

# STAD57: Time Series Analysis

## Problem Set 3 Solutions

### 1. Exercise 3.1 from the textbook.

**SOL:**

We have  $X_t = W_t + \theta W_{t-1}$ , so that

$$\begin{aligned}\gamma_X(h) &= \text{Cov}(X_{t+h}, X_t) = \text{Cov}(W_{t+h} + \theta W_{t+h-1}, W_t + \theta W_{t-1}) = \\ &= \text{Cov}(W_{t+h}, W_t + \theta W_{t-1}) + \theta \text{Cov}(W_{t+h-1}, W_t + \theta W_{t-1}) = \\ &= \text{Cov}(W_{t+h}, W_t) + \theta \text{Cov}(W_{t+h}, W_{t-1}) + \theta \text{Cov}(W_{t+h-1}, W_t) + \theta^2 \text{Cov}(W_{t+h-1}, W_{t-1}) = \\ &= \begin{cases} \sigma_W^2(1+\theta^2), & h=0 \\ \sigma_W^2\theta, & h=1 \\ 0, & h \geq 2 \end{cases} \Rightarrow \rho_X(h) = \begin{cases} 1, & h=0 \\ \theta/(1+\theta^2), & h=1 \\ 0, & h \geq 2 \end{cases}\end{aligned}$$

To find the maximum/minimum of  $\rho_X(1)$  we differentiate it w.r.t. to  $\theta$  and set to 0:

$$\frac{d}{d\theta} \rho_X(h) = \frac{d}{d\theta} \left( \frac{\theta}{1+\theta^2} \right) = \frac{(1+\theta^2) - \theta(2\theta)}{(1+\theta^2)^2} = \frac{1-\theta^2}{(1+\theta^2)^2} = 0 \Rightarrow \theta = \pm 1$$

Substituting  $\theta = \pm 1$  into  $\rho_X(1)$  we get that the maximum/minimum values are:

$$\rho_X(1) = \frac{\theta}{1+\theta^2} = \frac{\pm 1}{1+(\pm 1)^2} = \pm \frac{1}{2}$$

(Note: to be technically correct we also have to check that the 2<sup>nd</sup> derivatives are negative/positive at the maximum/minimum, but you can check this is the case).

### 2. Exercise 3.2 from textbook.

**SOL:**

$$\text{We have } \begin{cases} X_1 = W_1 \\ X_t = \phi X_{t-1} + W_t, t \geq 2 \end{cases} \Rightarrow X_t = \phi X_{t-1} + W_t = \phi(\phi X_{t-2} + W_{t-1}) + W_t =$$

$$= \phi^2(\phi X_{t-3} + W_{t-2}) + \phi W_{t-1} + W_t = \dots = \phi^{t-1} X_1 + \phi^{t-2} W_2 + \dots + \phi W_{t-1} + W_t \Rightarrow$$

$$\Rightarrow X_t = \phi^{t-1} W_1 + \phi^{t-2} W_2 + \dots + \phi W_{t-1} + W_t = \sum_{j=0}^{t-1} \phi^j W_{t-j}$$

$$\text{a) } \mathbb{E}[X_t] = \mathbb{E}\left[\sum_{j=0}^{t-1} \phi^j W_{t-j}\right] = \sum_{j=0}^{t-1} \phi^j \underbrace{\mathbb{E}[W_{t-j}]}_{=0} = 0$$

$$\mathbb{V}[X_t] = \mathbb{V}\left[\sum_{j=0}^{t-1} \phi^j W_{t-j}\right] = \sum_{j=0}^{t-1} \phi^{2j} \underbrace{\mathbb{V}[W_{t-j}]}_{=\sigma_W^2} = \sigma_W^2 \sum_{j=0}^{t-1} \phi^{2j} = \sigma_W^2 \frac{1-\phi^{2t}}{1-\phi^2}.$$

Since the variance depends on  $t$ , the series is not stationary.

