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**Pedro L. Valls Pereira** 



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Pedro L. Valls Pereira

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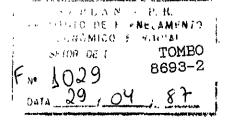
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# TESTING FOR FIRST ORDER SERIAL CORRELATION IN TEMPORALLY AGGREGATED REGRESSION MODELS

# Pedro L. Valls Pereira<sup>1</sup> INPES - IPEA

ABSTRACT

1,71: 1,22 111

This paper shows that the LM statistic for testing first order serial correlation in

regression models can be computed using the Kalman Filter.

It is shown that when there are missing observations, the LM statistic for this test is equivalent to the test statistic derived by Robinson (1985) using the likelihood conditional on the observation times.

The Kalman Filter approach is preferable because the test statistic for first order serial correlation in temporally aggregated regression models can be obtained as an extension of the previous case.

KEYWORDS: First Order Serial Correlation; LM test; Regression Models; Missing Observations; Temporal Aggregation; Kalman Filter.

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## 1 INTRODUCTION

The most common test against autocorrelated errors in regression models is the bounds test of Durbin & Watson (1950, 1951, 1971). Many authors - Breusch (1978) and Godfrey (1978) among others- observed that the Durbin's h statistic ( see Durbin (1970)) can be derived from the general theory of the LM test procedure.

When observations are missing in regression models, a modified Durbin - Watson

statistic was proposed by Savin & White (1978) d Dufour & Dagenais (1985).

In a recent paper Robinson (1985) applied the LM principle and derived two test statistics, where one is obtained using an unconditional form of the likelihood and the other using the likelihood conditional on the observation times. Asymptotic distributions of the test statistics are established and analytic comparison of efficiencies are made.

The motivation for using the LM approach is that it produces an asymptotically locally most powerful test and the  $\chi^2$  asymptotic approximation to its distribution is robust to

departures from normality.

This paper follows Robinson's approach and a LM statistic for testing serial correlation in regression models with missing observations is obtained. This test statistic is obtained as a by - product of the Kalman Filter recursions. It is shown that a reparametrization of the parameter space is useful in obtaining this test statistic.

This approach is used because the extention to temporally aggregateds regression

models can be done very easily.

The structure of this paper is as follows. In section 2, the LM statistic for testing serial correlation in regression models with missing observations is obtained.

Section 3 extends the results to temporally aggregated regression models.

Concluding remarks are made in Section 4.

# 2 THE LM STATISTIC FOR TESTING FIRST ORDER SERIAL CORRELATION IN REGRESSION MODELS WITH MISSING OBSERVATIONS

Consider the regression model with autocorrelated errors, i. e.:

$$y(t) = x'(t)\beta + u(t)$$
  $t = 1, ..., T$  (2.1)

$$u(t) = \rho u(t-1) + \epsilon(t) \tag{2.2}$$

where x(t) is a k×1 vector of nonstochastic regressors, independent of u(t),  $\beta$  is a k×1 vector of unknown parameters and  $\epsilon(t)$  is i.i.d. N(0,  $\sigma^2$ ).

It is assumed that y(t) is observed each m time periods (sometimes called "skip sampling"), i. e.  $y_{\tau} = y(mt)$   $\tau = 1, \ldots, [\frac{T}{m}]$  is the observed endogenous variable and x(t) is

observed for each time period.

The proposition below proves that the regression model for the observed data has parameters  $\beta$ ,  $\rho_{\star} = \rho^{m}$  and  $\sigma_{\star} = (1 + \rho^{2} + \ldots + \rho^{2(m-1)})\sigma^{2}$ . Therefore the log likelihood function for this regression model will be a function of the new parameter set  $\Psi = (\rho_{\star}, \beta', \sigma_{\star}^{2})$  and the LM statistic for testing  $H_{0}: \rho = 0$  vs  $H_{a}: \rho \neq 0$  will be obtained using this reparametrization, because testing  $\rho = 0$  is equivalent to test  $\rho_{\star} = 0$ .

# Proposition I.

Consider the regression model with autocorrelated errors given by (2.1-2). If the endonegous variable y(t) is observed only for  $t \equiv 0 \pmod{m}$ , the observed model is given by:

 $y_{\tau} = x_{\tau}^{\prime} \beta + u_{\tau} \quad \tau = 1, \dots, \left[ \frac{T}{m} \right]$  (2.3)

$$u_{\tau} = \rho_{\tau} u_{\tau-1} + \epsilon_{\tau}^{\star} \tag{2.4}$$

and  $\epsilon_r^*$  is i. i. d. N (0,  $\sigma_\star^2$ ) where  $\sigma_\star^2 = (1 + \rho^2 + \ldots + \rho^{2(m-1)})\sigma^2$  and  $\rho_\star = \rho^m$ .

Proof: see Appendix I

In order to obtain the log likelihood function for the model (2.3-4) the Kalman Filter recursions will be used.

The state space representation of the model (2.3-4) is given by the measurement equation, i. e.:

$$y_{\tau} = (1,0)\alpha_{\tau} \quad \tau = 1, \dots, \left[\frac{T}{m}\right]$$

$$\Leftrightarrow y_{\tau} = Z_{\tau}\alpha_{\tau}$$
(2.5)

and the transition equation, i. e.:

$$\alpha_{r} = \begin{bmatrix} 0 & \rho_{\star} \\ 0 & \rho_{\star} \end{bmatrix} \alpha_{r-1} + \begin{bmatrix} x_{r}^{t} \beta \\ 0 \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \end{bmatrix} \epsilon_{r}^{\star}$$

$$\Leftrightarrow \alpha_{r} = \Phi \alpha_{r-1} + c_{r} + R \epsilon_{r}^{\star}$$
(2.6)

where  $Var(\epsilon_r^*) = \sigma_*^2 Q$ .

