

Assignment 2

Derek Situ (62222807)

Question 1

(1a)

We will show that the process $X_t = \frac{2}{5}X_{t-1} + \frac{1}{4}X_{t-2} - \frac{1}{10}X_{t-3} + Z_t$ is stationary by showing that the roots of the characteristic polynomial lie outside the unit circle in the complex plane. We can rearrange to get

$$\begin{aligned} Z_t &= X_t - \frac{2}{5}X_{t-1} - \frac{1}{4}X_{t-2} + \frac{1}{10}X_{t-3} \\ &= (1 - \frac{2}{5}B - \frac{1}{4}B^2 + \frac{1}{10}B^3)X_t. \end{aligned}$$

Now we have to find b such that

$$1 - \frac{2}{5}b - \frac{1}{4}b^2 + \frac{1}{10}b^3 = 0.$$

We multiply both sides by 20 to work with integers, and group the terms like so:

$$\begin{aligned} (2b^3 - 5b^2) - (8b - 20) &= 0 \\ b^2(2b - 5) - 4(2b - 5) &= 0 \\ (b^2 - 4)(2b - 5) &= 0 \end{aligned}$$

Now we can see that this is satisfied when $b = -2, 2, \frac{5}{2}$. And since all of these lie outside the unit circle on the complex plane, $\{X_t\}_{t \in \mathbb{N}}$ is stationary.

(1b)

First we multiply both sides of $X_t = \frac{2}{5}X_{t-1} + \frac{1}{4}X_{t-2} - \frac{1}{10}X_{t-3} + Z_t$ by X_{t-k} for $k = 1, 2, 3$ to get 3 equations

$$X_t X_{t-1} = \frac{2}{5}X_{t-1}X_{t-1} + \frac{1}{4}X_{t-2}X_{t-1} - \frac{1}{10}X_{t-3}X_{t-1} + Z_t X_{t-1} \quad (1)$$

$$X_t X_{t-2} = \frac{2}{5}X_{t-1}X_{t-2} + \frac{1}{4}X_{t-2}X_{t-2} - \frac{1}{10}X_{t-3}X_{t-2} + Z_t X_{t-2} \quad (2)$$

$$X_t X_{t-3} = \frac{2}{5}X_{t-1}X_{t-3} + \frac{1}{4}X_{t-2}X_{t-3} - \frac{1}{10}X_{t-3}X_{t-3} + Z_t X_{t-3} \quad (3)$$

Next we take expectation of both sides of each equation to get equations 4-6

$$\gamma(1) = \frac{2}{5}\gamma(0) + \frac{1}{4}\gamma(1) - \frac{1}{10}\gamma(2) \quad (4)$$

$$\gamma(2) = \frac{2}{5}\gamma(1) + \frac{1}{4}\gamma(0) - \frac{1}{10}\gamma(1)$$

$$= \frac{1}{4}\gamma(0) + \frac{3}{10}\gamma(1) \quad (5)$$

$$\gamma(3) = \frac{2}{5}\gamma(2) + \frac{1}{4}\gamma(1) - \frac{1}{10}\gamma(0) \quad (6)$$

Substituting (5) into (4) yields

$$\gamma(1) = \frac{2}{5}\gamma(0) + \frac{1}{4}\gamma(1) - \frac{1}{10}[\frac{1}{4}\gamma(0) + \frac{3}{10}\gamma(1)]$$

and after some simple algebraic manipulations we see

$$\gamma(1) = \frac{150}{312}\gamma(0).$$

Dividing $\gamma(1)$, as well as equations (5) and (6), by $\gamma(0)$ yields

$$\begin{aligned} \rho(1) &= \frac{\frac{150}{312}\gamma(0)}{\gamma(0)} \\ &= \frac{150}{312} = 0.4807692 \end{aligned} \quad (7)$$

$$\begin{aligned} \rho(2) &= \frac{3}{10}\rho(1) + \frac{\frac{1}{4}\gamma(0)}{\gamma(0)} \\ &= \frac{3}{10}(\frac{150}{312}) + \frac{1}{4} = 0.3942308 \end{aligned} \quad (8)$$

$$\begin{aligned} \rho(3) &= \frac{2}{5}\rho(2) + \frac{1}{4}\rho(1) - \frac{\frac{1}{10}\gamma(0)}{\gamma(0)} \\ &= \frac{2}{5}[\frac{3}{10}(\frac{150}{312}) + \frac{1}{4}] + \frac{1}{4}(\frac{150}{312}) - \frac{1}{10} = 0.1778846 \end{aligned} \quad (9)$$

