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**BOUNDS ON THE EFFECT OF HETEROSCEDASTICITY ON  
THE CHOW TEST FOR STRUCTURAL CHANGE**

**David Giles and Offer Lieberman**

***Discussion Paper***

**No. 9101**

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**January 1991**

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BOUNDS ON THE EFFECT OF  
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and

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January 1991

Abstract

This paper considers the effect of heteroscedastic regression errors on the size of the Chow test for structural stability. We show that bounds can be placed on the true size of this test in the light of such misspecification, and on the true critical value needed to achieve any desired significance level when using the test under various degrees of heteroscedasticity. These bounds are data-independent, and some cases are tabulated. An example is given to illustrate the practical application of the critical value bounds.

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## I. Introduction

The Chow (1960) test for the constancy of the regression coefficient vector over the sample is one of the most widely used diagnostic tests in applied econometrics. In its various forms, this test amounts to one of the validity of particular exact restrictions on the regression coefficients (e.g., Fisher (1970)).

It is well known that the statistics associated with the various forms of the Chow test are F-distributed under the null hypothesis of parameter stability (and non-central F under the alternative hypothesis), provided that certain conditions are satisfied. The usual assumption of normal errors can be relaxed to one of spherical symmetry (provided that the errors are homoscedastic) without affecting the null distributions of these statistics, but their distributions under the alternative (and hence their power) are sensitive to this relaxation (e.g., Ullah and Phillips (1986), Giles (1991)). The assumption of homoscedastic disturbances over the full sample cannot be relaxed without distorting both the null and alternative distributions of the test statistics even with normal errors. In the face of heteroscedasticity, we have a form of the Behrens-Fisher problem.

Several studies have considered the effect of this misspecification of the model on the Chow test. For example, Toyoda (1974) approximates the distribution of the test statistic in this case, and Schmidt and Sickles (1977) provide exact evidence. Other authors have proposed alternative tests which might be robust to heteroscedasticity or which allow for its presence in some way. For example Jayatissa (1977) suggests a finite-sample test which has been criticised by Honda (1982) and others. Watt (1979) proposes an asymptotic Wald test whose exact distribution is discussed by Ohtani and Toyoda (1985), and finite-sample bounds for which are described by Ohtani and Kobayashi (1986). A further test is suggested by Weerahandi (1987).



MacKinnon (1989) derives heteroscedasticity-robust variants of the Chow test which have asymptotic validity, but whose finite-sample properties are rather mixed.

Given its ease of construction, the Chow test continues to be used widely in favour of the proposed alternatives, even in situations where the homoscedasticity assumption is unreasonable. Following Schmidt and Sickles (1977) it is quite straightforward to determine the true (as opposed to nominal) size of the Chow test for any specific data matrix and known actual level of heteroscedasticity, by using the techniques of Imhof (1961) or Davies (1980). This is somewhat analogous to computing an exact Durbin-Watson test rather than using the tabulated bounds on the critical values, and can be undertaken with the SHAZAM package (White et al. (1990)). However, the size distortion is data-specific, and most applied researchers (who may not have easy access to software for computing the distribution of ratios of quadratic forms in normal random vectors) are unlikely to proceed in this way, even given an estimate of the degree of heteroscedasticity.

Instead, it is common for the Chow test to be applied without allowance for possible heteroscedasticity despite its well known inadequacy in this case. Accordingly, for different degrees of heteroscedasticity, it would be helpful to have bounds on the true critical values for the test (or, equivalently, bounds on its true size) which are independent of the data values in the sample. In this paper we use the results of Kiviet (1980) to construct such bounds. The problem and notation are formalised in the next section. Section III details the construction of the bounds, and Section IV reports our results. Some concluding comments appear in Section V.

## II. Model and Notation

Consider a sample of  $T = T_1 + T_2$  observations and the model

$$y_i = X_i \beta_i + u_i \quad ; \quad i = 1, 2 \quad (1)$$

where  $y_i$  is  $(T_i \times 1)$ ;  $X_i$  is  $(T_i \times K)$ , non-stochastic and of rank  $K$  ( $< T_i$ ); and  $\beta_i$  is  $(K \times 1)$ ;  $i = 1, 2$ . The same variables enter the model in each sub-sample but (typically) with different values.