The log likelihood function can be obtained using the prediction error decomposition (see Harvey (1981)) and is given by:

$$l(\Psi) = \ln L(\rho_{\star}, \beta', \sigma_{\star}^{2}) = -\frac{T}{2m} \ln 2\pi - \frac{T}{2m} \ln \sigma_{\star}^{2}$$

$$-\frac{1}{2} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \ln f_{\tau} - \frac{1}{2\sigma_{\star}^{2}} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \frac{\nu_{\tau}^{2}}{f_{\tau}}$$
(2.7)

where  $\nu_{\tau}$  is the prediction error and  $\sigma_{\star}^2 f_{\tau}$  is the variance of the prediction error and they are obtained by the Kalman Filter recursions (see Harvey (1931) pages 107 - 111). These recursions are given by two sets of equations: the prediction equations and the updating equations. In order to start these recursions, the mean,  $a_0$ , and the variance – covariance matrix,  $\sigma_{\star}^2 P_0$ , of the state space vector at time t = 0 are needed. The log likelihood (2.7) is obtained when these staring values are given by:

$$a_0 = 0$$

$$P_0 = \begin{bmatrix} \kappa & 0 \\ 0 & \frac{1}{1-\rho_s^2} \end{bmatrix}$$
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where  $\kappa$  is a large number, e. g.  $\kappa = 10^6$ , which is equivalent to use a diffuse prior for the initial condictions (see Ansley & Kohn (1985)).

It is easy to see that

$$\nu_1 = y_1 - x_1' \beta \tag{2.8}$$

$$\nu_{\tau} = (y_{\tau} - x_{\tau}'\beta) - \rho_{\star}(y_{\tau-1} - x_{\tau-1}'\beta) \quad \tau = 2, \dots, \left[\frac{T}{m}\right]$$
 (2.9)

and

$$f_1 = \frac{1}{1 - \rho^2} \tag{2.10}$$

$$f_{\tau} = 1 \qquad \tau = 2, \dots, \left[\frac{T}{m}\right] \tag{2.11}$$

In order to derive the LM statistic for testing for first order autocorrelation, the first derivatives of the log likelihood function and the information matrix are needed. Pagan (1978) or Engle & Watson (1981) have shown that the information matrix can be computed using only first derivatives of the likelihood function. These derivatives are computed using auxiliary recursions (see Appendix II).

By Breusch & Pagan (1978), the LM statistic for testing for first order autocorrelation

is given by:

$$LM = \widetilde{D}_1 \left( \widetilde{I}_{\rho_{\star}\rho_{\star}} - \widetilde{I}_{\rho_{\star}\beta} \widetilde{I}_{\beta\beta}^{-1} \widetilde{I}_{\beta\rho_{\star}} \right)^{-1} \widetilde{D}_1$$
 (2.12)

where

$$\widetilde{D} = \begin{bmatrix} \frac{\partial l}{\partial \rho_*} \\ \frac{\partial l}{\partial \beta} \\ \frac{\partial l}{\partial \sigma_*^2} \end{bmatrix} (\widetilde{\Psi}) = \begin{bmatrix} \widetilde{D}_1 \\ 0 \\ 0 \end{bmatrix}$$

and

$$\widetilde{I} = \begin{bmatrix} I_{\rho_{+}\rho_{+}} & I_{\rho_{+}\beta} & I_{\rho_{+}\sigma_{+}^{2}} \\ I_{\beta\rho_{+}} & I_{\beta\beta} & I_{\beta\sigma_{+}^{2}} \\ I_{\sigma_{+}^{2}\rho_{+}} & I_{\sigma_{+}^{2}\beta} & I_{\sigma_{+}^{2}\sigma_{+}^{2}} \end{bmatrix} (\widetilde{\Psi})$$

is the Fisher information matrix evaluated at  $\widetilde{\Psi} = \Psi_{H_0} = (0, \widetilde{\beta}, \widetilde{\sigma}_{\star}^2)$  and

$$\widetilde{\beta} = \left[ (X^*)'(X^*) \right]^{-1} (X^*)'Y^*$$

$$X^* = (x(m), x(2m), \dots, x(T))$$

$$Y^* = (y(m), y(2m), \dots, y(T))$$

$$\widetilde{\sigma}_{\star}^2 = \frac{1}{T} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \left( y_{\tau} - x_{\tau}' \widetilde{\beta} \right)^2 = \frac{1}{T} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \widetilde{u}_{\tau}^2$$

From Appendix III it follows that:

$$LM = \frac{T}{m} (r_1^*)^2 (2.13)$$

where

$$r_1^{\star} = \frac{\sum_{\tau=1}^{\left[\frac{T}{m}\right]} \widetilde{u}_{\tau} \widetilde{u}_{\tau-1}}{\sum_{\tau=1}^{\left[\frac{T}{m}\right]} \widetilde{u}_{\tau}} \tag{2.14}$$

which is exactly the s<sub>1</sub> statistic derived by Robinson (1985).

# 3 THE LM STATISTICS FOR TESTING FIRST ORDER SERIAL CORRELATION IN TEMPORALLY AGGREGATED REGRESSION MODELS

Now it is assumed that the endogenous variable y(t) in (2.1) is a flow variable and instead of observing y(t) every time period, is  $y_t$  which is observed, i. e.:

$$y_{\tau} = \sum_{i=0}^{m-1} y(m\tau - j) \quad \tau = 1, ..., \left[\frac{T}{m}\right]$$
 (3.1)

Given that model (2.1-2) is the disaggregated model, i. e. for times t = 1, ..., t, Proposition II below derives the aggregated model, i. e. for  $\tau = 1, ..., \left[\frac{T}{m}\right]$  and the endogenous variable is given by (3.1).