$$\text{b) } \text{Cor}(X_t, X_{t-h}) = \text{Cor}\left(\sum_{j=0}^{t-1} \phi^j W_{t-j}, \sum_{k=0}^{t-1} \phi^k W_{t-h-k}\right) = \sum_{j=0}^{t-1} \sum_{k=0}^{t-h-1} \phi^j \phi^k \underbrace{\text{Cov}(W_{t-j}, W_{t-h-k})}_{=\begin{cases} \sigma_w^2, & j=h+k \\ 0, & j \neq h+k \end{cases}} =$$

$$= \sigma_w^2 \sum_{k=0}^{t-h-1} \phi^{k+h} \phi^k = \phi^h \left( \sigma_w^2 \sum_{k=0}^{t-h-1} \phi^{2k} \right) = \phi^h \left( \sigma_w^2 \frac{1-\phi^{2(t-h)}}{1-\phi^2} \right) = \phi^h \mathbb{V}[X_{t-h}] \Rightarrow$$

$$\Rightarrow \text{Cor}(X_t, X_{t-h}) = \frac{\text{Cov}(X_t, X_{t-h})}{\sqrt{\mathbb{V}[X_t] \mathbb{V}[X_{t-h}]} } = \frac{\phi^h \mathbb{V}[X_{t-h}]}{\sqrt{\mathbb{V}[X_t] \mathbb{V}[X_{t-h}]} } = \phi^h \sqrt{\frac{\mathbb{V}[X_{t-h}]}{\mathbb{V}[X_t]}}$$

c) Simply take limits of the above expressions as  $t \rightarrow \infty$ . We have:

$$\lim_{t \rightarrow \infty} \mathbb{V}[X_t] = \lim_{t \rightarrow \infty} \left( \sigma_w^2 \frac{1-\phi^{2t}}{1-\phi^2} \right) = \sigma_w^2 \frac{1-\lim_{t \rightarrow \infty}(\phi^{2t})}{1-\phi^2} = \sigma_w^2 \frac{1}{1-\phi^2} \quad (\text{since } |\phi| < 1)$$

$$\lim_{t \rightarrow \infty} \text{Cor}(X_t, X_{t-h}) = \lim_{t \rightarrow \infty} \left( \sigma_w^2 \phi^h \frac{1-\phi^{2(t-h)}}{1-\phi^2} \right) = \sigma_w^2 \phi^h \frac{1-\lim_{t \rightarrow \infty}(\phi^{2(t-h)})}{1-\phi^2} = \sigma_w^2 \frac{\phi^h}{1-\phi^2}.$$

d) To generate  $n$  values from the Gaussian AR(1) series, you can iteratively generate  $N \gg n$  values starting from  $X_1 = W_1 = \sigma_w Z_1 \sim N(0, \sigma_w^2)$  and using  $X_t = \phi X_{t-1} + W_t = \phi X_{t-1} + \sigma_w Z_t$ ,  $\forall t \geq 2$ , where  $\sigma_w Z_t = W_t \sim N(0, \sigma_w^2)$ , but only keep the last  $n$  values.

e) If you start with  $X_t = \frac{W_1}{\sqrt{1+\phi^2}} \sim N\left(0, \frac{\sigma_w^2}{1-\phi^2}\right)$ , the resulting series will be stationary. We have

$X_t = \phi^{t-1} W_1 + \phi^{t-2} W_2 + \dots + \phi W_{t-1} + W_t = \phi^{t-1} W_1 + \sum_{j=0}^{t-2} \phi^j W_{t-j}$ . The mean will trivially be zero again, but the variance and covariances will be given by:

$$\begin{aligned} \mathbb{V}[X_t] &= \mathbb{V}\left[\phi^{t-1} X_1 + \sum_{j=0}^{t-2} \phi^j W_{t-j}\right] = \phi^{t-1} \mathbb{V}[X_1] + \sum_{j=0}^{t-2} \phi^j \mathbb{V}[W_{t-j}] = \\ &= \sigma_w^2 \frac{\phi^{t-1}}{1-\phi^2} + \sigma_w^2 \sum_{j=0}^{t-2} \phi^j = \sigma_w^2 \left( \frac{\phi^{t-1}}{1-\phi^2} + \frac{1-\phi^{t-1}}{1-\phi^2} \right) = \frac{\sigma_w^2}{1-\phi^2} \quad (\text{indep. of } t) \end{aligned}$$