The most common form of the "Chow test" for parameter stability considers  $H_0: \beta_1 = \beta_2$  vs.  $H_A: \beta_1 \neq \beta_2$ . This may be expressed as a standard test of linear restrictions by writing (1) as

$$y \equiv \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} X_1 & 0 \\ 0 & X_2 \end{pmatrix} \begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix} + \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} \\ \equiv X\beta + u,$$

and the null hypothesis as  $H_0: R\beta = r$ , where  $R = (I_K, -I_K)$  and  $r = 0$ . The Chow test statistic is

$$f = \left( \frac{T-2K}{K} \right) (R\hat{\beta} - r)' [R(X'X)^{-1}R']^{-1} (R\hat{\beta} - r) / e'e,$$

where  $\hat{\beta} = (X'X)^{-1}X'y$  and  $e = y - X\hat{\beta}$ .

If  $u \sim N(0, \sigma^2 I_T)$  then  $f$  is F-distributed with  $K$  and  $(T-2K)$  degrees of freedom if  $H_0$  is true, and it is non-central F with these degrees of freedom and non-centrality parameter  $\phi = (R\beta - r)' [R(X'X)^{-1}R']^{-1} (R\beta - r) / 2\sigma^2$  under  $H_A$ . It is readily verified that under  $H_0$ ,

$$f = \left( \frac{T-2K}{K} \right) u' Au / u' Mu, \quad (2)$$

where

$$M = I_T - X(X'X)^{-1}X'$$

$$A = X(X'X)^{-1}R' \left[ R(X'X)^{-1}R' \right]^{-1} R(X'X)^{-1}X'$$



and both  $M$  and  $A$  are idempotent. If  $u \sim N(0, \sigma^2 \Omega)$ , for arbitrary positive definite symmetric  $\Omega \neq I_T$ , then the above distributional results no longer hold, though (2) is still valid.

As the Chow test is a special case of the usual test of  $h$  linear restrictions on  $\beta$ , the results of Kiviet (1980) can be used to derive bounds on  $f$  when  $u \sim N(0, \sigma^2 \Omega)$ . The bounds on  $f$  can be used to construct bounds on its critical value for any chosen significance level, or on the true significance level of the test constructed by rejecting  $H_0$  if  $f > F^c(\alpha)$ , where  $F^c(\alpha)$  is the  $100\alpha\%$  critical value based on the (wrongly) assumed  $F_{K, T-2K}$  null distribution. All of these bounds depend on  $T$ ,  $K$  and  $\Omega$ , but they are independent of  $X$ .

As noted in the Introduction, the assumption that  $u \sim N(0, \sigma^2 I_T)$  is often unreasonable when applying the Chow test. A more realistic assumption is that  $u_i \sim N(0, \sigma_i^2 I_{T_i})$ , or  $u \sim N(0, \sigma^2 \Omega)$  where  $\Omega = \text{diag.}(\omega_i)$  and

$$\omega_i = \begin{cases} 1 & ; i = 1, \dots, T_1 \\ \theta = \sigma_2^2 / \sigma_1^2 & ; i = T_1 + 1, \dots, T \end{cases}$$

As  $\Omega$  depends on  $T_1$  and  $T_2$  here, these separate values partially determine the various bounds.

### III. Calculation of the Bounds

Recalling the form of  $f$  in (2) and applying the principal theorem of Kiviet (1980, p.354), it follows that under the null hypothesis  $f_L \leq f \leq f_u$ , where

$$f_L = \left( \frac{T-2K}{K} \right) \left( \frac{\sum_{i=1}^K \lambda_i x_i^2}{\sum_{i=2K+1}^T \lambda_i x_i^2} \right) \quad (3)$$

$$f_u = \left( \frac{T-2K}{K} \right) \left( \frac{\sum_{i=1}^K \lambda_{T-K+i} x_i^2}{\sum_{i=2K+1}^T \lambda_{i-2K} x_i^2} \right) \quad (4)$$

The  $\chi_i^2$  are independent central Chi square variates, each with one degree of freedom and  $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_T$  are the eigenvalues of  $\Omega$ . As  $\Omega$  is diagonal here, the  $\lambda_i$ 's are its (appropriately ordered) diagonal elements.