Now we wish to find the roots d such that $d^3 - \frac{2}{5}d^2 - \frac{1}{4}d + \frac{1}{10} = 0$. We find the roots in the same fashion as in question (1a):

$$20d^3 - 8d^2 - 5d + 2 = 0$$

$$4d^2(5d - 2) - (5d - 2) = 0$$

$$(4d^2 - 1)(5d - 2) = 0$$

We see that this is satisfied when $d = -\frac{1}{2}, \frac{2}{5}, \frac{1}{2}$. Armed with the values of $\rho(1), \rho(2), \rho(3), d_1, d_2, d_3$ we can solve for A_1, A_2, A_3 in the following system of equations:

$$\begin{aligned} \rho(1) &= A_1(\frac{-1}{2}) + A_2(\frac{2}{5}) + A_3(\frac{1}{2}) \\ \rho(2) &= A_1(\frac{-1}{2})^2 + A_2(\frac{2}{5})^2 + A_3(\frac{1}{2})^2 \\ \rho(3) &= A_1(\frac{-1}{2})^3 + A_2(\frac{2}{5})^3 + A_3(\frac{1}{2})^3 \end{aligned}$$

We can represent this system of equations as a matrix and reduce it to identify A_1, A_2 , and A_3 :

```
library(pracma)
d <- c(-1/2, 2/5, 1/2)
rho <- c(0.4807692, 0.3942308, 0.1778846)
mat <- matrix(c(d, rho[1], d^2, rho[2], d^3, rho[3]),
              nrow = 3, ncol = 4, byrow = TRUE)
A <- rref(mat)[,4]
A
```

```
## [1] 0.1794873 -1.6025639 2.4230768
```

We see that $A_1 = 0.1794873$, $A_2 = -1.6025639$, and $A_3 = 2.4230768$ so that

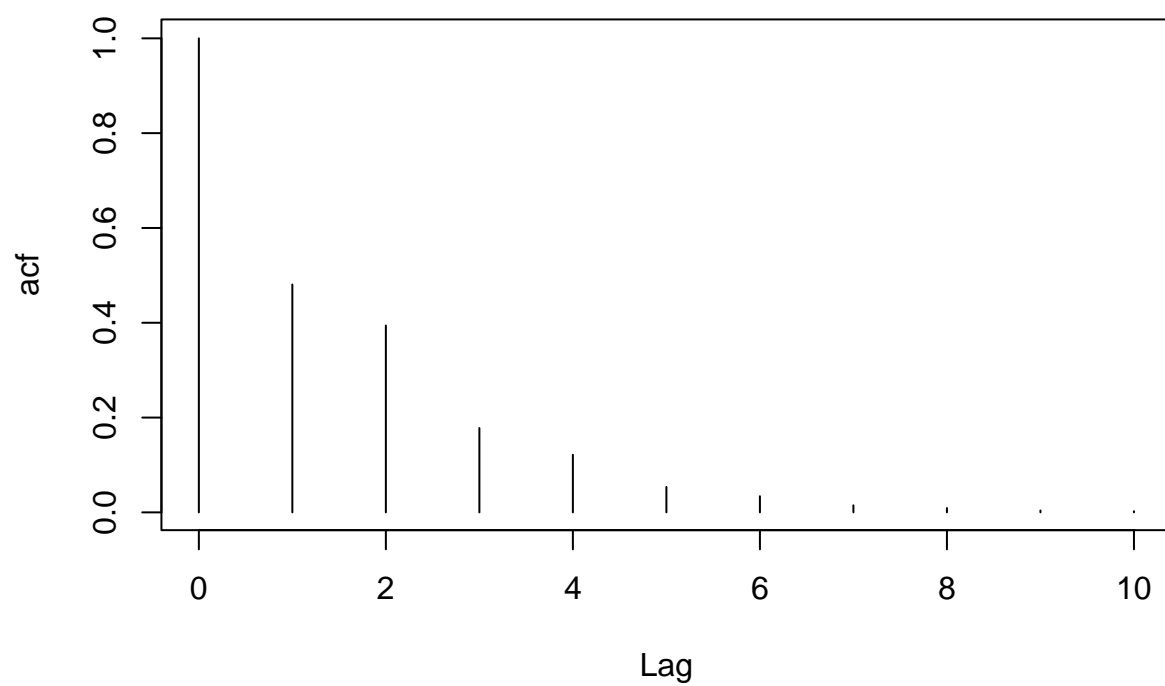
$$\rho(h) = 0.1794873\left(\frac{-1}{2}\right)^{|h|} - 1.6025639\left(\frac{2}{5}\right)^{|h|} + 2.4230768\left(\frac{1}{2}\right)^{|h|}.$$

