

## COMPARISON OF WHITTLE TYPE PORTMANTEAU TESTS

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**ABSTRACT.** For an ARMA adequacy test, Box and Pierce (1970) proposed a portmanteau test  $T_{BP}$ . However, because the accuracy of  $T_{BP}$  by  $\chi^2$ -approximation is not good, various modifications of  $T_{BP}$  have been introduced by many authors. Taniguchi and Amano (2008) proposed an important portmanteau test  $T_{WLR}$  of natural Whittle type which is always asymptotically  $\chi^2$  distributed under the null hypothesis that ARMA model is adequate. This paper compares  $T_{WLR}$  with another famous portmanteau tests Ljung-Box's  $T_{LB}$ , Li-McLeod's  $T_{LM}$  and Monti's  $T_{MN}$  and proves its accuracy by simulation. Empirical powers of those portmanteau tests are also compared numerically.

**1. Introduction** One of the most important stages of building a model in time series is to verify the adequacy of a fitted model. In particular, sample residual autocorrelations are usually used. For ARMA adequacy test, Box and Pierce (1970) proposed a test statistic  $T_{BP}$  which is the squared sum of  $m$  sample autocorrelations of the estimated residual process of ARMA(p,q). Under the null hypothesis that the ARMA(p,q) model is adequate, it is suggested that  $T_{BP}$  is approximately distributed as  $\chi^2_{m-p-q}$ . However, Davies et al. (1977) claimed that the  $\chi^2_{m-p-q}$ -approximation is not adequate and Ljung and Box (1978) and Li-McLeod (1981) proposed test statistics  $T_{LB}$  and  $T_{LM}$  as a modification of  $T_{BP}$ . Recently Monti (1994) proposed a portmanteau test  $T_{MN}$  using the residual partial autocorrelations. Various modified versions of  $T_{BP}$  (see Li (2004)) have been proposed. Under the null hypothesis that ARMA(p,q) is adequate, these test statistics are much closer to chi-square distribution than  $T_{BP}$ .

The test statistic  $T_{BP}$  and modifications of  $T_{BP}$  are called the portmanteau test and have been widely used. Taniguchi and Amano (2008) proved that  $T_{BP}$  does not converge to  $\chi^2_{m-p-q}$  distribution for fixed  $m$  and for ARMA adequacy test, proposed a portmanteau test of natural Whittle type  $T_{WLR}$  and showed that  $T_{WLR}$  is always asymptotically chi-square distributed. This paper compares  $T_{WLR}$  with another famous portmanteau test statistics  $T_{BP}$ ,  $T_{LB}$  and  $T_{MN}$  and we observe that  $T_{WLR}$  behaves well numerically.

This paper is organized as follows. Section 2 describes the construction of  $T_{WLR}$  and its asymptotics. In Section 3, we compare the means and variances of  $T_{WLR}$  with those of other portmanteau tests  $T_{BP}$ ,  $T_{LB}$  and  $T_{MN}$  by simulation. Then the empirical significance levels and the empirical powers under contiguous alternatives are compared numerically.

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## 2. Asymptotics of $T_{WLR}$

A stationary process  $\{X_t\}$  is assumed to satisfy

$$\sum_{j=0}^p \alpha_j X_{t-j} = \sum_{j=0}^q \beta_j u_{t-j}, \quad (\alpha_0 = \beta_0 = 1, \alpha_p \neq 0, \beta_q \neq 0), \quad (2.1)$$

where  $\{u_t\}$  is an  $m$ -dependent sequence with autocovariance  $\{\theta_{2,j}\}$  ( $\theta_{2,0} \equiv 1, \theta_{2,-j} \equiv \theta_{2,j}$ ) and the innovation process of  $\{u_t\}$  is identically distributed with mean 0, variance  $\sigma_u^2$  and fourth-order cumulant  $\kappa_4$ . Let  $\alpha(z) \equiv \sum_{j=0}^p \alpha_j z^j$  and  $\beta(z) \equiv \sum_{j=0}^q \beta_j z^j$ , and they are assumed to satisfy  $\alpha(z) \neq 0$  and  $\beta(z) \neq 0$  on  $\mathbf{D} = \{z \in \mathbf{C} : |z| \leq 1\}$  and the equations  $\alpha(z) = 0$  and  $\beta(z) = 0$  have no common roots. We define  $\theta_1 = (\theta_{1,1}, \dots, \theta_{1,p+q})' \equiv (\alpha_1, \dots, \alpha_p, \beta_1, \dots, \beta_q)', \theta_2 = (\theta_{2,1}, \dots, \theta_{2,m})'$  and  $\theta = (\theta_1', \theta_2')'$ , then the spectral density of  $\{X_t\}$  is