Because (3) and (4) hold only under  $H_0$ , it follows that the true 100 $\alpha$ % upper critical value,  $C(\alpha)$ , of  $f$  satisfies  $C_L(\alpha) \leq C(\alpha) \leq C_U(\alpha)$ , where

$$\Pr. \left( f_L \geq C_L(\alpha) \right) = \alpha \quad (5)$$

$$\Pr. \left( f_U \geq C_U(\alpha) \right) = \alpha \quad (6)$$

Using (3) and (4),  $C_L(\alpha)$  and  $C_U(\alpha)$  may be computed by noting they are values satisfying

$$\Pr. \left( \sum_{i=1}^T w_i \chi_i^2 \leq 0 \right) = 1 - \alpha \quad (7)$$

and

$$\Pr. \left( \sum_{i=1}^T z_i \chi_i^2 \leq 0 \right) = 1 - \alpha, \quad (8)$$

where

$$w_i = \begin{cases} \lambda_1 (T-2K)/K & i=1, \dots, K \\ -\lambda_1 C_L(\alpha) & i=2K+1, \dots, T \\ 0 & \text{otherwise} \end{cases}$$

$$z_i = \begin{cases} \lambda_{T-K+1} (T-2K)/K & i=1, \dots, K \\ -\lambda_{1-2K} C_U(\alpha) & i=2K+1, \dots, T \\ 0 & \text{otherwise} \end{cases}$$

The calculation of (7) and (8) is a standard problem. We use a FORTRAN version of Davies' (1980) algorithm to calculate these probabilities, searching over  $C_L(\alpha)$  and  $C_U(\alpha)$  to satisfy (7) and (8).

Similarly, the true size ( $\alpha_0$ ) of the test which is constructed by rejecting  $H_0$  if  $f > F^C(\alpha)$ , satisfies  $\alpha_L \leq \alpha_0 \leq \alpha_U$ , where

$$\alpha_L = \Pr. \left\{ f_L \geq F^C(\alpha) \right\} \quad (9)$$

$$\alpha_U = \Pr. \left\{ f_U \geq F^C(\alpha) \right\}. \quad (10)$$

Using (3) and (4),

$$\alpha_L = 1 - \Pr. \left( \sum_{i=1}^T w_i^* \chi_i^2 \leq 0 \right) \quad (11)$$

$$\alpha_U = 1 - \Pr. \left( \sum_{i=1}^T z_i^* \chi_i^2 \leq 0 \right) \quad (12)$$

where the  $w_i^*$ 's and  $z_i^*$ 's equal the  $w_i$ 's and  $z_i$ 's respectively, but with  $F^C(\alpha)$  replacing both  $C_L(\alpha)$  and  $C_U(\alpha)$ . Davies' routine facilitates the direct calculation of  $\alpha_L$  and  $\alpha_U$ .

The usefulness of the above bounds is that they are independent of the data - they depend only<sup>1</sup> on  $\alpha$ ,  $K$ ,  $T_1$ ,  $T_2$  and  $\theta$ . Finally, if the errors are homoscedastic,  $\lambda_i = 1$  for all  $i$ . Then recalling the additivity properties of independent Chi square variates, it follows that  $f_L = f = f_U$ ,  $C_L(\alpha) = F^C(\alpha) = C_U(\alpha)$ , and  $\alpha_L = \alpha = \alpha_U$ .

#### IV. RESULTS

Given the form of  $\Omega$ , the bounds will obviously exhibit certain symmetries. For example, if  $T_1 = T_2$ , the results for  $\theta = 0.1$  are identical to those for  $\theta = 10$ , etc. Similarly, those for  $T_1 = 10$  and  $T_2 = 20$  when  $\theta = 0.1$  are identical to those for  $T_2 = 10$  and  $T_1 = 20$  when  $\theta = 10$ , etc.