# Proposition II.

Given model (2.1-2) and assume that  $y_{\tau}$  is given by (3.1). Then the aggregated model is given by:

$$y_{\tau} = z_{\tau}' \beta + u_{\tau} \quad \tau = 1, \dots, \left[ \frac{T}{m} \right]$$
 (3.2)

$$u_r = \rho_{\star} u_{r-1} + \epsilon_r^{\star} + \theta_{\star} \epsilon_{r-1}^{\star} \tag{3.3}$$

where

$$z_{\tau}' = \sum_{j=0}^{m-1} x'(m\tau - j)$$

$$\rho_{\star} = \rho^{m}$$

$$\epsilon_{\tau}^{\star} \sim N(0, \sigma_{\tau}^{2})$$

and  $\theta_{\star}$  and  $\sigma_{\star}^2$  are given by the solutions of the following equations:

$$(1+\theta_{\star}^{2})\sigma_{\star}^{2} = [1+(1+\rho)^{2}+(1+\rho+\rho^{2})^{2}+\ldots+(1+\rho+\ldots+\rho^{m-1})^{2}+ + \rho^{2}(1+\rho+\ldots+\rho^{m-2})^{2}+\rho^{4}(1+\rho+\ldots+\rho^{m-3})^{2}+ + \ldots+\rho^{2(m-1)}]\sigma^{2}$$

$$(3.4)$$

$$\theta_{\star}\sigma_{\star}^{2} = [\rho(1+\rho+\ldots+\rho^{m-2})+(1+\rho)\rho^{2}(1+\rho+\ldots+\rho^{m-3})+ + \ldots+(1+\rho+\ldots+\rho^{m-3})\rho^{m-2}(1+\rho)+ + (1+\rho+\ldots+\rho^{m-2})\rho^{m-1}]\sigma^{2}$$

$$(3.5)$$

$$|\theta_{\star}| < 1$$

Proof: : see Appendix IV

The LM statistic for testing first order serial correlation will be derived using the Kalman Filter recursions. In order to use these recursions the state space representation of model (3.2-3) is needed.

This representation is given by the measurement equation, i. e.:

$$y_{\tau} = (0, 0, 1) \alpha_{\tau} \quad \tau = 1, \dots, \left[ \frac{T}{m} \right]$$

$$\Leftrightarrow y_{\tau} = Z_{\tau} \alpha_{\tau} \tag{3.7}$$

and the transition equation, i. e.:

$$\alpha_{\tau} = \begin{bmatrix} \rho_{\star} & 1 & 0 \\ 0 & 0 & 0 \\ \rho_{\star} & 1 & 0 \end{bmatrix} \alpha_{\tau-1} + \begin{bmatrix} 0 \\ 0 \\ z_{\tau}^{t} \beta \end{bmatrix} + \begin{bmatrix} 1 \\ \theta_{\star} \\ 1 \end{bmatrix} \epsilon_{\tau}^{\star}$$

$$\Rightarrow \alpha_{\tau} = \Phi \alpha_{\tau-1} + c_{\tau} + R \epsilon_{\tau}^{\star}$$

$$(3.8)$$

where  $Var(\epsilon_{\tau}^{\star}) = \sigma_{\star}^{2}Q$ .

The Kalman Filter recursions are started by

$$a_{0} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$P_{0} = \begin{bmatrix} \frac{1+2\rho_{*}\theta_{*}+\theta_{*}^{2}}{1-\rho_{*}^{2}} & \theta_{*} & 0 \\ \theta_{*} & \theta_{*}^{2} & 0 \\ 0 & 0 & \kappa \end{bmatrix}$$

and the log likelihood function is given by (2.7).

It is easy to see that:

$$\nu_1 = y_1 - z_1' \beta \tag{3.9}$$

$$\nu_{\tau} = (y_{\tau} - z_{\tau}^{\prime} \beta) - \rho_{\star} (y_{\tau-1} - z_{\tau-1}^{\prime} \beta) - \theta_{\star} f_{\tau-1}^{-1} \nu_{\tau-1} \quad \tau = 2, \dots, \left[ \frac{T}{\tau} \right]$$
 (3.10)

and

$$f_1 = \frac{1 + 2\rho_{\star}\theta_{\star} + \theta_{\star}^2}{1 - \rho_{\star}^2} \tag{3.11}$$

$$f_{\tau} = 1 + \theta_{\star}^{2} \left( 1 - f_{\tau-1}^{-1} \right) \qquad \tau = 2, \dots, \left[ \frac{T}{m} \right]$$
 (3.12)

It is easy to see that under  $H_0$  ( and under some conditions on the x(t)), i. e.  $\rho = 0 \Leftrightarrow \rho_{\star} = 0$ :

(i) the OLS estimate of  $\beta$ , i. e.  $\widetilde{\beta}$  is given by:

$$\widetilde{\beta} = (Z'Z)^{-1}Z'Y^* \tag{3.13}$$

where  $Z = (z_1, \ldots, z_{\lceil \frac{n}{2} \rceil}), z_{\tau} = \sum_{j=0}^{m-1} x(m\tau - j);$ 

(ii) the prediction error is given  $\nu_{\tau} = y_{\tau} - z_{\tau}' \widetilde{\beta} \quad \forall \tau$ ; (iii) the variance of the prediction error is given by  $f_{\tau} = 1$  (iv) and  $\widetilde{\sigma}_{\star}^2 = m\widetilde{\sigma}^2 = \frac{1}{L} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} (y_{\tau} - z_{\tau}' \widetilde{\beta})^2 = \frac{1}{L} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \widetilde{u}_{\star}^2$ . From Appendix V, it follows that:

$$LM = \frac{T}{m} (r_1^*)^2 {(3.14)}$$

where

$$r_1^{\star} = \frac{\sum_{\tau=1}^{\left[\frac{T}{m}\right]} \widetilde{u}_{\tau} \widetilde{u}_{\tau-1}}{\sum_{\tau=1}^{\left[\frac{T}{m}\right]} \widetilde{u}_{\tau}} \tag{3.15}$$

# **4 CONCLUDING REMARKS**

It was shown that the LM statistic for testing first order serial correlation can be derived as a by - product of the Kalman Filter recursions.

In order to use the Kalman Filter recursions to derive the LM statistic a reparametriza-

tion was necessary.

It was shown that the LM statistic for testing first order serial correlation in regression models with missing observations, is equivalent to the s<sub>1</sub> statistic derived by Robinson (1985). This statistic is obtained from the likelihood conditional on the observation times. The LM statistic can be interpreted as  $\left[\frac{T}{m}\right]$  times  $R^2$ , the coefficient of determination of the regression of  $\tilde{u}_{\tau}$  in  $\tilde{u}_{\tau-1}$ , where  $\tilde{u}_{\tau} = y_{\tau} - x_{\tau}^{\prime}$ , i. e. the residuals, under  $H_0$ , of model (2.1) for every m time period. An extension for regression models where the endogenous variable is observed at every time period for  $t=1,\ldots,p_0T$   $(1 \le p_0 < 1)$  and for  $t=p_0T+1,\ldots,T$ is observed at every m time period, is given in Pereira (1985).

