

22-03-10

Econ 469 - Econometrics

MID-TERM EXAM

SOLUTIONS

(20) 1. (a) This is a MA(1) model, so that

$$\gamma(0) = (1 + \theta^2)\sigma^2,$$

$$\gamma(1) = -\theta\sigma^2,$$

$$\rho(1) = -\theta/(1 + \theta^2).$$

The derivative of $\rho(1)$ with respect to θ is:

$$\frac{\partial \rho(1)}{\partial \theta} = \frac{-(1 + \theta^2) + \theta(2\theta)}{(1 + \theta^2)^2} = \frac{-1 + \theta^2}{(1 + \theta^2)^2}$$

hence

$$\begin{aligned} \frac{\partial \rho(1)}{\partial \theta} &> 0 && \text{if } \theta < -1 \\ &= 0 && \text{if } \theta = -1 \\ &< 0 && \text{if } |\theta| < 1 \\ &= 0 && \text{if } \theta = 1 \\ &> 0 && \text{if } \theta > 1 \end{aligned}$$

This entails that $\rho(1)$ has a minimum at $\theta = -1$ and a maximum at $\theta = 1$

Since

$$\rho(1) = 0.5 \quad \text{if } \theta = -1,$$

$$\rho(1) = -0.5 \quad \text{if } \theta = 1,$$

$$\rho(1) \rightarrow 0 \quad \text{as } \theta \rightarrow -\infty,$$

$$\rho(1) \rightarrow 0 \quad \text{as } \theta \rightarrow +\infty,$$

it follows that

$$|\rho(1)| \leq 0.5.$$

(b) The upper bound of $|\rho(1)|$
is attained at $\theta = -1$ and $\theta = 1$
with

$$\rho(1) = -0.5 \quad \text{if } \theta = -1$$

$$\rho(1) = 0.5 \quad \text{if } \theta = +1,$$

2. We can consider the model:

$$X_t = 10 + M_t - 0.75 M_{t-1} + 0.125 M_{t-2}$$

$$\{M_t : t \in \mathbb{Z}\} \stackrel{i.i.d.}{\sim} N(0, 1)$$

(a) Yes. This process is an $MA(2)$ and consequently, it is stationary.

(b) Yes.

We have:

$$X_t = 10 + \theta(B) M_t$$

where

$$\begin{aligned} \theta(B) &= 1 - 0.75B + 0.125B^2 \\ &= (1 - 0.5B)(1 - 0.25B) \end{aligned}$$

The roots of the equation

$$\theta(z) = 0$$

$$\text{are } z_1 = 1/0.5 = 2.0, \quad z_2 = 1/0.25 = 4$$

Since $|z_1| > 1$ and $|z_2| > 1$, both the roots are outside the unit circle.

$$2.(c) i) E(X_t) = 10$$

$$ii) \gamma(0) = \text{Var}(X_t) = [1 + (0.75)^2 + (0.125)^2] = 1.578125$$

$$\gamma(1) = E[(X_t - 10)(X_{t-1} - 10)]$$

$$= E[(M_t - 0.75M_{t-1} + 0.125M_{t-2})(M_{t-1} - 0.75M_{t-2} + 0.125M_{t-3})]$$

$$= 0.75 - (0.125)(0.75) = -0.84375$$

$$\gamma(2) = E[(X_t - 10)(X_{t-2} - 10)]$$

$$= E[(M_t - 0.75M_{t-1} + 0.125M_{t-2})(M_{t-2} - 0.75M_{t-3} + 0.125M_{t-4})]$$

$$= 0.125$$

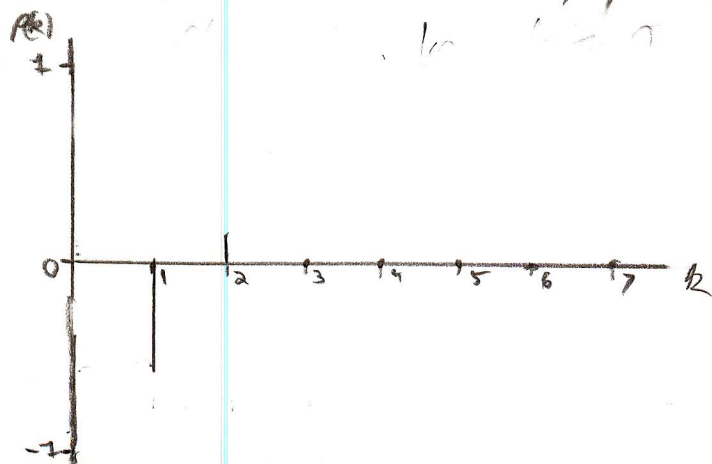
$$\gamma(k) = 0, \text{ for } k \geq 3$$

Thus: $\gamma(0) = 1.578125$, $\gamma(1) = -0.84375$, $\gamma(2) = 0.125$
 $\gamma(k) = 0$, for $k \geq 3$

$$iii) \rho(k) = \gamma(k) / \gamma(0)$$

$$\rho(1) = -0.53465, \rho(2) = 0.07921, \rho(k) = 0, \text{ for } k \geq 3$$

(d)