$$f_\theta(\lambda) \equiv f_{(\theta_1, \theta_2)}(\lambda) = \frac{|\sum_{j=0}^q \beta_j e^{ij\lambda}|^2}{|\sum_{j=0}^p \alpha_j e^{ij\lambda}|^2} \cdot \frac{\sigma_u^2}{2\pi} \left\{ \sum_{j=-m}^m \theta_{2,j} e^{-ij\lambda} \right\}$$

For the construction of a portmanteau test, Let  $\vec{X}_n = (X_1, \dots, X_n)'$  be an observed stretch from (1), and write the periodogram as

$$I_n(\lambda) = \frac{1}{2\pi n} \left| \sum_{t=1}^n X_t e^{it\lambda} \right|^2, \quad \lambda \in [-\pi, \pi]. \quad (2.2)$$

By use of Whittle likelihood

$$D(f_\theta, I_n) = -\frac{1}{4\pi} \int_{-\pi}^{\pi} \{ \log f_\theta(\lambda) + \frac{I_n(\lambda)}{f_\theta(\lambda)} \} d\lambda \quad (2.3)$$

estimators for  $(\theta_1', \theta_2')$  are given by

$$\hat{\theta}_1 \equiv \arg \max_{\theta_1} D(f_{(\theta_1, 0)}, I_n), \quad (\tilde{\theta}_1, \tilde{\theta}_2) \equiv \arg \max_{(\theta_1, \theta_2)} D(f_{(\theta_1, \theta_2)}, I_n), \quad (2.4)$$

where 0 in (4) is the  $m$ -dimensional zero vector. As an adequacy test for ARMA(p,q) model, a portmanteau test of natural Whittle likelihood type

$$T_{WLR} \equiv 2n[D(f_{(\tilde{\theta}_1, \tilde{\theta}_2)}, I_n) - D(f_{(\hat{\theta}_1, 0)}, I_n)] \quad (2.5)$$

was proposed in Taniguchi and Amano (2008).

The following lemmas are due to Taniguchi and Amano (2008).

**Lemma 2.1.** Write  $F \equiv \frac{1}{4\pi} \int_{-\pi}^{\pi} \frac{\partial}{\partial \theta} \log f_\theta(\lambda) \frac{\partial}{\partial \theta'} \log f_\theta(\lambda) d\lambda = \begin{pmatrix} F_{11} & F_{12} \\ F_{21} & F_{22} \end{pmatrix}$ . Suppose that  $F$  is positive definite. If ARMA(p,q) model is adequate, then for any fixed  $m = \dim \theta_2$ , it holds that

$$T_{WLR} \rightarrow \chi_m^2, \quad \text{in distribution as } n \rightarrow \infty. \quad (2.6)$$

**Lemma 2.2.** Under  $A_G^{(n)} : \theta_2 = \frac{1}{\sqrt{n}} h$ , where  $h$  is a fixed  $m$ -dimensional vector, the following holds

$$T_{WLR} \rightarrow \chi_m^2(h' F_{22 \cdot 1} h) \quad \text{in distribution as } n \rightarrow \infty \quad (2.7)$$

where  $F_{22 \cdot 1} = F_{22} - F_{21} F_{11}^{-1} F_{12}$ , and  $\chi_m^2(h' F_{22 \cdot 1} h)$  is a noncentral chi-square random variable with  $m$  degrees of freedom and noncentrality parameter  $h' F_{22 \cdot 1} h$ .

**3. Numerical study** In this section, we give a comparison of the test statistic  $T_{WLR}$  with another portmanteau tests

$$T_{LB} = n(n+2) \sum_{k=1}^m \frac{\hat{r}_k^2}{n-k}, \quad (3.1)$$

$$T_{LM} = \frac{m(m+1)}{2n} + n \sum_{k=1}^m \hat{r}_k^2 \quad (3.2)$$

and

$$T_{MN} = n(n+2) \sum_{k=1}^m \frac{\hat{\pi}_k^2}{n-k}, \quad (3.3)$$

by simulation. Here,  $\hat{r}_k$  and  $\hat{\pi}_k$  are the  $k$ th sample autocorrelations and sample partial autocorrelations of the estimated residual process of ARMA(p,q) model, respectively. Under the null hypothesis that ARMA(p,q) is adequate, these portmanteau tests  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  are supposed to be approximated by  $\chi_{m-p-q}^2$ -distribution.

