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Due on Tuesday, February 4th at 11:59 PM

Question:	1	2	3	4	5	Total
Points:	10	10	10	10	10	50
Score:						

- (10 points) If  $X$  and  $Y$  are dependent but  $\text{Var}(X) = \text{Var}(Y)$ , find  $\text{Cov}(X + Y, X - Y)$
- (10 points) Let  $Y_1, Y_2, \dots, Y_n$  be iid with  $E(Y_i) = \mu$  and  $\text{Var}(Y_i) = \sigma^2$ . Recall that  $\bar{Y} = \frac{1}{n} \sum_{i=1}^n Y_i$ . Find  $E[(\bar{Y} - E(\bar{Y}))^2]$
- (10 points) Suppose  $Y_t = X$  for all  $t$  where  $E(X) = \mu$  and  $\text{Var}(X) = \sigma^2$ .
  - Show that  $\{Y_t\}$  is stationary.
  - Find the autocovariance function  $\gamma_k$  for  $\{Y_t\}$ .
- (10 points) Recall: To show that a process  $\{Y_t\}$  is not stationary, try to show that  $E(Y_t)$  depends on  $t$ . If this fails, try to show that  $\text{Var}(Y_t)$  depends on  $t$ . If this fails, show that  $\gamma_k = \text{Cov}(Y_t, Y_{t-k})$  depends on  $t$ .
  - Let  $Y_t = \sum_{i=1}^t e_i$  where the  $e_i$  are iid with  $E(e_i) = \mu > 0$  and  $\text{Var}(e_i) = \sigma^2$ . Show that  $\{Y_t\}$  is not stationary.
  - Let  $Y_t = \sum_{i=1}^t e_i$  where the  $e_i$  are iid with  $E(e_i) = 0$  and  $\text{Var}(e_i) = \sigma^2$ . Show that  $\{Y_t\}$  is not stationary.
- (10 points) Consider the  $MA(3)$  model with  $\theta_1 = 0.8$ ,  $\theta_2 = 0.6$ , and  $\theta_3 = 0.4$ .
  - Find the theoretical ACF.
  - Generate  $n = 150$  observations of this  $MA(3)$  time series. Then, plot the sample autocorrelation function (sample ACF).

$$(1) \quad \text{COV}(X+Y, X-Y) = \text{COV}(X, X) - \text{COV}(X, Y) + \text{COV}(Y, X) - \text{COV}(Y, Y)$$

Given  $\boxed{\begin{aligned} \text{COV}(X, X) &= \text{Var}(X) \\ \text{COV}(Y, Y) &= \text{Var}(Y) \end{aligned}}$

$$= \text{Var}(X) - \cancel{\text{COV}(X, Y)} + \cancel{\text{COV}(X, Y)} - \text{Var}(Y)$$

$$= \text{Var}(X) - \text{Var}(Y)$$

Given  $\boxed{\text{Var}(X) = \text{Var}(Y)}$   $\therefore$

therefore

$$\boxed{\text{COV}(X+Y, X-Y) = 0}$$

$$(2) \quad E[(\bar{Y} - E(\bar{Y}))^2] = \text{Var}(\bar{Y})$$

Given  $\boxed{Y_1, Y_2, \dots, Y_n \text{ is iid with variance } \sigma^2}$

therefore  $\text{Var}(\bar{Y}) = \frac{\sigma^2}{n}$

therefore

$$\boxed{E[(\bar{Y} - E(\bar{Y}))^2] = \frac{\sigma^2}{n}}$$

3A

Given  $\boxed{\begin{aligned} Y_t &= X \text{ for all } t \\ E(X) &= \mu \\ \text{Var}(X) &= \sigma^2 \end{aligned}}$

therefore  $E(X) = E(Y_t) = \mu \leftarrow \text{constant mean}$

$\text{Var}(X) = \text{Var}(Y_t) = \sigma^2 \leftarrow \text{constant variance}$

$\text{COV}(X, X) = \text{COV}(Y_t, Y_{t-k}) = \sigma^2 \leftarrow \text{constant covariance for all } t, k$

all conditions  
for  
Stationarity  
are satisfied

3B

$$\gamma_k = \text{COV}(Y_t, Y_{t-k})$$

Given  $\boxed{Y_t = X \text{ for all } t}$

$$\boxed{\text{then } \gamma_k = \text{COV}(X, X) = \sigma^2}$$

(4A)

$$E[Y_t] = E\left[\sum_{i=1}^t e_t\right] = \sum_{i=1}^t E[e_t] = \mu t \quad \begin{array}{l} \text{depends on } t \\ \text{so not stationary} \end{array}$$

(4B)

$$E[Y_t] = E\left[\sum_{i=1}^t e_t\right] = \sum_{i=1}^t E[e_t] = 0 \quad \begin{array}{l} \text{constant so} \\ \text{check variance} \end{array}$$

$$\text{Var}(Y_t) = \text{Var}\left(\sum_{i=1}^t e_t\right) = \sum_{i=1}^t \text{Var}(e_t) = \sigma^2 t \quad \begin{array}{l} \text{depends on } t \\ \text{so not stationary} \end{array}$$

(5A)

$$Y_t = e_t + \theta_1 e_{t-1} + \theta_2 e_{t-2} + \theta_3 e_{t-3}$$

$e_t$  is iid with  $\mu_{e_t}=0$  and  $\text{variance} = \sigma^2$

$$\rho_k = \frac{\gamma_k}{\gamma_0} \quad \gamma_0 = \text{Var}(Y_t)$$

$$\gamma_0 = \sigma^2 (1 + 0.8^2 + 0.6^2 + 0.4^2) = 2.16\sigma^2$$

$$\gamma_1 = \sigma^2 (\theta_1 + \theta_1 \theta_2 + \theta_2 \theta_3) = \sigma^2 (0.8 + (0.8 \cdot 0.6) + (0.6 \cdot 0.4)) = 1.52\sigma^2$$

$$\gamma_2 = \sigma^2 (\theta_2 + \theta_1 \theta_3) = \sigma^2 (0.6 + (0.8 \cdot 0.4)) = 0.92\sigma^2$$

$$\gamma_3 = \sigma^2 \theta_3 = 0.4\sigma^2$$

$$\rho_0 = 1$$

$$\rho_1 = \frac{1.52\sigma^2}{2.16\sigma^2} \approx .704$$

$$\rho_2 = \frac{0.92\sigma^2}{2.16\sigma^2} \approx .426$$

$$\rho_3 = \frac{0.4\sigma^2}{2.16\sigma^2} \approx .185$$

$$\rho_k = 0 \text{ for all } k > 3$$

# HW 2

Farooq Mahmud

## Problem 5b

```
set.seed(42)
theta <- c(0.8, 0.6, 0.4)

ma3_series <- arima.sim(n = 150, model = list(ma = theta))
acf(ma3_series)
```

Series ma3\_series