Using the Kalman Filter recursions, it was possible to extend this result to temporally aggregated regression models. As before the LM statistic can be interpreted as  $\left[\frac{T}{m}\right]$  times  $R^2$  coefficient of determination of the regression of  $\tilde{u}_{\tau}$  in  $\tilde{u}_{\tau-1}$ , where  $\tilde{u}_{\tau} = y_{\tau} - z_{\tau}'$  and

 $z_{\tau} = \sum_{j=0}^{m-1} x(m\tau - j)$ , i. e. the residuals under  $H_0$  of model (3.2-3).

### APPENDIX I

(2.1-2) can be rewritten as follows:

$$(1 - \rho L)y(t) = (1 - \rho L)x'(t)\beta + \epsilon(t)$$
(A.I.1)

Multiplying both sides of (A.I.1) by  $(1 + \rho L + \ldots + \rho^{m-1} L^{m-1})$ , it follows:

$$(1 - \rho^m L^m)y(t) = (1 - \rho^m L^m)x'(t)\beta + (1 + \rho L + \dots + \rho^{m-1} L^{m-1})\epsilon(t)$$
 (A.I.2)

Now define

$$B = L^{m}$$

$$\rho_{\pi} = \rho^{m}$$

$$\epsilon_{\tau}^{*} = (1 + \rho L + \dots + \rho^{m-1} L^{m-1}) \epsilon(t) \qquad \tau = 1, \dots, \left[ \frac{T}{m} \right]$$

then (A.I.2) can be rewritten as follows:

$$(1 - \rho_{\star} \beta) y_{\tau} = (1 - \rho_{\star} \beta) x_{\tau}^{t} \beta + \epsilon_{\tau}^{\star}$$
 (A.I.3)

1

$$y_{\tau} = x_{\tau}^{\prime} \beta + u_{\tau} \tag{A.I.4}$$

$$u_{\tau} = \rho_{\star} u_{\tau - 1} + \epsilon_{\tau} \tag{A.I.5}$$

Using the results of Tiao(1972) or Pereira(1986) it follows that  $\epsilon_{\tau}^{\star} \sim NID(0, \sigma_{\star}^{2})$  where  $\sigma_{\star}^{2} = (1 + \rho^{2} + \ldots + \rho^{2(m-1)})\sigma^{2}$ .

### APPENDIX II

Taking derivatives of (2.3) with respect to the parameters, it follows that:

$$\frac{\partial l}{\partial \psi_{i}} = \frac{1}{2\sigma_{\star}^{2}} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \frac{\nu_{\tau}^{2}}{f_{\tau}} \frac{\partial f_{\tau}}{\partial \psi_{i}} - \frac{1}{2\sigma_{\star}^{2}} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \frac{\nu_{\tau}}{f_{\tau}} \frac{\partial \nu_{\tau}}{\partial \psi_{i}} - \frac{1}{2} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \frac{1}{f_{\tau}} \frac{\partial f_{\tau}}{\partial \psi_{i}} \quad i = 1, \dots, k+1 \tag{A.II.1}$$

$$\frac{\partial l}{\partial \sigma_{\star}^2} = \frac{T}{2m\sigma_{\star}^2} + \frac{1}{2\sigma_{\star}4} \sum_{\tau=1}^{\left[\frac{T}{4\pi}\right]} \frac{\nu_{\tau}^2}{f_{\tau}} \tag{A.II.2}$$

and the information matix is given by:

$$I_{\psi_{i}\psi_{j}} = \frac{1}{\sigma_{\star}^{2}} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \frac{1}{f_{\tau}} \left[ \frac{\partial \nu_{\tau}}{\partial \psi_{i}} \right] \left[ \frac{\partial \nu_{\tau}}{\partial \psi_{j}} \right]$$

$$- \frac{1}{2} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \frac{1}{f_{\tau}^{2}} \frac{\partial f_{\tau}}{\partial \psi_{i}} \frac{\partial f_{\tau}}{\partial \psi_{j}} \quad i, j = 1, \dots, k+1$$
(A.II.3)

$$I_{\sigma_{\pi}^{2}\psi_{j}} = \frac{1}{\sigma_{\pi}^{4}} \sum_{i=1}^{\left[\frac{\pi}{m}\right]} \frac{\nu_{\tau}}{f_{\tau}} \frac{\partial \nu_{\tau}}{\partial \psi_{i}} - \frac{1}{2\sigma_{\pi}^{4}} \sum_{i=1}^{\left[\frac{\pi}{m}\right]} \frac{\nu_{\tau}^{2}}{f_{\tau}} \frac{\partial f_{\tau}}{\partial \psi_{i}} \quad i = 1, \dots, k+1$$
(A.II.4)

$$I_{\sigma_{\pi}^2 \sigma_{\pi}^2} = \frac{T}{2m\sigma_{\pi}^2} \tag{A.II.5}$$

In order to obtain (A.II.1-4), it is needed auxiliarly recursions for the derivatives. They are:

**Prediction Equations** 

$$\frac{\partial a_{\tau/\tau-1}}{\partial \psi_{i}} = \frac{\partial \Phi}{\partial \psi_{i}} a_{\tau-1} + \Phi \frac{\partial a_{\tau-1}}{\partial \psi_{i}} + \frac{\partial c_{\tau}}{\partial \psi_{i}} 
\frac{\partial P_{\tau/\tau-1}}{\partial \psi_{i}} = \frac{\partial \Phi}{\partial \psi_{i}} P_{\tau-1} \Phi' + \Phi \frac{\partial P_{\tau-1}}{\partial \psi_{i}} \Phi' + \Phi P_{\tau-1} \left(\frac{\partial \Phi}{\partial \psi_{i}}\right)' 
+ \frac{\partial R}{\partial \psi_{i}} Q R' + R \frac{\partial Q}{\partial \psi_{i}} R' + R Q \left(\frac{\partial R}{\partial \psi_{i}}\right)' \quad i = 1, \dots, k+1 \quad (A.II.7)$$