Table 1 illustrates the bounds on  $C(\alpha)$  and  $\alpha_0$  for  $\alpha = 0.05$  and the values of  $T_1$ ,  $T_2$ ,  $K$  and  $\theta$  considered in Table I of Schmidt and Sickles (1977, p. 1295). Their results provide exact values of  $\alpha_0$  for certain specific  $X$  matrices. It can be verified that  $\alpha_L$  and  $\alpha_U$  in Table 1 bound all of their reported  $\alpha_0$ 's (and the nominal  $\alpha = 0.05$ ). Also, for fixed  $T_1$ ,  $T_2$  and  $K$ , as  $\theta$  departs from unity (and  $H_0$  becomes increasingly false), the values of  $(C_U(\alpha) -$

TABLE 1.-BOUNDS ON TRUE CRITICAL VALUES AND SIZES OF CHOW TEST WHEN  
THE NOMINAL SIGNIFICANCE LEVEL IS 5%

$T_1$	$T_2$	$k$	$\theta$	$C_L$	$C_u$	$\alpha_L$	$\alpha_u$
10	10	2	0.01	0.065	13.134	0.000	0.318
			0.1	0.595	9.884	0.000	0.260
			1	3.634	3.634	0.050	0.050
25	25	2	0.01	0.062	7.476	0.000	0.250
			0.1	0.564	6.583	0.000	0.214
			1	3.200	3.200	0.050	0.050
20	30	2	0.01	0.079	10.171	0.000	0.345
			0.1	0.687	8.297	0.000	0.287
			1	3.200	3.200	0.050	0.050
			10	0.478	5.444	0.000	0.160
			100	0.050	5.904	0.000	0.181
40	10	2	0.01	0.037	4.151	0.000	0.095
			0.1	0.365	4.034	0.000	0.089
			1	3.200	3.200	0.050	0.050
			10	1.191	16.422	0.001	0.513
			100	0.177	33.667	0.000	0.658

$C_L(\alpha)$  and  $(\alpha_u - \alpha_L)$  increase, as expected. This is consistent with the results in Tables 4-6 of Kiviet (1980, pp.356-7) for the general F-test in the case of AR(1) or MA(1) errors<sup>2</sup>.

The effects on the bounds of varying  $T_1$ ,  $T_2$  and  $K$  are best seen by considering Table 1 in conjunction with the Appendix tables. The latter provide bounds on  $C(0.05)$  for a range of situations likely to be encountered in the applied work. In practice, having computed the Chow test statistic, we require a critical value in order to implement the test. In the face of possible heteroscedasticity, the Appendix provides bounds on such critical values.<sup>3</sup>

For  $\theta < 1$ , increasing  $T_1$  (ceteris paribus) leads to decreases in  $C_L(0.05)$ ,  $C_u(0.05)$  and their difference. The converse result emerges if  $\theta > 1$ . For increasing  $T_2$ , these results are reversed in general. Exceptions<sup>4</sup> can be seen for  $C_u(0.05)$  for  $\theta = 10$  or  $100$  in Table A4. Changing the value of  $K$ , ceteris paribus, produces less clear patterns in the results. This is to be expected as  $K$  determines both the numerator and denominator degrees of freedom of  $f$ , and so its effect on  $F^C(0.05)$  and the bounds depends on  $T_1$  and  $T_2$ .

The usefulness of Tables A1 - A4 is best seen with an example. Gujarati (1972) discusses<sup>5</sup> a structural shift in the relationship between unemployment (UN) and vacancies (VAC) in Great Britain in 1966. Using the quarterly data given by Gujarati (1988, p.465) we have fitted the model  $UN_i = \beta_1 + \beta_2 VAC_i + u_i$  over the periods 1959(2) - 1970(2), 1959(2) - 1966(3) and 1966(4) - 1970(2). The corresponding OLS sums of squared residuals are 2.72600, 0.22581 and 0.35241. As  $T_1 = 30$ ,  $T_2 = 15$  and  $K = 2$ , the Chow test statistic takes the value 76.147. The corresponding 5% tabulated F-value (for 2 and 4 degrees of freedom) is 3.226, so we would reject the null hypothesis of structural stability. However, dividing the sub-sample sums of squared residuals by their respective degrees of freedom, and taking their ratio, gives an estimate