$$\begin{aligned} \text{Cor}(X_t, X_{t-h}) &= \text{Cor}\left(\phi^{t-1} X_1 + \sum_{j=0}^{t-2} \phi^j W_{t-j}, \phi^{t-h-1} X_1 + \sum_{k=0}^{t-h-2} \phi^k W_{t-h-k}\right) = \\ &= \phi^{t-1} \phi^{t-h-1} \underbrace{\text{Cov}(X_1, X_1)}_{=\mathbb{V}[X_1]} + \sum_{j=0}^{t-2} \sum_{k=0}^{t-h-2} \phi^j \phi^k \underbrace{\text{Cov}(W_{t-j}, W_{t-h-k})}_{=\begin{cases} \sigma_w^2, & j=h+k \\ 0, & j \neq h+k \end{cases}} \Rightarrow \end{aligned}$$

$$\begin{aligned} \text{Cor}(X_t, X_{t-h}) &= \sigma_w^2 \phi^h \frac{\phi^{2(t-h-1)}}{1-\phi^2} + \sigma_w^2 \sum_{k=0}^{t-h-2} \phi^{k+h} \phi^k = \sigma_w^2 \phi^h \left( \frac{\phi^{2(t-h-1)}}{1-\phi^2} + \sum_{k=0}^{t-h-2} \phi^{2k} \right) = \\ &= \sigma_w^2 \phi^h \left( \frac{\phi^{2(t-h-1)}}{1-\phi^2} + \frac{1-\phi^{2(t-h-1)}}{1-\phi^2} \right) = \sigma_w^2 \frac{\phi^h}{1-\phi^2} \quad (\text{indep. of } t) \end{aligned}$$

$$\Rightarrow \text{Cor}(X_t, X_{t-h}) = \frac{\text{Cov}(X_t, X_{t-h})}{\sqrt{\mathbb{V}[X_t] \mathbb{V}[X_{t-h}]} } = \frac{\sigma_w^2 \frac{\phi^h}{1-\phi^2}}{\sigma_w^2 \frac{1}{1-\phi^2}} = \phi^h, \quad \forall h \geq 0$$

### 3. Exercise 3.4 from the textbook.

**SOL:**

$$\begin{aligned} \text{a)} \quad X_t &= .8X_{t-1} - .15X_{t-2} + W_t - .3W_{t-1} \Leftrightarrow X_t - .8X_{t-1} + .15X_{t-2} = W_t - .3W_{t-1} \Leftrightarrow \\ &\Leftrightarrow (1 - .8B + .15B^2)X_t = (1 - .3B)W_t \Leftrightarrow (1 - .5B) \cancel{(1 - .3B)} X_t = \cancel{(1 - .3B)} W_t \Leftrightarrow \\ &\Leftrightarrow (1 - .5B)X_t = W_t \Leftrightarrow X_t = .5X_{t-1} + W_t \end{aligned}$$

which is an AR(1) model with  $\phi_1 = .5$ . Since  $|\phi_1| < 1$ , the model is causal ( $\rightarrow$  stationary) and invertible (b/c all pure AR models are invertible).

$$\text{b)} \quad X_t = X_{t-1} - .5X_{t-2} + W_t - W_{t-1} \Leftrightarrow X_t - X_{t-1} + .5X_{t-2} = W_t - W_{t-1} \Leftrightarrow \underbrace{(1 - B + .5B^2)}_{\phi(B)} X_t = \underbrace{(1 - B)}_{\theta(B)} W_t$$

The roots of the AR polynomial are given by:

$$\begin{aligned} \phi(z) = 0 &\Rightarrow .5z^2 - z + 1 = 0 \Rightarrow z = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a} = \frac{-(-1) \pm \sqrt{(-1)^2 - 4(.5)(1)}}{2(.5)} \\ &\Rightarrow z = \frac{1 \pm \sqrt{1-2}}{1} = 1 \pm \sqrt{-1} = 1 \pm i \end{aligned}$$

These are conjugate complex roots with norm  $|1 \pm i| = \sqrt{1^2 + 1^2} = \sqrt{2} > 1$ , so the model is causal. However, the root of the MA polynomial  $\theta(z) = 1 - z = 0$  is trivially  $z = 1$ , which lies *on* the unit circle  $\rightarrow$  the model is *not* invertible.