Prediction Error

$$\frac{\partial f_{\tau}}{\partial \psi_{i}} = Z_{\tau}^{i} \frac{\partial P_{\tau/\tau-1}}{\partial \psi_{i}} Z_{\tau} \quad i = 1, \dots, k+1$$
 (A.II.8)

Variance of the Prediction Error

$$\frac{\partial \nu_{\tau}}{\partial \psi_{i}} = -Z_{\tau}^{\prime} \frac{\partial a_{\tau/\tau-1}}{\partial \psi_{i}} \quad i = 1, \dots, k+1$$
 (A.II.9)

**Updating Equations** 

$$\frac{\partial a_{\tau}}{\partial \psi_{i}} = \frac{\partial a_{\tau/\tau-1}}{\partial \psi_{i}} + \frac{\partial P_{\tau/\tau-1}}{\partial \psi_{i}} Z_{\tau} f_{\tau}^{-1} \nu_{\tau} 
- P_{\tau/\tau-1} Z_{\tau} f_{\tau}^{-2} \frac{\partial f_{\tau}}{\partial \psi_{i}} \nu_{\tau} + P_{\tau/\tau-1} Z_{\tau} f_{\tau}^{-1} \frac{\partial \nu_{\tau}}{\partial \psi_{i}} \quad i = 1, \dots, k+1 \quad (A.II.10)$$

$$\frac{\partial P_{\tau}}{\partial \psi_{i}} = \frac{\partial P_{\tau/\tau-1}}{\partial \psi_{i}} - \frac{\partial P_{\tau/\tau-1}}{\partial \psi_{i}} Z_{\tau} f_{\tau}^{-1} Z_{\tau}^{i} P_{\tau/\tau-1} 
+ P_{\tau/\tau-1} Z_{\tau} f_{\tau}^{-2} \frac{\partial f_{\tau}}{\partial \psi_{i}} Z_{\tau}^{i} P_{\tau/\tau-1} 
- P_{\tau/\tau-1} Z_{\tau} f_{\tau}^{-1} Z_{\tau}^{i} \frac{\partial P_{\tau/\tau-1}}{\partial \psi_{i}} \quad i = 1, \dots, k+1 \quad (A.II.11)$$

and the initial conditions are given by:

$$\begin{split} \frac{\partial a_0}{\partial \psi_i} &= 0 \quad i = 1, \dots, k+1 \\ \frac{\partial P_0}{\partial \beta} &= 0 \\ \frac{\partial P_0}{\partial \rho_{\star}} &= \begin{bmatrix} 0 & 0 \\ 0 & \frac{-2\rho_{\star}}{(1-\rho_{\star}^2)^2} \end{bmatrix} \\ \frac{\partial P_0}{\partial \rho_{\star}} &= \begin{bmatrix} \frac{2(\theta_{\star} + (\theta_{\star} + \rho_{\star}) \begin{bmatrix} \frac{\partial \theta_{\star}}{\partial \rho_{\star}} \end{bmatrix})(1-\rho_{\star}^2) + 2\rho_{\star}(1+2\rho_{\star}\theta_{\star} + \theta_{\star}^2)}{(1-\rho_{\star}^2)^2} & \frac{\partial \theta_{\star}}{\partial \rho_{\star}} & 0 \\ \frac{\partial \theta_{\star}}{\partial \rho_{\star}} &= \begin{bmatrix} \frac{2(\theta_{\star} + (\theta_{\star} + \rho_{\star}) \begin{bmatrix} \frac{\partial \theta_{\star}}{\partial \rho_{\star}} \end{bmatrix})(1-\rho_{\star}^2) + 2\rho_{\star}(1+2\rho_{\star}\theta_{\star} + \theta_{\star}^2)}{(1-\rho_{\star}^2)^2} & \frac{\partial \theta_{\star}}{\partial \rho_{\star}} & 0 \\ 0 & 0 & 0 \end{bmatrix} \end{split}$$

### APPENDIX III

It is easy to see that

$$\frac{\partial \nu_{\tau}}{\partial \psi_{i}} = \begin{bmatrix} \frac{\partial \nu_{\tau}}{\partial \rho_{\star}}, \frac{\partial \nu_{\tau}}{\partial \beta}, 0 \end{bmatrix} = \begin{cases} (0, -x'_{1}, 0) & \tau = 1 & \text{(A.III.1)} \\ -(y_{\tau-1} - x'_{\tau-1}\beta, x'_{\tau} - \rho_{\star} x'_{\tau-1}, 0) & \tau > 1 & \text{(A.III.2)} \end{cases}$$

$$\frac{\partial f_{\tau}}{\partial \psi_{i}} = \begin{bmatrix} \frac{\partial f_{\tau}}{\partial \rho_{\star}}, \frac{\partial f_{\tau}}{\partial \beta}, 0 \end{bmatrix} = \begin{cases} (\frac{-2\rho_{\star}}{1 - \rho_{\star}^{2}}, 0, 0) & \tau = 1 & \text{(A.III.3)} \\ 0 & \tau > 1 & \text{(A.III.4)} \end{cases}$$
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Substituting (A.III.1-4) and (2.8-11) into (A.II.1), it follows that:

$$\frac{\partial l}{\partial \rho_{\star}} = \frac{-\rho_{\star}(y_{1} - x_{1}^{l}\beta)^{2}}{\sigma_{\star}^{2}} \\
+ \frac{1}{\sigma_{\star}^{2}} \sum_{\tau=2}^{\left[\frac{T}{m}\right]} \left[ (y_{\tau} - x_{\tau}^{l}\beta) - \rho_{\star}(y_{\tau-1} - x_{\tau-1}^{l}\beta) \right] (y_{\tau-1} - x_{\tau-1}^{l}\beta) \\
+ \frac{\rho_{\star}}{1 - \rho_{\star}^{2}} \\
\frac{\partial l}{\partial \beta} = \frac{1 - \rho_{\star}^{2}}{\sigma_{\star}^{2}} x_{1}^{l} (y_{1} - x_{1}^{l}\beta) \\
+ \frac{1}{\sigma_{\star}^{2}} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} (x_{\tau} - \rho_{\star} x_{\tau}^{l})^{l} \left[ (y_{\tau} - x_{\tau}^{l}\beta) - \rho_{\star}(y_{\tau-1} - x_{\tau-1}^{l}\beta) \right] \tag{A.III.6}$$