2. (e) Since the model is already in MA form, the coefficients of the MA representation are the same:

$$X_t = 10 + \sum_{k=0}^{\infty} \psi_k u_{t-k}$$

where

$$\psi_0 = 1, \quad \psi_1 = -0.75, \quad \psi_2 = 0.125,$$

$$\psi_k = 0, \quad \text{for } k \geq 3.$$

(f) The first partial autocorrelation is

$$\alpha_{11} = \rho(1) = -0.53465$$

The second partial autocorrelation is obtained by solving the equations:

$$\rho(1) = \rho(0)\alpha_{21} + \rho(1)\alpha_{22} = \alpha_{21} + \rho(1)\alpha_{22}$$

$$\rho(2) = \rho(1)\alpha_{21} + \rho(0)\alpha_{22} = \alpha_{21}\rho(1) + \alpha_{22}$$

or

$$\begin{bmatrix} \rho(1) \\ \rho(2) \end{bmatrix} = \begin{bmatrix} 1 & \rho(1) \\ \rho(1) & 1 \end{bmatrix} \begin{bmatrix} \alpha_{21} \\ \alpha_{22} \end{bmatrix}$$

hence

$$\alpha_{22} = \frac{\begin{vmatrix} 1 & \rho(1) \\ \rho(1) & \rho(2) \end{vmatrix}}{\begin{vmatrix} 1 & \rho(1) \\ \rho(1) & 1 \end{vmatrix}} = \frac{\rho(2) - \rho(1)^2}{1 - \rho(1)^2}$$

$$= \frac{-0.20664}{0.71415} = -0.28935$$

2.(g) $X_8 = 11$ is the only information available to make the forecast.

i) We need to use the formula

$$P_L(X_{t+1} | X_t) = \beta_0 + \beta_1 X_t$$

$$\beta_1 = \frac{\text{Cov}(X_{t+1}, X_t)}{V(X_t)} = \frac{\gamma(1)}{\gamma(0)} = \rho(1) = -0.53465$$

$$\beta_0 = 10 - \beta_1(10) = 10(1 - \beta_1) = 15.3465$$

$$P_L(X_9 | X_8) = 15.3465 + (-0.53465)(11)$$

$$= 10 - 0.53465 = 9.46535$$

1.1) Since the process is Gaussian,

$$E(X_{t+1} | X_t) = P_L(X_{t+1} | X_t)$$

hence

$$E(X_9 | X_8) = 9.46535$$

(h) i) X_{15} is uncorrelated with all X_t , $t \leq 10$,
therefore

$$P_L(X_{15} | X_t, t \leq 10) = E(X_{15}) = 10.$$

ii) By the Gaussian assumption,

X_{15} is independent of $\{X_t, t \leq 10\}$.

therefore

$$E(X_{15} | X_t, t \leq 10) = E(X_{15}) = 10.$$

3. (a) i) The sample autocorrelations are

$$r_k = \sum_{t=1}^{T-k} (x_{t+k} - \bar{x})(x_t - \bar{x}) / \sum_{t=1}^T (x_t - \bar{x})^2$$

where $\bar{x} = \frac{1}{T} \sum_{t=1}^T x_t$

ii) The partial autocorrelations are obtained by solving the equation

$$r_j = \sum_{k=1}^j a_{kj} r_{j-k}, \quad j=1, 2, \dots, k$$

to obtain a_{jk} ($k=1, 2, \dots$)

(b) i) For any $m \geq 1$,

$$\sqrt{T}[(r_1 - \rho_1), \dots, (r_m - \rho_m)]' \xrightarrow{T \rightarrow \infty} N[0, I_m]$$

when $\rho_1 = \dots = \rho_m = 0$,
or that

$$\sqrt{T}[r_1, \dots, r_m]' \xrightarrow{T \rightarrow \infty} N[0, I_m]$$

Similarly

$$\sqrt{T}(a_{11}, \dots, a_{mm})' \xrightarrow{T \rightarrow \infty} N[0, I_m]$$

(b) ii) For any $m \geq 1$

$$\sqrt{T}[(\lambda_1 - \rho_1), \dots, (\lambda_m - \rho_m)] \rightarrow N[0, \Sigma_m]$$

where the elements of Σ_m are given by the Bartlett formula.

$$\text{Further, } \Sigma_m = [\sigma_{ij}]_{i,j=1,\dots,m}$$

$$\text{where } \sigma_{ii} = 1 + 2 \sum_{j=1}^q \rho_j^2 \quad \text{for } i \geq q+1$$

The sample partial autocorrelations are asymptotically normal.

(c) In question 2, we have an MA(2) process so that $\rho_1 \neq 0, \rho_2 \neq 0$ and

$$\rho_k = 0, \quad \text{for } k \geq 3$$

The order of the process can be identified by looking at the correlogram. We check whether

$$t_k = \left| \frac{\sqrt{T} \hat{\lambda}_k}{\hat{\sigma}_k} \right| > \kappa(2/2), \quad k=1, 2, \dots$$

$$\hat{\sigma}_k^2 = 1 + 2 \sum_{j=1}^{k-1} \hat{\rho}_j^2$$

where $\alpha(2/2)$ is the normal critical value for a test with level α .

For an $MA(2)$ process, we expect that the first two of these tests be non-significant, while the other ones (for $k \geq 3$) will not be significant.