of  $\hat{\theta} = 0.297$ . This suggests that the regression errors are moderately heteroscedastic. From Table A1, when  $\theta = 0.1$ ,  $C_L(0.05) = 0.434$  and  $C_U(0.05) = 4.959$ . As  $76.147 > 4.959$  we can reject the null hypothesis when  $0.1 < \theta < 1$ , given the patterns in the bounds. In this case, the outcome of the test is not affected by the heteroscedasticity.

## V. CONCLUSIONS

While it is widely recognised that the Chow test for structural stability is invalid in the face of heteroscedastic regression errors, it continues to be used widely. To compensate for this, our tabulated critical value bounds should help applied researchers. However, they also illustrate that the appropriate choice of critical value in this case can be dramatically different from the assumed one. This highlights the extent to which a conventional application of the Chow test can be distortive, regardless of the data matrix, when the errors are heteroscedastic.

These bounds apply only to that form of the test which allows for a structural shift in the full coefficient vector, and where there are positive degrees of freedom in each sub-sample. The methods we have described can also be used if these requirements are relaxed, but this would necessitate a very extensive set of tables.

The same approach is not fruitful as a means of bounding the power of the test. It is easily shown that under the alternative hypothesis, the bounds are no longer independent of the data, so they are of little value. However, Kiviet's approach can be used on a wide range of other testing problems of importance to applied econometricians, and work in progress considers some other such cases for various forms of model misspecification.

## FOOTNOTES

- We are grateful to Judy Giles and John Small for their helpful comments, and to Robert Davies for supplying the FORTRAN code for his algorithm.
- 1. As  $f$  in (2) is invariant to scale, the bounds are independent of the separate values of  $\sigma_1^2$  and  $\sigma_2^2$ .
- 2. A glimpse of some bounds for the critical value of the form of the Chow test considered here, when the errors are AR(1) can be obtained from the entries for (his)  $k = 4$ ,  $h = 2$  in Kiviet's Table 5.
- 3. Corresponding tables of bounds on  $\alpha_0$  are available from the authors on request. The  $\alpha_0$  values reported in Table II of Schmidt and Sickles (1977, p.1296) lie within the appropriate bounds in our table.
- 4. Similar exceptions arise for critical value upper bounds reported for the F-test with ARMA (1,1) errors in Table 7 of Kiviet (1980, p.357).
- 5. See, also, Gujarati (1988, pp.449-450).



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## APPENDIX

Bounds on Chow Test Critical Values  
When Errors Are HeteroscedasticTABLE A1.-NOMINAL SIGNIFICANCE LEVEL = 5%;  $k = 2$ 