Substituting (2.8-11) into (A.II.2), it follows that:

$$\begin{split} \frac{\partial l}{\partial \sigma_{\star}^{2}} &= -\frac{T}{2m\sigma_{\star}^{2}} \\ &+ \frac{1 - \rho_{\star}^{2}}{2\sigma_{\star}^{4}} x_{1}^{t} (y_{1} - x_{1}^{t} \beta)^{2} \\ &+ \frac{1}{\sigma_{\star}^{4}} \sum_{\tau=2}^{\left[\frac{T}{m}\right]} \left[ (y_{\tau} - x_{\tau}^{t} \beta) - \rho_{\star} (y_{\tau-1} - x_{\tau-1}^{t} \beta) \right]^{2} \end{split}$$

Substituting (2.8-11) and (A.III.1-4) into (A.II.3), it follows that:

$$I_{\rho_{+}\rho_{+}} = \frac{1}{\sigma_{+}2} \sum_{\tau=2}^{\left[\frac{T}{m}\right]} (y_{\tau-1} - x_{\tau-1}^{t}\beta)^{2}$$

$$+ \frac{2\rho_{+}^{2}}{1 - \rho_{+}^{2}} \qquad (A.III.7)$$

$$I_{\rho_{+}\beta} = \frac{1}{\sigma_{+}^{2}} \sum_{\tau=2}^{\left[\frac{T}{m}\right]} (x_{\tau} - \rho_{+}x_{\tau}^{t})^{t} (y_{\tau-1} - x_{\tau-1}^{t}\beta) \qquad (A.III.8)$$

$$I_{\beta\beta} = \frac{1}{\sigma_{+}^{2}} \sum_{\tau=2}^{\left[\frac{T}{m}\right]} (x_{\tau} - \rho_{+}x_{\tau}^{t}) (x_{\tau} - \rho_{+}x_{\tau-1}^{t})^{t}$$

$$+ \frac{1 - \rho_{+}^{2}}{\sigma_{+}^{2}} (x_{1}x_{1}^{t}) \qquad (A.III.9)$$

Sunstituting (A.III.1-4) and (2.8-11) into (A.II.4), it follows that:

$$I_{\rho_{\pi}\sigma_{\pi}^{2}} = \frac{-\rho_{\pi}}{\sigma_{\pi}^{4}} (y_{1} - x_{1}^{i}\beta)$$

$$- \frac{1}{\sigma_{\pi}^{4}} \sum_{\tau=2}^{\left[\frac{T}{m}\right]} \left[ (y_{\tau} - x_{\tau}^{i}\beta) - \rho_{\pi} (y_{\tau-1} - x_{\tau-1}^{i}\beta) \right] (y_{\tau-1} - x_{\tau-1}^{i}\beta) \quad \text{(A.III.10)}$$

$$I_{\beta\sigma_{\pi}^{2}} = \frac{1 - \rho_{\pi}^{2}}{\sigma_{\pi}^{4}} x_{1}^{i} (y_{1} - x_{1}^{i}\beta)$$

$$- \frac{1}{\sigma_{\pi}^{4}} \sum_{\tau=2}^{\left[\frac{T}{m}\right]} (x_{\tau} - \rho_{\pi} x_{\tau}^{i})^{i} \left[ (y_{\tau} - x_{\tau}^{i}\beta) - \rho_{\pi} (y_{\tau-1} - x_{\tau-1}^{i}\beta) \right] \quad \text{(A.III.11)}$$

Under  $H_0: \rho = 0 \Leftrightarrow H_0; \rho_* = 0$ , it follows that:

$$\frac{\partial l}{\partial \beta}(\psi_{H_0}) = 0$$

$$\frac{\partial l}{\partial \sigma_{\star}^2}(\psi_{H_0}) = 0$$

$$\frac{\partial l}{\partial \rho_{\star}}(\psi_{H_0}) = \frac{1}{\tilde{\sigma}_{\star}^2} \sum_{\tau=2}^{\left[\frac{T}{m}\right]} \tilde{u}_{\tau} \tilde{u}_{\tau-1} \tag{A.III.12}$$

Taking plim's in (A.III.7-11), and under appropriated regularity conditions on the x'(t), it follows that:

$$\begin{aligned} plim \left[ \widetilde{I}_{\rho*\rho*} - \frac{1}{\widetilde{\sigma}_{*}^{2}} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \widetilde{u}_{\tau-1}^{2} \right] &= 0 \\ plim \left[ \widetilde{I}_{\beta\beta} - \frac{1}{\widetilde{\sigma}_{*}^{2}} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} x_{\tau} x_{\tau}^{\prime} \right] &= 0 \\ plim \widetilde{I}_{\rho*\beta} &= 0 \\ plim \widetilde{I}_{\rho*\beta} &= 0 \\ plim \widetilde{I}_{\beta\sigma_{*}^{2}} &= 0 \end{aligned}$$

It follows that

$$LM = \begin{bmatrix} \frac{1}{\widetilde{\sigma}_{\star}^2} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \widetilde{u}_{\tau} \widetilde{u}_{\tau-1} \end{bmatrix} \begin{bmatrix} \frac{1}{\widetilde{\sigma}_{\star}^2} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \widetilde{u}_{\tau-1}^2 \end{bmatrix}^{-1} \begin{bmatrix} \frac{1}{\widetilde{\sigma}_{\star}^2} \sum_{\tau=1}^{\left[\frac{T}{m}\right]} \widetilde{u}_{\tau} \widetilde{u}_{\tau-1} \end{bmatrix}$$

but

$$\left[\frac{1}{\widetilde{\sigma}_{\star}^{2}}\sum_{\tau=1}^{\left[\frac{T}{m}\right]}\widetilde{u}_{\tau-1}^{2}\right]=\frac{T}{m}$$

then

$$LM = \left\lceil \frac{T}{m} \right\rceil^2 r_1^{\star^2} \left\lceil \frac{T}{m} \right\rceil^{-1} = \left\lceil \frac{T}{m} \right\rceil r_1^{\star^2}$$

where

$$r_1^{\star^2} = \frac{\sum_{\tau=1}^{\left[\frac{T}{m}\right]} \widetilde{u}_{\tau} \widetilde{u}_{\tau-1}}{\sum_{\tau=1}^{\left[\frac{T}{m}\right]} \widetilde{u}_{\tau}^2}$$

