

# STAT4005 Solution to Assignment 3

## Question 1

Time series, ACF, and PACF of  $Y$  are plotted in Figure 1.

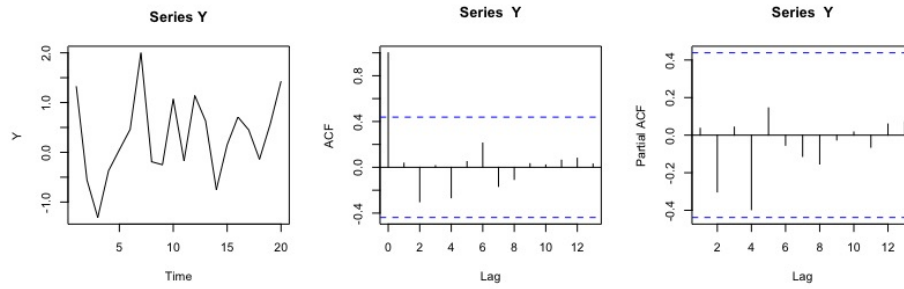


Figure 1: Values, ACF, and PACF of  $Y$

## Question 2

First and second moments of the  $MA(1)$  model:  $Y_t = Z_t + \theta Z_{t-1}$  are

$$\begin{cases} \gamma(0) &= (1 + \theta^2)\sigma^2, \\ \gamma(1) &= \theta\sigma^2 \end{cases}$$

Hence, moment estimates of  $\theta$  and  $\sigma^2$  satisfy

$$\begin{cases} C(0) &= (1 + \hat{\theta}^2)\hat{\sigma}^2, \\ C(1) &= \hat{\theta}\hat{\sigma}^2 \end{cases}$$

where  $C(k) = \frac{1}{n} \sum_{t=1}^{n-k} (Y_t - \bar{Y})(Y_{t+k} - \bar{Y})$  is the  $k$ -th sample moment,  $k \geq 0$ . Hence,

$$\frac{\hat{\theta}}{1 + \hat{\theta}^2} = \frac{C(1)}{C(0)} = r(1)$$

$$r(1)\hat{\theta}^2 - \hat{\theta} + r(1) = 0$$

Hence,

$$\hat{\theta}_1 = 0.03629, \quad \hat{\theta}_2 = 27.56$$

$\hat{\theta}_2$  is rejected since  $|\hat{\theta}_2| > 1$  implies non-invertibility of the model. That is,  $\hat{\theta} = 0.03629$ , and

$$\hat{\sigma}^2 = \frac{C(1)}{\hat{\theta}} = 0.6415$$

### Question 3

Let  $\mathbf{Y}_t = (Y_t, Y_{t+1}, \dots, Y_{17+t})'$ ,  $t = 1, 2, 3$ , then least squares estimate of  $\phi = (\phi_2, \phi_1)$  are found from

$$\mathbf{Y}_3 = \mathbf{X}\phi + \epsilon$$

where  $\mathbf{X} = (\mathbf{Y}_2, \mathbf{Y}_1)$ . Hence,

$$\hat{\phi} = \begin{pmatrix} \hat{\phi}_2 \\ \hat{\phi}_1 \end{pmatrix} = (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{Y}_3 = \begin{pmatrix} 0.2354 \\ -0.2230 \end{pmatrix}$$

$$\hat{\sigma}^2 = \frac{1}{n-2}(\mathbf{Y}_3 - \mathbf{X}\hat{\phi})'(\mathbf{Y}_3 - \mathbf{X}\hat{\phi}) = 0.6432$$

Asymptotic distribution of  $\hat{\phi}$  is  $\sqrt{n}(\hat{\phi} - \phi) \sim \mathbf{N}(\mathbf{0}, \sigma^2\mathbf{\Gamma}_2^{-1})$ , where

$$\mathbf{\Gamma}_2 = \begin{bmatrix} \gamma(0) & \gamma(1) \\ \gamma(1) & \gamma(0) \end{bmatrix}$$

Hence,

$$\widehat{var}(\hat{\phi}) = \frac{\hat{\sigma}^2}{n} \begin{bmatrix} C(0) & C(1) \\ C(1) & C(0) \end{bmatrix}^{-1} = \frac{\hat{\sigma}^2}{n} \begin{bmatrix} 1.559 & -0.05649 \\ -0.05649 & 1.559 \end{bmatrix} = \begin{bmatrix} 0.05014 & -0.001817 \\ -0.001817 & 0.05014 \end{bmatrix}$$

95% confidence interval for  $\hat{\phi}_k$  is  $\hat{\phi}_k \pm 2\sqrt{\widehat{var}(\hat{\phi}_k)}$ . Hence, 95% C.I. for  $\hat{\phi}_2$  is  $[-0.2125, 0.6832]$ , that for  $\hat{\phi}_1$  is  $[-0.6709, 0.2248]$ .

