

ARIMA model validation *

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1. Problem

$$X_t \sim ARIMA(p, d, q)$$
$$\varphi(B) \nabla^d X_t = \varphi_0 + \Theta(B) a_t .$$

After estimation of the parameters. we expect that residuals \hat{a}_t be approximately a white noise.

The criterion of success for ARIMA model is: to reduce a time series to the white noise structure.

Let us examine how we can test whether a series a_1, \dots, a_N is a white noise.

2. Correlogram of a white noise

Let

$$a_1, \dots, a_N \sim BB(0, \sigma_a^2) . \quad (2.1)$$

Then

$$r_k(a) = \frac{\sum_{t=1}^{N-k} a_t a_{t+k}}{\sum_{t=1}^N a_t^2} \quad (2.2)$$

is an estimator of $E(a_t a_{t+k}) / E(a_t^2)$.

For N large,

$$r_k(a) \sim N\left[0, \frac{1}{N}\right] \quad (2.3)$$

$$\frac{r_k(a)}{1/\sqrt{N}} \sim N[0, 1] . \quad (2.4)$$

Furthermore, we can show that $r_k(a), k = 1, \dots, K$, where $K < N$, are independent. Hence :

$$Q(r) = \sum_{k=1}^K \left[\frac{r_k(a)}{1/\sqrt{N}} \right]^2 = N \sum_{k=1}^K r_k(a)^2 \sim \chi^2(K) .$$

We can test whether a_1, \dots, a_N constitute a white noise.

3. Correlogram of residuals

Instead of a_1, \dots, a_N , we have $\hat{a}_1, \dots, \hat{a}_N$. We wish to test

$$H_0 : \text{an ARIMA } (p, d, q) \text{ is adequate.} \quad (3.1)$$

Let us examine the autocorrelations:

$$r_k(\hat{a}), \quad k = 1, \dots, K.$$

For N large,

$$\sqrt{N} r_k(\hat{a}) \stackrel{a}{\sim} N[0, 1], \quad k = 1, \dots, K.$$

but they are not independent.

However, one can show [Box and Pierce (1970)] that

$$\hat{r} \simeq (I - D) r$$

where

$$r = \begin{pmatrix} r_1(a) \\ \vdots \\ r_K(a) \end{pmatrix}, \quad \hat{r} = \begin{pmatrix} r_1(\hat{a}) \\ \vdots \\ r_K(\hat{a}) \end{pmatrix} \quad (3.2)$$

and

$$I_K - D \text{ is an idempotent matrix of rank } K - \ell, \quad \ell = p + q. \quad (3.3)$$

Thus

$$\sqrt{N} r \stackrel{a}{\sim} N_K(0, I_K) \quad (3.4)$$

$$\sqrt{N} \hat{r} \simeq (I_K - D) \sqrt{N} r \stackrel{a}{\sim} N_K(0, I_K - D) \quad (3.5)$$

$$Q(\hat{r}) = N \sum_{k=1}^K r_k(\hat{a})^2 \sim \chi^2(K - \ell)$$

$p + q$ does not include the constant.

4. Ljung-Box statistic

For relatively short series, approximating the distribution of Q by a $\chi^2(K - \ell)$ distribution can yield very unreliable results [see Davies, Triggs and Newbold (1977)]. In particular,

$$E(Q) < E[\chi^2(K - \ell)]. \quad (4.1)$$

Ljung and Box (1978) have proposed a modification which improves the approximation. Consider first the case of white noise:

$$\begin{aligned} \text{Var} [r_k(a)] &= \frac{N-k}{N(N+2)}, \quad k = 1, 2, \dots, K, \\ \frac{r_k(a)}{\sqrt{\frac{N-k}{N(N+2)}}} &\sim N(0, 1) \end{aligned} \quad (4.2)$$

where

$$\frac{N-k}{N(N+2)} < \frac{1}{N}. \quad (4.3)$$

$$\tilde{Q}(r) = \sum_{k=1}^K \left[\frac{r_k(a)}{\sqrt{\frac{N-k}{N(N+2)}}} \right]^2 = N(N+2) \sum_{k=1}^K \frac{r_k(a)^2}{N-k}.$$

With estimated residuals, we use:

$$\tilde{Q}(\hat{r}) = N(N+2) \sum_{k=1}^K (N-k)^{-1} r_k(\hat{a})^2 \sim \chi^2(K-\ell).$$

This statistic is called the Ljung-Box statistic

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