### APPENDIX IV

# Proof of Proposition II:

Multiplying both sides of (A.I.2) by  $\sum_{j=1}^{m-1} L^j$ , it follows that:

$$(1 - \rho^m B) y_\tau = (1 - \rho^m) (\sum_{j=0}^{m-1} x^j (t - j)) \beta + \omega_\tau$$
 (A.IV.1)

where

$$\omega_{\tau} = \left(\sum_{j=0}^{m-1}\right) \left(\sum_{j=1}^{m-1} \rho^{j} L^{j}\right) \epsilon(t) \tag{A.IV.2}$$

Now (A.IV.1) can be rewritten as:

$$y_{\tau} = z_{\tau}^{\prime} \beta + u_{\tau} \tag{A.IV.3}$$

$$u_{\tau} = \rho_{\star} u_{\tau-1} + \omega_{\tau} \tag{A.IV.4}$$

where

$$z_{\tau}^{\prime} = \sum_{j=0}^{m-1} x(m\tau - j)$$

$$\rho_{\tau} = \rho^{m}$$

and  $\omega_{\tau}$  is given by (A.IV.2).

Using the results of Tiao (1972) or Pereira (1986), (A.IV.4) follows an ARMA(1,1), the expression (3.3), where the AR parameter is  $\rho_{\star}$  and the MA parameter,  $\theta_{\star}$  and the variance of the error term,  $\sigma_{\star}^2$ , are given by the solution of equation (3.4-6).

### APPENDIX V

Derivating (3.9-12) with respect to  $\beta$ , it follows that:

$$\frac{\partial \nu_{\tau}}{\partial \beta} = \begin{cases} -z_1^{\ell} & \text{(A.V.1)} \\ -\left[z_{\tau} - \rho_{\star} z_{\tau-1}\right] + \theta_{\star} \frac{\partial f_{\tau-1}}{\partial \beta} f_{\tau-1}^{-2} \nu_{\tau-1} - \theta_{\star} f_{\tau-1}^{-1} \frac{\partial \nu_{\tau-1}}{\partial \beta} & \tau > 1 \text{ (A.V.2)} \end{cases}$$

and

$$\frac{\partial f_{\tau}}{\partial \beta} = 0 \tag{A.V.3}$$

Now using (A.V.3) into (A.V.2), it follows that:

$$\frac{\partial \nu_{\tau}}{\partial \beta} = \begin{cases} -z_1^t & \text{(A.V.1)} \\ -\left[z_{\tau} - \rho_{\star} z_{\tau-1}\right] - \theta_{\star} f_{\tau-1}^{-1} \frac{\partial \nu_{\tau-1}}{\partial \beta} & \tau > 1 & \text{(A.V.2)} \end{cases}$$

Now derivating (3.9-12) with respect to  $\rho_*$ , it follows that:

$$\frac{\partial \nu_{\tau}}{\partial \rho_{\star}} = \begin{cases} 0 & \tau = 1 \text{ (A.V.4)} \\ -\left[y_{\tau-1} - z_{\tau-1}^{t}\beta\right] - \frac{\partial \theta_{\star}}{\partial \rho_{\star}} f_{\tau-1}^{-1} \nu_{\tau-1} \\ + \theta_{\star} \frac{\partial f_{\tau-1}}{\partial \rho_{\star}} f_{\tau-1}^{-2} \nu_{\tau-1} - \theta_{\star} f_{\tau-1}^{-1} \frac{\partial \nu_{\tau-1}}{\partial \rho_{\star}} & \tau > 1 \text{ (A.V.5)} \end{cases}$$

and

$$\frac{\partial f_{\tau}}{\partial \rho_{\star}} = \begin{cases} \frac{\left[2\theta_{\star} + 2\rho_{\star} \frac{\partial \theta_{\star}}{\partial \rho_{\star}} + 2\theta_{\star} \frac{\partial \theta_{\star}}{\partial \rho_{\star}}\right] (1 - \rho_{\star}^{2}) - 2\rho_{\star} \left[1 + 2\rho_{\star}\theta_{\star} + \theta_{\star}^{2}\right]}{(1 - \rho_{\star}^{2})^{2}} & \tau = 1 \text{ (A.V.6)} \\ 2\theta_{\star} \frac{\partial \theta}{\partial \rho_{\star}} \left(1 - f_{\tau-1}^{-1}\right) + \theta_{\star}^{2} \frac{\partial f_{\tau-1}}{\partial \rho_{\star}} f_{\tau-1}^{-2} & \tau > 1 \text{ (A.V.7)} \end{cases}$$

Derivating (3.9-12) with respect to  $\sigma_{\star}^2$ , it follows that:

$$\frac{\partial f_{\tau}}{\partial \sigma_{\tau}^2} = \frac{\partial \nu_{\tau}}{\partial \sigma_{\tau}^2} = 0$$

Then under  $H_0$ , it follows that:

$$\frac{\partial \nu_{\tau}}{\partial \rho_{\star}}(\psi_{H_0}) = \begin{cases} 0 & \tau = 1 \text{ (A.V.8)} \\ -\frac{m+1}{m} \left[ y_{\tau-1} - z_{\tau-1}' \widetilde{\beta} \right] & \tau > 1 \text{ (A.V.9)} \end{cases}$$

$$\frac{\partial f_{\tau}}{\partial \rho_{\star}}(\psi_{H_0}) = 0 \quad \forall \tau \tag{A.V.10}$$

$$\frac{\partial \nu_{\tau}}{\partial \beta}(\psi_{H_0}) = \begin{cases}
-z_1 & \tau = 1 \text{ (A.V.11)} \\
-z_{\tau} & \tau > 1 \text{ (A.V.12)}
\end{cases}$$

$$\frac{\partial f_{\tau}}{\partial \beta}(\psi_{H_0}) = 0 \quad \forall \tau \tag{A.V.13}$$