### Question 4

Yule-Walker equations for the  $AR(2)$  model:  $Y_t = \phi_1 Y_{t-1} + \phi_2 Y_{t-2} + Z_t$ , are

$$\gamma(1) = \phi_1 \gamma(0) + \phi_2 \gamma(1)$$

$$\gamma(2) = \phi_1 \gamma(1) + \phi_2 \gamma(0)$$

Hence, Yule-Walker estimates of  $\phi = (\phi_1, \phi_2)'$  is

$$\begin{pmatrix} \hat{\phi}_1 \\ \hat{\phi}_2 \end{pmatrix} = \begin{pmatrix} C(0) & C(1) \\ C(1) & C(0) \end{pmatrix}^{-1} \begin{pmatrix} C(1) \\ C(2) \end{pmatrix} = \begin{pmatrix} 0.04718 \\ -0.3020 \end{pmatrix}$$

## Question 5

For the  $ARMA(1, 1)$  model of  $Y_t - \phi Y_{t-1} = Z_t + \theta Z_{t-1}$ , conditional on  $Z_0 = 0, Y_0 = 0$ , we have

$$\begin{aligned} Z_1 &= Y_1 \\ Z_t &= Y_t - \phi Y_t - \theta Z_{t-1}, \quad t \geq 2 \end{aligned}$$

Let  $S_*(\phi, \theta) = \sum_{t=1}^{20} Z_t^2$ , then conditional least squares estimate of  $(\phi, \theta, \sigma^2)$  is given by

$$\begin{aligned} (\hat{\phi}, \hat{\theta}) &= \arg \min_{\phi, \theta} S_*(\phi, \theta) \\ \hat{\sigma}^2 &= \frac{1}{n-1} S_*(\hat{\phi}, \hat{\theta}) \end{aligned}$$

Hence,  $\hat{\phi} = -0.5109, \hat{\theta} = 0.8440, \hat{\sigma}^2 = 0.6810$ .

## Question 6

Assume  $Z_t \stackrel{i.i.d.}{\sim} N(0, \sigma^2)$ , maximum likelihood estimate of  $(\phi, \theta, \sigma^2)$  is given by

$$(\hat{\phi}, \hat{\theta}, \hat{\sigma}^2) = \arg \min_{\phi, \theta, \sigma^2} l(Y_1, Y_2, \dots, Y_{20}) = \arg \min_{\phi, \theta, \sigma^2} \frac{1}{(2\pi)^{\frac{20}{2}} |\Sigma|^{\frac{1}{2}}} \exp\left(-\frac{1}{2} \mathbf{Y}' \Sigma^{-1} \mathbf{Y}\right)$$

where  $\mathbf{Y} = (Y_1, Y_2, \dots, Y_{20})'$ , and

$$\Sigma = (\gamma(|i-j|)) = \begin{pmatrix} \gamma(0) & \gamma(1) & \cdots & \gamma(n-1) \\ \gamma(1) & \gamma(0) & \cdots & \gamma(n-2) \\ & & \ddots & \\ \gamma(n-1) & \gamma(n-2) & \cdots & \gamma(0) \end{pmatrix}$$

Hence,  $\hat{\phi} = -0.3883, \hat{\theta} = 0.9999, \hat{\sigma}^2 = 0.5254$ , and maximized likelihood =  $-23.0844$ .

## Question 7

$$FPE = \hat{\sigma}^2 \frac{n+p}{n-p}$$

where  $\hat{\sigma}^2$  is MLE of  $\sigma^2$ . According to Table 1,  $AR(1)$  has the smallest FPE.

## Question 8

$$AICC = -2 \log\text{-likelihood} + \frac{2(p+q+1)n}{n-p-q-2}$$

According to Table 2,  $MA(2)$  has the smallest AICC.

Table 1: FPE for  $AR(p)$ ,  $p = 1, 2, 3, 4, 5$ .

model	FPE
AR(1)	0.8042
AR(2)	0.8422
AR(3)	0.8882
AR(4)	0.9319
AR(5)	0.8477

Table 2: AICC for  $MA(q)$ ,  $q = 1, 2, 3, 4, 5$ .

model	AICC
MA(1)	53.3839
MA(2)	53.2022
MA(3)	56.2802
MA(4)	59.4228
MA(5)	63.5976

## Question 9

Fit the model by default method in R (maximum likelihood conditional on starting values), then residuals are  $\hat{Z}_t = (0.9836, -1.0610, -1.4551, -0.4459, -0.2031, 0.1643, 1.6398, -0.8093, -0.4393, 0.8158, -0.6446, 0.9218, 0.1368, -1.1054, 0.0127, 0.3764, 0.0525, -0.4806, 0.3230, 1.0419)$ . Now conduct the Ljung-Box test.

$H_0$  : the series of residuals exhibits no autocorrelation up to lag 10.

$H_1$  : some autocorrelation coefficient,  $\rho(k)$ ,  $k = 1, 2, \dots, 10$ , are nonzero.

The test statistic is  $Q(10) = n(n+1) \sum_{j=1}^{10} \frac{\hat{\rho}^2(j)}{n-j} = 11.7303$ . Under  $H_0$ ,  $Q(10) \rightarrow \chi^2(9)$ . Since  $Q(10) < \chi_{0.95}^2(9) = 16.9190$ , the null hypothesis of no autocorrelation up to lag 10 cannot be rejected.

## Question 10

Consider the family of models,  $ARMA(p, q)$ :

$$Y_t - \phi_1 Y_{t-1} - \dots - \phi_p Y_{t-p} = Z_t + \theta_1 Z_{t-1} + \dots + \theta_q Z_{t-q}$$

where  $p \in \{0, 1, 2, 3, 4, 5\}$ ,  $q \in \{0, 1, 2, 3, 4, 5\}$ ,  $p = q = 0$  excluded.

By AIC/AICC, the best model is  $ARMA(0, 2)$ . By BIC, the best model is  $ARMA(1, 4)$ . By FPE, the best model is  $ARMA(2, 4)$ . Since the data set is small, by consideration of parsimony,  $MA(2)$  can be chosen. Analysis of residuals shows that  $MA(2)$  provides a good fit to the data.