### 4. Exercise 3.5 from the textbook.

**SOL:**

For the AR(2) model  $\overbrace{(1 - \phi_1 B - \phi_2 B^2)}^{=\phi(B)} X_t = W_t$  to be stationary, we need the roots of the characteristic equation  $1 - \phi_1 z - \phi_2 z^2 = 0$  to be outside the unit disk. Let  $z_1, z_2$  be the possibly complex roots; we can equivalently write the equation as  $1 - \phi_1 z - \phi_2 z^2 = (1 - z_1^{-1} z)(1 - z_2^{-1} z) = 1 - (z_1^{-1} + z_2^{-1})z + (z_1^{-1} z_2^{-1})z^2$ , from which we get that  $\phi_1 = z_1^{-1} + z_2^{-1}$  &  $\phi_2 = -z_1^{-1} z_2^{-1}$ . We want  $|z_1|, |z_2| > 1 \Rightarrow |z_1^{-1}|, |z_2^{-1}| < 1$ , which implies:

$$\phi_1 = z_1^{-1} + z_2^{-1} \Rightarrow |\phi_1| = |z_1^{-1} + z_2^{-1}| \leq |z_1^{-1}| + |z_2^{-1}| < 1 + 1 = 2$$

$$\phi_2 = -z_1^{-1} z_2^{-1} \Rightarrow |\phi_2| = |-z_1^{-1} z_2^{-1}| \leq |z_1^{-1}| |z_2^{-1}| < 1$$

Moreover, the roots are given by  $z = \frac{\phi_1 \pm \sqrt{\phi_1^2 + 4\phi_2}}{-2\phi_2}$ , and they are real provided that

$$\phi_1^2 + 4\phi_2 \geq 0 \Rightarrow \phi_2 \geq -\frac{\phi_1^2}{4}. \text{ Assuming we have two real roots, let } z_1 \leq z_2 \text{ without loss of}$$

generality. For both roots to be outside the unit circle, we have 3 cases  $\begin{cases} a: & 1 < z_1 \leq z_2 \\ b: & z_1 \leq z_2 < -1 \\ c: & z_1 < -1 \text{ \& } 1 < z_2 \end{cases}$ . Let

$\varphi_2 < 0$ , so that  $z_1 = \frac{\varphi_1 - \sqrt{\varphi_1^2 + 4\varphi_2}}{-2\varphi_2} \leq \frac{\varphi_1 + \sqrt{\varphi_1^2 + 4\varphi_2}}{-2\varphi_2} = z_2$ ; the 3 case become:

$$a. 1 < z_1 \Rightarrow 1 < \frac{\varphi_1 - \sqrt{\varphi_1^2 + 4\varphi_2}}{-2\varphi_2} \Rightarrow \sqrt{\varphi_1^2 + 4\varphi_2} < \varphi_1 + 2\varphi_2 \Rightarrow \varphi_1^2 + 4\varphi_2 < (\varphi_1 + 2\varphi_2)^2 \Rightarrow$$

$$\Rightarrow \cancel{\varphi_1^2} + \cancel{4\varphi_2} < \cancel{\varphi_1^2} + \cancel{4\varphi_1\varphi_2} + \cancel{4\varphi_2^2} \stackrel{(\text{div } \varphi_2)}{\Rightarrow} \boxed{1 > \varphi_1 + \varphi_2}$$

$$b. z_2 < -1 \Rightarrow \frac{\varphi_1 + \sqrt{\varphi_1^2 + 4\varphi_2}}{-2\varphi_2} < -1 \Rightarrow \sqrt{\varphi_1^2 + 4\varphi_2} < 2\varphi_2 - \varphi_1 \Rightarrow \varphi_1^2 + 4\varphi_2 < (2\varphi_2 - \varphi_1)^2 \Rightarrow$$

$$\Rightarrow \cancel{\varphi_1^2} + \cancel{4\varphi_2} < \cancel{4\varphi_2^2} - \cancel{4\varphi_1\varphi_2} + \cancel{\varphi_1^2} \stackrel{(\text{div } \varphi_2)}{\Rightarrow} \boxed{1 > \varphi_2 - \varphi_1}$$