		$\theta$									
		1	0.01		0.1		10		100		
$T_1$	$T_2$		$C_L$	$C_U$	$C_L$	$C_U$	$C_L$	$C_U$	$C_L$	$C_U$	
10	10	3.634	0.065	13.134	0.595	9.884	0.595	9.884	0.065	13.134	
15	10	3.467	0.051	7.486	0.487	6.624	0.726	11.486	0.084	16.887	
20	10	3.369	0.045	5.852	0.436	5.428	0.841	12.802	0.104	20.494	
25	10	3.305	0.042	5.085	0.407	4.819	0.943	13.910	0.123	23.967	
30	10	3.259	0.040	4.642	0.388	4.453	1.034	14.861	0.141	27.314	
35	10	3.226	0.039	4.354	0.375	4.208	1.117	15.690	0.159	30.545	
40	10	3.200	0.037	4.151	0.365	4.034	1.191	16.422	0.177	33.667	
15	15	3.369	0.063	9.198	0.578	7.718	0.577	7.718	0.063	9.198	
20	15	3.305	0.053	6.946	0.505	6.224	0.660	8.698	0.075	10.885	
25	15	3.259	0.048	5.887	0.462	5.439	0.737	9.582	0.087	12.547	
30	15	3.226	0.045	5.274	0.434	4.959	0.808	10.386	0.098	14.185	
35	15	3.200	0.043	4.875	0.414	4.636	0.875	11.121	0.110	15.799	
40	15	3.179	0.041	4.595	0.399	4.403	0.937	11.797	0.121	17.390	
20	20	3.259	0.061	8.031	0.569	6.963	0.569	6.963	0.061	8.031	
25	20	3.226	0.055	6.684	0.514	6.026	0.630	7.652	0.071	9.106	
30	20	3.200	0.050	5.904	0.478	5.444	0.687	8.297	0.079	10.171	
35	20	3.179	0.047	5.395	0.452	5.048	0.742	8.901	0.087	11.228	
40	20	3.162	0.045	5.037	0.433	4.762	0.790	9.469	0.095	12.275	
25	25	3.200	0.061	7.476	0.564	6.583	0.564	6.583	0.061	7.476	
30	25	3.179	0.056	6.530	0.520	5.909	0.612	7.112	0.068	8.263	
35	25	3.162	0.052	5.912	0.488	5.446	0.657	7.616	0.075	9.045	
40	25	3.148	0.049	5.478	0.465	5.110	0.701	8.096	0.081	9.822	
30	30	3.162	0.061	7.153	0.561	6.355	0.561	6.355	0.061	7.153	
35	30	3.148	0.056	6.428	0.524	5.831	0.600	6.784	0.067	7.773	
40	30	3.136	0.053	5.918	0.496	5.447	0.638	7.196	0.072	8.390	
35	35	3.136	0.061	6.941	0.558	6.203	0.558	6.203	0.061	6.941	
40	35	3.126	0.057	6.356	0.527	5.776	0.592	6.563	0.065	7.453	
40	40	3.117	0.061	6.792	0.557	6.095	0.557	6.095	0.061	6.792	

TABLE A2.-NOMINAL SIGNIFICANCE LEVEL = 5%; k = 3

		$\theta$									
		1	0.01		0.1		10		100		
$T_1$	$T_2$		$C_L$	$C_u$	$C_L$	$C_u$	$C_L$	$C_u$	$C_L$	$C_u$	
10	10	3.344	0.052	20.713	0.484	12.131	0.484	12.131	0.052	20.713	
15	10	3.127	0.041	7.973	0.401	6.736	0.607	13.591	0.069	26.795	
20	10	3.009	0.037	5.668	0.362	5.167	0.713	14.705	0.087	32.367	
25	10	2.934	0.035	4.738	0.340	4.450	0.807	15.596	0.104	37.506	
30	10	2.883	0.033	4.239	0.326	4.044	0.890	16.334	0.121	42.273	
35	10	2.845	0.032	3.929	0.316	3.784	0.964	16.958	0.137	46.718	
40	10	2.816	0.031	3.718	0.309	3.603	1.032	17.496	0.153	50.879	
15	15	3.009	0.052	9.962	0.483	7.853	0.483	7.853	0.052	9.962	
20	15	2.934	0.045	6.810	0.423	5.954	0.558	8.823	0.063	11.909	
25	15	2.883	0.041	5.535	0.389	5.049	0.628	9.677	0.073	13.815	
30	15	2.845	0.038	4.849	0.366	4.524	0.692	10.437	0.083	15.681	
35	15	2.816	0.036	4.423	0.351	4.184	0.751	11.119	0.094	17.511	
40	15	2.794	0.035	4.132	0.339	3.945	0.807	11.735	0.104	19.304	
20	20	2.883	0.052	7.940	0.481	6.676	0.481	6.676	0.052	7.940	
25	20	2.845	0.046	6.325	0.435	5.611	0.536	7.340	0.060	9.058	
30	20	2.816	0.043	5.456	0.405	4.982	0.587	7.954	0.067	10.164	
35	20	2.794	0.040	4.915	0.384	4.568	0.635	8.524	0.074	11.257	
40	20	2.776	0.038	4.546	0.368	4.277	0.681	9.055	0.082	12.339	
25	25	2.816	0.053	7.110	0.480	6.140	0.480	6.140	0.052	7.110	
30	25	2.794	0.047	6.060	0.443	5.405	0.522	6.639	0.058	7.889	
35	25	2.776	0.044	5.405	0.416	4.938	0.563	7.111	0.064	8.663	
40	25	2.761	0.042	4.958	0.397	4.597	0.602	7.559	0.069	9.431	
30	30	2.775	0.052	6.660	0.479	5.835	0.479	5.835	0.052	6.660	
35	30	2.761	0.048	5.893	0.448	5.295	0.514	6.233	0.057	7.257	
40	30	2.748	0.045	5.368	0.424	4.908	0.548	6.615	0.061	7.851	
35	35	2.748	0.052	6.378	0.478	5.638	0.478	5.638	0.052	6.378	
40	35	2.737	0.049	5.778	0.451	5.209	0.508	5.970	0.056	6.862	
40	40	2.728	0.052	6.185	0.478	5.501	0.478	5.501	0.052	6.185	