Substituting (A.V.8-13) into (A.II.1), it follows that:

$$\frac{\partial l}{\partial \beta}(\psi_{H_0}) = 0$$

$$\frac{\partial l}{\partial \sigma_{\star}^2}(\psi_{H_0}) = 0$$

$$\frac{\partial l}{\partial \rho_{\star}}(\psi_{H_0}) = \frac{m+1}{m\tilde{\sigma}_{\star}^2} \sum_{\tau=2}^{\left[\frac{T}{m}\right]} \tilde{u}_{\tau} \tilde{u}_{\tau-1} \tag{A.V.14}$$

Taking plim's in the expressions fot the information matrix and under appropriated regularity conditions on the x'(t), it follows that:

$$plim \left[ \widetilde{I}_{\rho_{+}\rho_{+}} - \left[ \frac{m+1}{m} \right]^{2} \frac{1}{\widetilde{\sigma}_{+}^{2}} \sum_{\tau=1}^{\left[ \frac{T}{m} \right]} \widetilde{u}_{\tau-1}^{2} \right] = 0$$

$$plim \left[ \widetilde{I}_{\beta\beta} - \frac{1}{\widetilde{\sigma}_{+}^{2}} \sum_{\tau=1}^{\left[ \frac{T}{m} \right]} z_{\tau} z_{\tau}^{i} \right] = 0$$

$$plim \widetilde{I}_{\rho_{+}\beta} = 0$$

$$plim \widetilde{I}_{\rho_{+}\beta} = 0$$

$$plim \widetilde{I}_{\beta\sigma_{+}^{2}} = 0$$

It follows that

$$LM = \left[ \left[ \frac{m+1}{m} \right] \frac{1}{\widetilde{\sigma}_{\star}^2} \sum_{\tau=1}^{\left[ \frac{T}{m} \right]} \widetilde{u}_{\tau} \widetilde{u}_{\tau-1} \right] \left[ \left[ \frac{m+1}{m} \right]^2 \frac{1}{\widetilde{\sigma}_{\star}^2} \sum_{\tau=1}^{\left[ \frac{T}{m} \right]} \widetilde{u}_{\tau-1}^2 \right]^{-1} \left[ \left[ \frac{m+1}{m} \right] \frac{1}{\widetilde{\sigma}_{\star}^2} \sum_{\tau=1}^{\left[ \frac{T}{m} \right]} \widetilde{u}_{\tau} \widetilde{u}_{\tau-1} \right]$$

but

$$\left[\frac{1}{\widetilde{\sigma}_{\star}^{2}}\sum_{\tau=1}^{\left[\frac{\tau}{m}\right]}\widetilde{u}_{\tau-1}^{2}\right]=\frac{T}{m}$$

then

$$LM = \left[\frac{T}{m}\right]^2 r_1^{\star^2} \left[\frac{T}{m}\right]^{-1} = \left[\frac{T}{m}\right] r_1^{\star^2}$$

where

$$r_1^{\star^2} = \frac{\sum_{\tau=1}^{\left[\frac{T_1}{T_1}\right]} \widetilde{u}_{\tau} \widetilde{u}_{\tau-1}}{\sum_{\tau=1}^{\left[\frac{T_1}{T_1}\right]} \widetilde{u}_{\tau}^2}$$

and  $\widetilde{u}_{\tau} = y_{\tau} - z_{\tau}^{t} \widetilde{\beta}$  and  $\widetilde{\beta}$  is given by (3.3):

### REFERENCES

ANSLEY, C. I. & KOHN, R. (1985) Estimation, Filtering and Smoothing in State Space Models With Incompletely Specified Initial Conditions, Annals of Statistics, (to appear).

BREUSCH, T. S. (1978) Testing for Autocorrelation in Dynamic Linear Models,

Australian Economic Paper, 17, 334 - 355.

BREUSCH, T. S. & PAGÁN, A. R. (1980) The Lagrange Multiplier Test and Its Applications to Model Specification in Econometrics, Review of Economic Studies, 47, 239 - 254.

DUDOUR, J - M & DAGENAIS, M. G. (1985) Durbin - Watson Tests for Serial Correlation in Regressions With Missing Observations, Journal of Econometrics, 27, 371 - 381

DURBIN, J. & WATSON, G. S. (1950) Testing for Serial Correlation in Least Squares

Regressions I, Biometrika, 37, 409 - 426.

DURBIN, J. & WATSON, G. S. (1951) Testing for Serial Correlation in Least Squares

Regressions II, Biometrika, 38, 159 - 178.

DURBIN, J. & WATSON, G. S. (1971) Testing for Serial Correlation in Least Squares Regressions III, Biometrika, 58, 1 - 19.

DURBIN, J. (1970) Testing for Serial Correlation in Least Squares Regressions When Some of the Regressors are Lagged Dependent Variables, Econometrica, 38, 410 - 421.

ENGLE, R. & WATSON, M. (1981) A One Factor Multivariate Time Series Model of Metropolitan Wage Rates, Journal of the American Statistical Association, 76, 774 - 781.

GODFREY, L. G. (1978) Testing Against general Autoregressive and Moving Average Error Models When the Regressors Include Lagged Dependent variables. <u>Econometrica</u>, 46, 1293 - 1301.

HARVEY, A. C. (1981) Time Series Models. Deddington: Philip Allan.

PAGAN, A. (1978) A Unified Approach to Estimation and Inference for Stochastically

Varying Coefficient Regression Models, Core Discussion Paper nº 7814.

PEREIRA, P. L. V. (1985) Testing for Serial Correlation in Regression Models with Missing Observations. Paper presented at the I Econometrics and Time Series Workshop, July 1985, IMPA, Rio de Janeiro, Brasil.

PEREIRA, P. L. V. (1986) The Loss of Efficiency in Estimation of Stationary Economic Time Series Models With Missing Observations, submitted to Journal of Time Series

Analysis.

ROBINSON, P. M. (1985) Testing for Serial Correlation in Regression with Missing Observations, Journal of the Royal Statistical Society, Series B, 47, 429 - 437.

SAVIN, N. E. & WHITE, K. J. (1978) Testing for Serial Correlation With Missing

Observations, Econometrica, 46, 59 - 67.

TIAO, G. C. (1972) Asymptotic Behaviour of Time Series Aggregated, Biometrika, 59, 523 - 531.

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