \dots, m-q$ . This result could also have been gotten by contour integration through the results of Gurland [6] or by Madow's generalization of Anderson's result.

3. The distribution of the Durbin and Watson Statistic in the non-null case. Using the result (2.8) and letting

(3.1) 
$$R = \sum_{i=1}^{2m-1} x_{i+1} x_i / \sum_{i=1}^{2m} x_i^2,$$

where

(3.2) 
$$\Lambda_{1} = \begin{pmatrix} 1 + \rho^{2} & -\rho & \\ -\rho & \cdot & & \\ & \cdot & & \\ & & -\rho & \\ & & -\rho & 1 + \rho^{2} \end{pmatrix},$$

we may find the distribution of R. For  $a_j = \cos(j\pi)/(m+1)$ ,  $\lambda_j = 1 + \rho^2 - 2\rho a_j$ ,  $b_j = 1$ . Now  $\prod_{j=1}^m \lambda_j = (1 - \rho^{2m+2})(1 - \rho^2)^{-1}$  and  $\lambda_k - \lambda_j = (a_k - a_j) \cdot (1 + \rho^2 - 2\rho z)$  and by Geisser [5]

(3.3) 
$$\prod_{\substack{j=1\\j\neq k}}^{m} (a_k - a_j) = (m+1)(-1)^{k+1}2^{-m}\csc^2\frac{k\pi}{m+1}.$$

Therefore

(3.4) 
$$G(z;\rho) = 1 - (1 - \rho^{2m+2})2^{m}(m+1)^{-1}(1 - \rho^{2})^{-1}(1 + \rho^{2} - 2\rho z)^{1-m} \\ \cdot \sum_{k=1}^{L} (-1)^{k+1} \left(\cos\frac{k\pi}{m+1} - z\right)^{m-1} \sin^{2}\frac{k\pi}{m+1} \left(1 + \rho^{2} - \frac{2}{\rho}\cos\frac{k\pi}{m+1}\right)^{-1}$$

and

(3.5) 
$$G'(z; \rho) = g(R; \rho) = (1 - \rho^{2m+2}) 2^m (m-1) (m+1)^{-1} (1 - \rho^2)^{-1} (1 + \rho^2 - 2\rho R)^{-m} \cdot \sum_{k=1}^{L} (-1)^{k+1} \left(\cos \frac{k\pi}{m+1} - R\right)^{m-2} \sin^2 \frac{k\pi}{m+1},$$

where

$$\cos\frac{(L+1)\pi}{m+1} \le R \le \cos\frac{L\pi}{m+1}.$$

For  $\rho = 0$  it is clear that

(3.6) 
$$g(R) = 2^{m}(m-1)(m+1)^{-1} \sum_{k=1}^{L} (-1)^{k+1} \cdot \left(\cos\frac{k\pi}{m+1} - R\right)^{m-2} \sin^{2}\frac{k\pi}{m+1}$$

and hence

(3.7) 
$$g(R;\rho) = (1 - \rho^{2m+2})(1 - \rho^2)^{-1}(1 + \rho^2 - 2\rho R)^{-m}g(R).$$

4. Approximations. In a paper by T. W. Anderson and R. L. Anderson [3] in which the circular serial correlation coefficient is discussed for fitted trigonometric series for the mean, they have fitted the trigonometric series for semi-annual data to correct for variation of period two and get a quadratic form

$$q = X'CX / X'BX$$

for n = 2m.

They reduce q to the form

(4.2) 
$$\sum_{j=1}^{2m-2} c_j y_j^2 / \sum_{j=1}^{2m-2} y_j^2,$$

where the  $c_i$  are identical with the  $a_i$  of the previous section.

Therefore the distribution in this particular case for 2m observations is exactly the same as that for the non-circular case of Durbin and Watson for 2m-2 observations when  $\rho=0$ . They also give the approximate distribution of their circular statistic as a beta distribution, and if we put 2m-2 in place of 2m we get the approximate distribution density of R for 2m-2 observations

when  $\rho = 0$  to be

(4.3) 
$$g(R) \sim K(1 - R^2)^p$$
, where  $p = (m^2 + m)(m - 1)^{-1} - 3/2$ .

**5. Moments of a ratio.** The previous work was based on the assumption that  $\mathcal{E}x_i = \mu$  was known. However, if  $\mu$  is unknown, one of the statistics used is the ratio of the mean square successive difference to the variance;

$$\eta = \frac{\delta^2}{s^2}$$

The distribution of  $\eta$  is for the present too difficult for explicit evaluation. The moments of  $\eta$  have been found by Williams [10] and much light shed on the distribution by Von Neumann [8]. However, the expression for the rth moment given by Williams is in terms of the rth derivative of a function.