In Example 3.1, the empirical means and variances of  $T_{WLR}$  for  $m = 1$  are compared with those of  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  for  $m = 2$  under null hypothesis. In Example 3.2, we compare the significance levels of  $T_{WLR}$  for  $m = 1$  with those of  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  for  $m = 2, 20$  under null hypothesis. Then we can observe that the test statistic  $T_{WLR}$  is more accurate than  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$ . In Example 3.3, local powers of the test  $T_{WLR}$  for  $m = 1$  are compared with those of  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  for  $m = 10, 20$  under local alternative and we can see that our test  $T_{WLR}$  is more powerful than  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$ .

**Example 3.1.** Let  $\{X_t\}$  be the AR(1) process

$$X_t + \alpha X_{t-1} = u_t \quad (3.4)$$

where  $u_t$ 's are independent and identically distributed as  $N(0, 1)$ . For (4), we compare the empirical means and variances of  $T_{WLR}$  for  $m = 1$  with those of  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  for  $m = 2$ , respectively. The parameter values are chosen as  $0.85 \leq \alpha \leq 0.99$ . The empirical means and variances are calculated based on length of observations  $n = 200$  and 1000 times simulation.

In Figure 1, the empirical means of  $T_{WLR}$  for  $m = 1$  and  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  for  $m = 2$  ( $0.85 \leq \alpha \leq 0.99$ ) are plotted.

In Figure 2, the empirical variances of  $T_{WLR}$  for  $m = 1$  and  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  for  $m = 2$  ( $0.85 \leq \alpha \leq 0.99$ ) are plotted.

From Figure 1, the empirical means of  $T_{WLR}$  for  $m = 1$  are closer to 1 than those of  $T_{LB}$ ,  $T_{ML}$  and  $T_{MN}$  for  $m = 2$ . From Figure 2, the empirical variances of  $T_{WLR}$  for  $m = 1$  are closer to 2 than those of  $T_{LB}$ ,  $T_{ML}$  and  $T_{MN}$  for  $m = 2$ . Due to Lemma 2.1,  $T_{WLR}$  for  $m = 1$  is approximated by  $\chi_1^2$ -distribution and  $T_{LB}$ ,  $T_{ML}$  and  $T_{MN}$  for  $m = 2$  is supposed to be approximated by  $\chi_1^2$ -distribution. Hence Figures 1 and 2 imply  $T_{WLR}$  is more accurate than another portmanteau tests  $T_{LB}$ ,  $T_{ML}$  and  $T_{MN}$ .

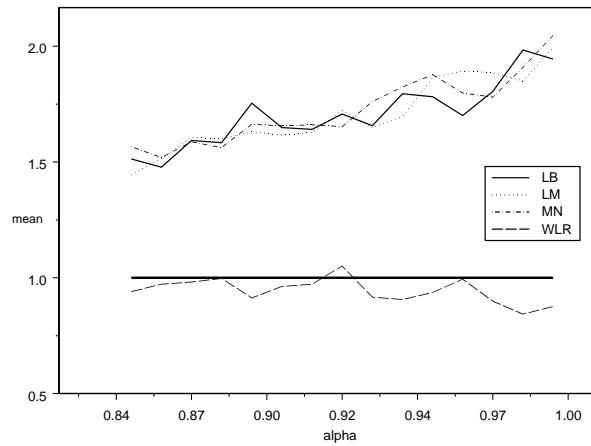


Figure 1: The means of  $T_{WLR}$ ,  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  in Example 3.1 ( $0.85 \leq \alpha \leq 0.99$ )

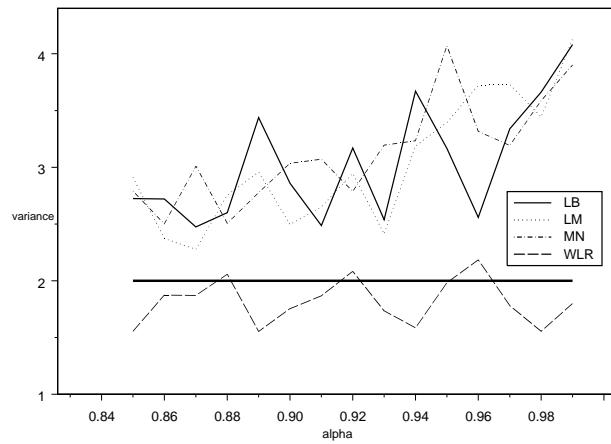


Figure 2: The variances of  $T_{WLR}$ ,  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  in Example 3.1 ( $0.85 \leq \alpha \leq 0.99$ )