TABLE A3.-NOMINAL SIGNIFICANCE LEVEL = 5%; k = 4

		θ									
T <sub>1</sub>	T <sub>2</sub>	1	0.01		0.1		10		100		
			C <sub>L</sub>	C <sub>U</sub>	C <sub>L</sub>	C <sub>U</sub>	C <sub>L</sub>	C <sub>U</sub>	C <sub>L</sub>	C <sub>U</sub>	
10	10	3.259	0.041	62.560	0.402	18.127	0.402	18.127	0.041	62.560	
15	10	2.965	0.035	9.652	0.339	7.518	0.523	18.610	0.058	73.853	
20	10	2.817	0.031	5.888	0.311	5.237	0.626	19.020	0.071	82.573	
25	10	2.728	0.030	4.667	0.295	4.335	0.717	19.367	0.091	89.669	
30	10	2.668	0.029	4.071	0.285	3.860	0.796	19.662	0.107	95.646	
35	10	2.626	0.028	3.720	0.277	3.569	0.867	19.916	0.122	100.840	
40	10	2.594	0.027	3.489	0.272	3.372	0.931	20.138	0.138	105.339	
15	15	2.817	0.044	12.276	0.418	8.707	0.418	8.707	0.044	12.276	
20	15	2.728	0.039	7.174	0.369	6.062	0.490	9.695	0.054	14.812	
25	15	2.668	0.035	5.507	0.341	4.944	0.556	10.535	0.064	17.266	
30	15	2.626	0.033	4.693	0.322	4.338	0.617	11.262	0.074	19.644	
35	15	2.594	0.032	4.213	0.309	3.962	0.673	11.900	0.083	21.948	
40	15	2.570	0.031	3.896	0.300	3.705	0.725	12.465	0.093	24.183	
20	20	2.668	0.045	8.442	0.423	6.804	0.423	6.804	0.045	8.442	
25	20	2.626	0.041	6.340	0.384	5.510	0.474	7.476	0.052	9.692	
30	20	2.594	0.037	5.311	0.359	4.792	0.523	8.088	0.059	10.924	
35	20	2.570	0.035	4.703	0.341	4.338	0.568	8.649	0.066	12.140	
40	20	2.550	0.034	4.303	0.327	4.026	0.611	9.165	0.073	13.339	
25	25	2.594	0.046	7.165	0.426	6.039	0.426	6.039	0.046	7.165	
30	25	2.570	0.042	5.924	0.394	5.222	0.465	6.534	0.051	7.984	
35	25	2.550	0.039	5.191	0.371	4.699	0.503	6.998	0.056	8.795	
40	25	2.534	0.037	4.707	0.354	4.336	0.540	7.435	0.062	9.600	
30	30	2.550	0.046	6.535	0.427	5.631	0.427	5.631	0.046	6.535	
35	30	2.534	0.043	5.676	0.400	5.045	0.460	6.020	0.051	7.141	
40	30	2.520	0.040	5.110	0.379	4.636	0.491	6.391	0.055	7.444	
35	35	2.520	0.046	6.160	0.428	5.378	0.428	5.378	0.046	6.160	
40	35	2.509	0.043	5.512	0.404	4.925	0.455	5.698	0.050	6.641	
40	40	2.499	0.046	5.911	0.429	5.206	0.429	5.206	0.046	5.911	