Durbin and Watson [4] suggested a modified statistic in this case for n = 2m. Let

(5.2) 
$$R = X'AX / X'BX$$
 or  $R = 4(m-1)\delta_0^2/[(2m-1)s^2],$ 

where

with latent roots

$$a_j = 4 \sin^2 \frac{(m-j)\pi}{2m} = 4 \cos^2 \frac{j\pi}{2m},$$

and

$$B = \frac{1}{2m} \begin{bmatrix} 2m - 1 & -1 & \cdots & -1 \\ -1 & & & \ddots \\ & \ddots & & & \ddots \\ & \ddots & & & -1 \\ & -1 & & \ddots & 2m - 1 \end{bmatrix}.$$

The distribution of R is given to be

(5.3) 
$$P(R > R') = \sum_{k=1}^{L} (a_k - R')^{m-3/2} a_k^{-\frac{1}{2}} \prod_{\substack{j=1\\j > k}}^{m-1} (a_k - a_j)^{-1}$$

for  $a_{k+1} \leq R \leq a_k$ . This result is based on a result of Anderson's where in addition to double roots there is a single root which is less than all of the double roots. By simplification it can be reduced to

$$(5.4) P(R > R') = 4m^{-1} \sum_{k=1}^{L} (-1)^{k+1} (a_k - R')^{m-3/2} \sin^2 \frac{k\pi}{2m} \cos \frac{k\pi}{2m}$$

for  $a_{k+1} \leq R \leq a_k$ . The moments of this ratio can be easily found for r < 3m - 1, since we already have evaluated the moments of the numerator in a previous paper (Geisser [5]) and the moments of the denominator are well known. When the successive observations are independent, the moments of this ratio are the ratio of the moments [9]. Therefore

(5.5) 
$$8R^r = \mu_r = \frac{2^{r+1}(2m+2r-2)!(2m^2-m-r)}{[(2m+r)!(2m-1)(2m+1)\cdots(2m+2r-3)]}.$$

6. The distribution of the modified von Neumann ratio. If we consider the ratio

(6.1) 
$$\eta_0 = \frac{2\delta_0^2}{s_0^2} = \frac{\sum_{\substack{i=1\\i \ge m}}^{2m-1} (x_{i+1} - x_i)^2}{\sum_{\substack{1\\i}}^{m} (x_i - \bar{x}_1)^2 + \sum_{\substack{m=1\\m+1}}^{2m} (x_i - \bar{x}_2)^2},$$

i.e., twice the ratio of the modified mean square successive difference to the pooled variance, we are able to use the Double Root Result and find the distribution of  $\eta_0$  for the non-null alternative given by T. W. Anderson [2] if we further consider the model to be made up of two independent sets and

(6.2) 
$$\Lambda = \begin{pmatrix} \Lambda_{1} & 0 \\ 0 & \Lambda_{1} \end{pmatrix},$$

$$(6.3) \quad \Lambda_{1} = \sigma^{-2} \begin{bmatrix} 1 + \rho^{2} - \rho & -\rho & & & \\ -\rho & 1 + \rho^{2} & & & \\ & & \ddots & & \\ & & & 1 + \rho^{2} & -\rho & \\ & & & -\rho & 1 + \rho^{2} - \rho \end{bmatrix}.$$

As was shown previously [2],  $\eta_0$  provides a uniformly most powerful one-sided test. By the double root result we get, letting  $\lambda'_i = \sigma^2 \lambda_i$ ,

$$G(\eta_0; \rho) = 1 - 2m^{-1} [1 + \rho^2 - 2\rho + \rho \eta_0]^{2-m} \prod_{j=1}^{m-1} \lambda_j'$$

$$\cdot \sum_{k=1}^{L} (-1)^{k+1} (a_k - \eta_0)^{m-2} \lambda_k^{-1} \sin^2 \frac{k\pi}{m}$$

and

$$G'(\eta_0; \rho) = g(\eta_0; \rho) = 2m^{-1}(m-2)[1+\rho^2-2\rho+\rho\eta_0]^{2-m} \prod_{j=1}^{m-1} \lambda_j'$$

$$(6.5) \qquad \qquad \cdot \sum_{k=1}^{L} (-1)^{k+1} (a_k-\eta_0)^{m-3} \sin^2 \frac{k\pi_0}{m^3}$$

where  $a_{L+1} \leq \eta_0 \leq a_L$ . For  $\rho = 0$ ,

(6.6) 
$$G(\eta_0) = 2(m-2)m^{-1} \sum_{k=1}^{L} (a_k - \eta_0)^{m-2} (-1)^{k+1} \sin^2 \frac{k\pi}{m}$$

and

(6.7) 
$$G'(\eta_0) = g(\eta_0) = 2(m-2)m^{-1} \sum_{k=1}^{L} (-1)^{k+1} (a_k - \eta_0)^{m-3} \sin^2 \frac{k\pi}{m}$$

for  $a_{L+1} \leq \eta_0 \leq a_L$ . Hence

(6.8) 
$$G'(\eta_0; \rho) = g(\eta_0; \rho) = (1 + \rho^2 - 2\rho + \rho\eta_0)^{1-m} \left(\prod_{j=1}^{m-1} \lambda_j'\right) g(\eta_0);$$

and since

$$\prod_{j=1}^{m-1} \lambda_j' = \frac{(1-\rho)(1-\rho^{2m})}{(1+\rho)(1-\rho)^2} = \frac{1-\rho^{2m}}{1-\rho^2},$$
(6.9) 
$$G'(\eta_0; \rho) = (1+\rho^2 - 2\rho + \rho\eta_0)^{1-m}(1-\rho^{2m})(1-\rho^2)^{-1}g(\eta_0).$$

It is also quite easy to find the moments  $\xi \eta_o^r$  when  $\rho=0$  since we have already given the moments of the numerator for r<3m-1 and the moments of the denominator are well known. Hence for r<3m-1

(6.10) 
$$\epsilon \eta_0^r = \frac{2^{1-r}(2m^2 - m - r)(2m + 2r - 2)!(m - 2)!}{(m + r - 2)!(2m + r)!}.$$

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