$$c. z_2 > 1 \Rightarrow \frac{\varphi_1 + \sqrt{\varphi_1^2 + 4\varphi_2}}{-2\varphi_2} > 1 \Rightarrow \sqrt{\varphi_1^2 + 4\varphi_2} > 2\varphi_2 + \varphi_1 \Rightarrow \varphi_1^2 + 4\varphi_2 > (2\varphi_2 + \varphi_1)^2 \Rightarrow$$

$$\Rightarrow \cancel{\varphi_1^2} + \cancel{4\varphi_2} > \cancel{4\varphi_2^2} + \cancel{4\varphi_1\varphi_2} + \cancel{\varphi_1^2} \stackrel{(\text{div } \varphi_2)}{\Rightarrow} \boxed{1 < \varphi_2 + \varphi_1}, \text{ AND}$$

$$z_1 < -1 \Rightarrow \frac{\varphi_1 - \sqrt{\varphi_1^2 + 4\varphi_2}}{-2\varphi_2} < -1 \Rightarrow \sqrt{\varphi_1^2 + 4\varphi_2} > 2\varphi_2 + \varphi_1 \Rightarrow \varphi_1^2 + 4\varphi_2 > (2\varphi_2 + \varphi_1)^2 \Rightarrow$$

$$\Rightarrow \cancel{\varphi_1^2} + \cancel{4\varphi_2} > \cancel{4\varphi_2^2} + \cancel{4\varphi_1\varphi_2} + \cancel{\varphi_1^2} \stackrel{(\text{div } \varphi_2)}{\Rightarrow} \boxed{1 < \varphi_2 + \varphi_1}$$

Note that for case c. we need  $(1 < \varphi_2 + \varphi_1) \& (1 < \varphi_2 - \varphi_1) \Rightarrow \varphi_2 > 1$ , which the assumption  $\varphi_2 < 0$ , so we reject it. Repeating the above analysis for  $\varphi_2 > 0$ , we get similar results. So, overall, we have the conditions  $(|\varphi_2| < 1) \& (\varphi_2 + \varphi_1 < 1) \& (\varphi_2 - \varphi_1 < 1)$ .

## 5. Exercise 3.6 from the textbook.

**SOL:**

We have  $X_t = -.9X_{t-2} + W_t \Leftrightarrow X_t + .9X_{t-2} = W_t \Leftrightarrow \underbrace{(1 + 0B + .9B^2)}_{\varphi(B)} X_t = W_t$ . The roots of the AR

polynomial are:  $\varphi(z) = 1 + .9z^2 = 0 \Rightarrow z^2 = -\frac{1}{.9} \Rightarrow z = \pm \frac{i}{\sqrt{.9}}$ . The ACF is given by

$\rho(h) = -.9\rho(h-2), \forall h \geq 2$  with initial conditions  $\rho(0) = 1$   $\rho(1) = -.9\rho(-1) = -.9\rho(1) \Rightarrow \rho(1) = 0$

$$\Rightarrow \begin{cases} \rho(2) = -.9\rho(0) = -.9 \\ \rho(3) = -.9\rho(1) = 0 \\ \rho(4) = -.9\rho(2) = (-.9)^2 \Rightarrow \rho(h) = \begin{cases} (-.9)^{h/2}, & h = 0, 2, 4, \dots \\ 0, & h = 1, 3, 5, \dots \end{cases} \\ \rho(5) = -.9\rho(3) = 0 \\ \vdots \end{cases}$$

## 6. Exercise 3.7 from the textbook.

**SOL:**

a)  $X_t + 1.6X_{t-1} + .64X_{t-2} = W_t \Rightarrow X_t = -1.6X_{t-1} - .64X_{t-2} + W_t \Rightarrow \rho(h) = -1.6\rho(h-1) - .64\rho(h-2) \Rightarrow \rho(0) = 1$

$$\rho(1) = -1.6\rho(0) - .64\rho(-1) = -1.6 - .64\rho(1) \Rightarrow \rho(1) = -\frac{1.6}{1.64} = -0.9756098$$

$$\rho(2) = -1.6\rho(1) - .64\rho(0) = 1.6\frac{1.6}{1.64} - .64 = 0.9209756$$

$$\rho(3) = -1.6\rho(2) - .64\rho(1) = \dots = -0.8491707$$

$\vdots$

From R, we get:

> ARMAacf(ar=c(-1.6,-.64), ma=0, 10)