**Example 3.2.** For (4), we compare the empirical significance levels of  $T_{WLR}$  for  $m = 1$  with those of  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  for  $m = 2, 20$ , respectively. The parameter values are chosen as  $0.85 \leq \alpha \leq 0.99$ . The empirical significance levels are calculated based on length of observations  $n = 200$  and 1000 times simulations.

In Figure 3, the fractions of times that  $T_{WLR}$  for  $m = 1$  and  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  for  $m = 2$  ( $0.85 \leq \alpha \leq 0.99$ ) exceed the critical values of  $\chi^2_1$ -distribution for nominal level 5% are plotted.

In Figure 4, the fractions of times that  $T_{WLR}$  for  $m = 1$  and  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  for  $m = 20$  ( $0.85 \leq \alpha \leq 0.99$ ) exceed the critical values of  $\chi^2_1$  and  $\chi^2_{19}$ -distribution for nominal level 5% are plotted.

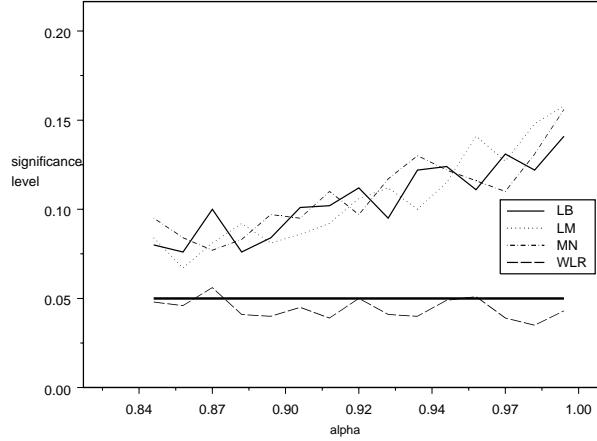


Figure 3: The significance levels with nominal size 5% of  $T_{WLR}$ ,  $T_{LB}$  ( $m = 2$ ),  $T_{LM}$  ( $m = 2$ ) and  $T_{MN}$  ( $m = 2$ ) in Example 3.2 ( $0.85 \leq \alpha \leq 0.99$ )

Due to Lemma 2.1,  $T_{WLR}$  for  $m = 1$  is approximated by  $\chi^2_1$ -distribution and  $T_{LB}$ ,  $T_{ML}$  and  $T_{MN}$  for  $m = 2$  are supposed to be approximated by  $\chi^2_{m-1}$ -distribution. From Figures 3 and 4, it is seen that  $T_{WLR}$  is closer to its asymptotic distribution than  $T_{LB}$ ,  $T_{ML}$  and  $T_{MN}$ .

**Example 3.3.** Let  $\{X_t\}$  be the AR(1) process

$$X_t + \alpha X_{t-1} = u_t \quad (3.5)$$

where  $\{u_t\}$  is the MR(1) with the mean 0, the variance 1 and the autocovariance function  $\{\frac{H}{\sqrt{n}}\}$  where  $H = \frac{3}{\sqrt{F_{22,1}}} = \frac{3}{\alpha}$ . If  $T_{WLR}$  for  $m = 1$  exceeds the 95% point of  $\chi^2_1$ , we reject the null hypothesis.  $T_{WLR}$  for  $m = 1$  is calculated with length of observations  $n = 200$ . By use of 1000 times simulation, we give the frequency that the test rejects the hypothesis. If the  $T_{LB}$  for  $m$  exceeds the 95% point of  $\chi^2_{m-1}$ , we reject the null hypothesis.  $T_{LB}$  for  $m$  is calculated with length of observations  $n = 200$ . By use of 1000 times simulations, we give