TABLE A4.-NOMINAL SIGNIFICANCE LEVEL = 5%; k = 5

		$\theta$										
$T_1$	$T_2$	1		0.01		0.1		10		100		
		$C_L$	$C_U$	$C_L$	$C_U$	$C_L$	$C_U$	$C_L$	$C_U$	$C_L$	$C_U$	
10	10	3.326	0.033	332.582	0.333	33.258	0.333	33.258	0.333	33.258	0.033	332.582
15	10	2.901	0.029	14.108	0.290	9.292	0.455	29.013	0.455	29.013	0.049	290.130
20	10	2.711	0.027	6.519	0.271	5.587	0.559	27.109	0.559	27.109	0.065	271.090
25	10	2.603	0.026	4.783	0.260	4.377	0.648	26.030	0.648	26.030	0.081	260.299
30	10	2.534	0.025	4.037	0.253	3.797	0.727	25.336	0.727	25.336	0.096	253.356
35	10	2.485	0.025	3.625	0.249	3.460	0.796	24.851	0.796	24.851	0.111	248.515
40	10	2.449	0.025	3.364	0.245	3.241	0.858	24.495	0.858	24.495	0.125	244.947
15	15	2.711	0.038	18.196	0.367	10.516	0.367	10.516	0.367	10.516	0.038	18.196
20	15	2.603	0.034	8.069	0.327	6.482	0.438	11.470	0.438	11.470	0.048	22.034
25	15	2.534	0.031	5.710	0.304	5.019	0.502	12.244	0.502	12.244	0.057	25.647
30	15	2.485	0.030	4.693	0.289	4.290	0.561	12.891	0.561	12.891	0.066	29.059
35	15	2.449	0.028	4.131	0.279	3.858	0.615	13.443	0.615	13.443	0.075	32.289
40	15	2.422	0.028	3.776	0.272	3.574	0.665	13.922	0.665	13.922	0.084	35.354
20	20	2.534	0.040	9.589	0.380	7.266	0.380	7.266	0.380	7.266	0.040	9.589
25	20	2.485	0.036	6.626	0.346	5.608	0.429	7.960	0.429	7.960	0.046	11.080
30	20	2.449	0.034	5.344	0.324	4.753	0.475	8.581	0.475	8.581	0.053	12.544
35	20	2.422	0.032	4.635	0.309	4.237	0.519	9.141	0.519	9.141	0.060	13.980
40	20	2.400	0.031	4.187	0.297	3.894	0.560	9.649	0.560	9.649	0.066	15.391
25	25	2.449	0.041	7.532	0.386	6.151	0.386	6.152	0.386	6.152	0.041	7.532
30	25	2.422	0.038	5.991	0.357	5.190	0.424	6.656	0.424	6.656	0.046	8.430
35	25	2.401	0.035	5.136	0.337	4.599	0.460	7.125	0.460	7.125	0.051	9.318
40	25	2.383	0.034	4.595	0.322	4.202	0.495	7.562	0.495	7.562	0.056	10.197
30	30	2.400	0.042	6.633	0.389	5.602	0.389	5.602	0.389	5.602	0.042	6.633
35	30	2.383	0.039	5.635	0.365	4.946	0.420	5.993	0.420	5.993	0.046	7.270
40	30	2.363	0.037	5.003	0.347	4.499	0.450	6.363	0.450	6.363	0.050	7.903
35	35	2.368	0.042	6.131	0.392	5.277	0.392	5.277	0.392	5.277	0.042	6.131
40	35	2.356	0.040	5.408	0.370	4.786	0.418	5.594	0.418	5.594	0.046	6.624
40	40	2.346	0.042	5.812	0.393	5.062	0.393	5.062	0.393	5.062	0.042	5.812

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