0	1	2	3	4
1.0000000	-0.9756098	0.9209756	-0.8491707	0.7692488
5	6	7	8	9
-0.6873288	0.6074068	-0.5319605	0.4623964	-0.3993796
10				
0.3430736				

b)  $X_t - .4X_{t-1} - .45X_{t-2} = W_t \Rightarrow X_t = .4X_{t-1} + .45X_{t-2} + W_t \Rightarrow \rho(h) = .4\rho(h-1) + .45\rho(h-2) \Rightarrow \rho(0) = 1$

$$\rho(1) = .4\rho(0) + .45\rho(-1) = .4 + .45\rho(1) \Rightarrow \rho(1) = \frac{.4}{.55} = 0.7272727$$

$$\rho(2) = .4\rho(1) + .45\rho(0) = .4\frac{.4}{.55} + .45 = 0.7409091$$

$$\rho(3) = .4\rho(2) + .45\rho(1) = \dots = 0.6236364$$

$\vdots$

> ARMAacf(ar=c(.4,.45), ma=0, 10)

0	1	2	3	4	5
1.0000000	0.7272727	0.7409091	0.6236364	0.5828636	0.5137818
6	7	8	9	10	
0.4678014	0.4183224	0.3778396	0.3393809	0.3057802	

$$\text{c) } X_t - 1.2X_{t-1} + .85X_{t-2} = W_t \Rightarrow X_t = 1.2X_{t-1} - .85X_{t-2} + W_t \Rightarrow \rho(h) = 1.2\rho(h-1) - .85\rho(h-2) \Rightarrow \rho(0) = 1$$

$$\rho(1) = 1.2\rho(0) - .85\rho(-1) = 1.2 - .85\rho(1) \Rightarrow \rho(1) = \frac{1.2}{1.85} = 0.64864865$$

$$\rho(2) = 1.2\rho(1) - .85\rho(0) = 1.2 \frac{1.2}{1.85} - .85 = -0.07162162$$

$$\rho(3) = 1.2\rho(2) - .85\rho(1) = \dots = -0.63729730$$

⋮

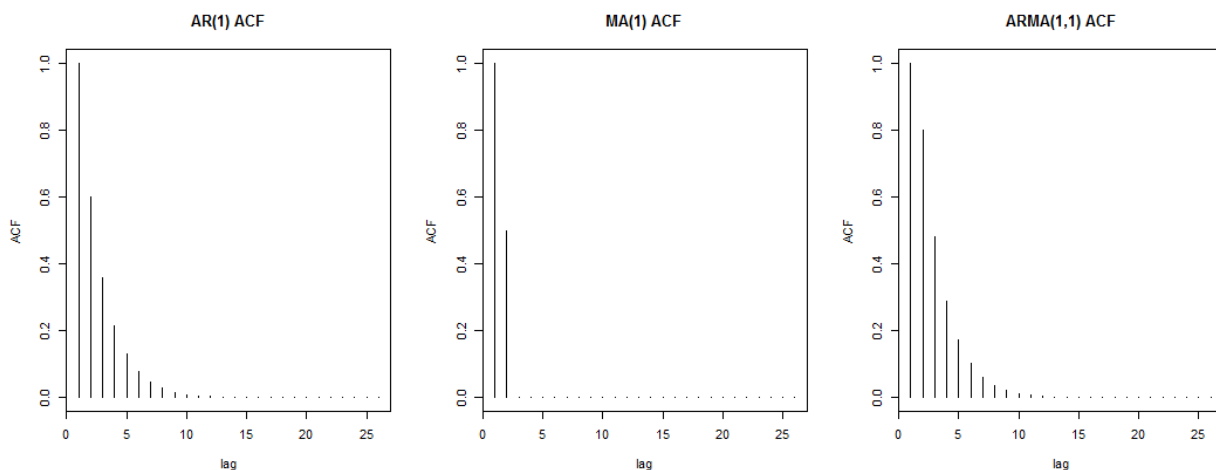
```
> ARMAacf(ar=c(1.2,-.85), ma=0, 10)
```


0	1	2	3	4
1.00000000	0.64864865	-0.07162162	-0.63729730	-0.70387838
5	6	7	8	9
-0.30295135	0.23475500	0.53921465	0.44751583	0.07868654
10				
-0.28596460				