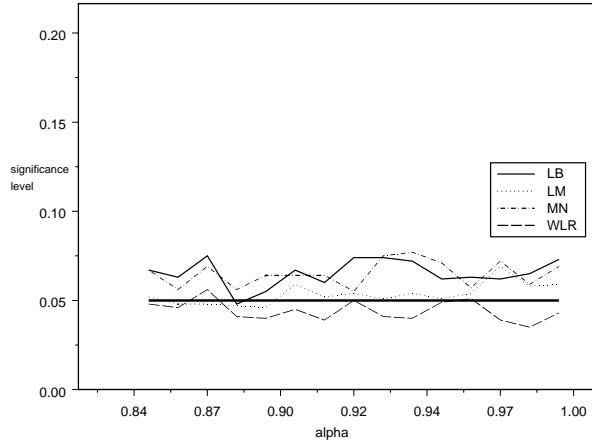


Figure 4: The significance levels with nominal size 5% of  $T_{WLR}$ ,  $T_{LB}$  ( $m = 20$ ),  $T_{LM}$  ( $m = 20$ ),  $T_{MN}$  ( $m = 20$ ) in Example 3.2 ( $0.85 \leq \alpha \leq 0.99$ )

*the frequency that the test rejects the hypothesis. Also, we give empirical powers of  $T_{LM}$  and  $T_{MN}$  for  $m$  similarly.*

In Figure 5, the empirical powers for a 5%-level test of  $T_{WLR}$  for  $m = 1$  and those of  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  for  $m = 10$  ( $0.45 \leq \alpha \leq 0.99$ ) are plotted.

In Figure 6, the empirical powers for a 5%-level test of  $T_{WLR}$  for  $m = 1$  and those of  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  for  $m = 20$  ( $0.45 \leq \alpha \leq 0.99$ ) are plotted.

From Figures 5 and 6, our test statistic  $T_{WLR}$  is more powerful than  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$ .

Simulation results imply that  $T_{WLR}$  is closer to theoretic  $\chi^2$ -distribution than another famous portmanteau tests  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  under null hypothesis that ARMA(p,q) model is adequate. It is implied that under contiguous alternative hypothesis, the ability of  $T_{WLR}$  to detect model misspecification is higher than that of another famous portmanteau tests  $T_{LB}$ ,  $T_{LM}$  and  $T_{MN}$  by simulation.

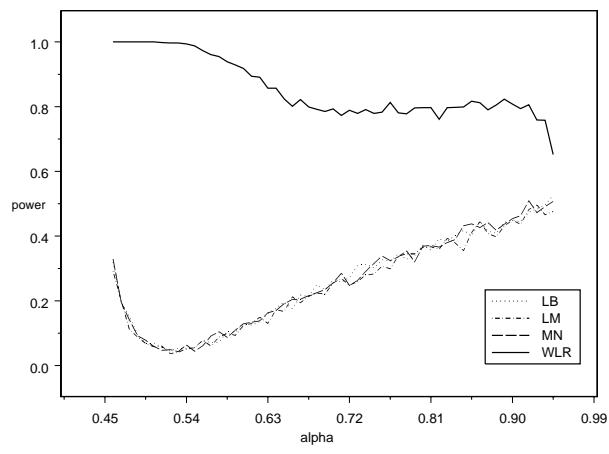


Figure 5: The empirical powers with level test 5% of  $T_{WLR}$ ,  $T_{LB}$  ( $m = 10$ ),  $T_{LM}$  ( $m = 10$ ),  $T_{MN}$  ( $m = 10$ ) in Example 3.3 ( $0.45 \leq \alpha \leq 0.99$ )

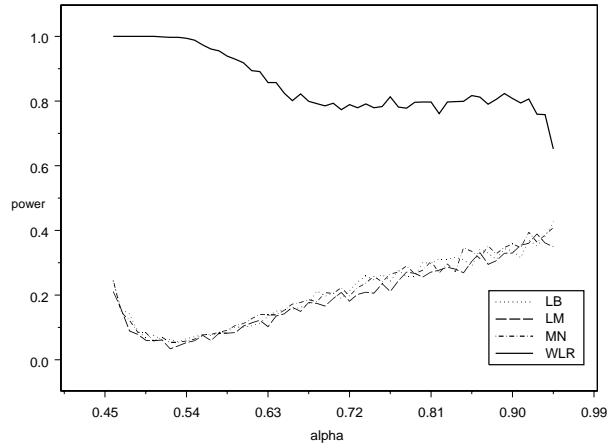


Figure 6: The empirical powers with level test 5% of  $T_{WLR}$ ,  $T_{LB}$  ( $m = 20$ ),  $T_{LM}$  ( $m = 20$ ),  $T_{MN}$  ( $m = 20$ ) in Example 3.3 ( $0.45 \leq \alpha \leq 0.99$ )

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