7. Exercise 3.8 from the textbook.

**SOL:**

The calculations of the general ARMA(1,1) ACF are verified in the example from Lecture 7, p. 14-15. The resulting ACF's for the various models using  $\phi = .6$  &  $\theta = .9$  are:



 **Code:**

# Q3.8

```
phi=.6; theta=.9
```

```
AR_1.acf=ARMAacf(ar=phi, lag=25)
```

```
MA_1.acf=ARMAacf(ma=theta, lag=25)
```

```
ARMA_1.1.acf=ARMAacf(ar=phi, ma=theta, lag=25)
```

```
par(mfrow=c(1,3))
```

```
plot(AR_1.acf, type='h', main="AR(1) ACF", xla='lag', ylab='ACF' )
plot(MA_1.acf, type='h', main="MA(1) ACF", xla='lag', ylab='ACF' )
plot(ARMA_1.1.acf, type='h', main="ARMA(1,1) ACF", xla='lag',
ylab='ACF' )
```

8. Determine whether the following AR models are stationary, and calculate the first 4 coefficients of their causal representation.

a.  $X_t = -.8X_{t-1} + .4X_{t-2} + W_t$

b.  $X_t = -.5X_{t-1} + .4X_{t-2} + W_t$

**SOL:**

**a.**

The AR model is stationary if its characteristic polynomial has all roots outside the unit circle.

We have  $X_t + .8X_{t-1} - .4X_{t-2} = W_t \Rightarrow \varphi(B)X_t = W_t$  where  $\varphi(z) = 1 + .8z - .4z^2 \Rightarrow$

$$z = \frac{-.8 \pm \sqrt{(.8)^2 - 4 \times (-.4) \times 1}}{2 \times (-.4)} = \frac{.8 \pm \sqrt{.64 + 1.6}}{.8} = 1 \pm \sqrt{\frac{2.24}{.8}} = 1 \pm \sqrt{3.5} = \begin{cases} 2.870829 \\ -0.870827 \end{cases}$$

Since only one root is outside the unit circle, the model is not stationary, and it does not have a causal representation (i.e. you can try to invert the AR polynomial operator, but the resulting  $\psi$  weights will be exploding to  $\infty$ ). The first 4 terms are  $\psi_1 = -0.8$ ,  $\psi_2 = 1.04$ ,  $\psi_3 = -1.152$ ,  $\psi_4 = 1.3376$ ,  $\psi_5 = -1.53088$ .

**b.**

We have  $X_t + .5X_{t-1} - .4X_{t-2} = W_t \Rightarrow \varphi(B)X_t = W_t$  where  $\varphi(z) = 1 + .5z - .4z^2 \Rightarrow$

$$z = \frac{-.5 \pm \sqrt{(.5)^2 - 4 \times (-.4) \times 1}}{2 \times (-.4)} = \frac{.5 \pm \sqrt{.25 + 1.6}}{.8} = \frac{.5 \pm \sqrt{1.85}}{.8} = \begin{cases} 2.325184 \\ -1.075184 \end{cases}$$

Since *both* roots are outside the unit disk (i.e.  $|z| > 1$ ), the model is stationary and its causal representation is given by:

$$\psi(B)\varphi(B) = 1 \Leftrightarrow (1 + \psi_1 B + \psi_2 B^2 + \dots)(1 + .5B - .4B^2) = 1$$

$$\Rightarrow 1 + (\psi_1 + .5)B + (\psi_2 + .5\psi_1 - .4)B^2 + \dots = 1$$

$$\Rightarrow \begin{cases} \psi_1 + .5 = 0 \Rightarrow \psi_1 = -.5 \\ \psi_2 + .5\psi_1 - .4 = 0 \Rightarrow \psi_2 = -.5\psi_1 + .4 = .5^2 + .4 = .65 \\ \psi_3 + .5\psi_2 - .4\psi_1 = 0 \Rightarrow \psi_3 = -.5\psi_2 + .4\psi_1 = -.5 \times .65 - .4 \times .5 = -.525 \\ \psi_4 + .5\psi_3 - .4\psi_2 = 0 \Rightarrow \psi_4 = -.5\psi_3 + .4\psi_2 = +.5 \times .525 + .4 \times .65 = .5225 \end{cases}$$