(1c)

```
# simulate Xt
set.seed(123)
sim <- arima.sim(list(ar = c(2/5, 1/4, -1/10)), n = 2000, sd = sqrt(1.96))

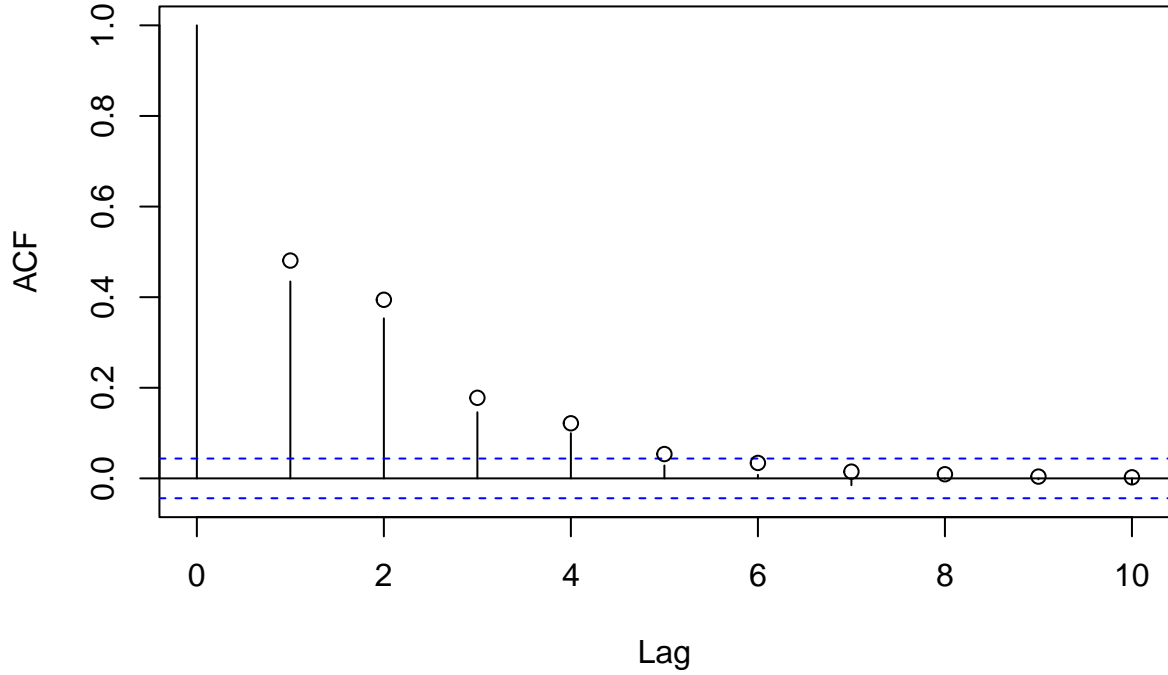
# plot just the theoretical acf
rho_h <- function(h) {
  A[1] * d[1] ^ h + A[2] * d[2] ^ h + A[3] * d[3] ^ h
}
acf_theory <- c(1)
for (i in 2:11) {
  acf_theory[i] <- rho_h(i-1)
}
plot(x = c(0:10), y = acf_theory, type = "h", xlab = "Lag", ylab = "acf",
     main = "Theoretical ACF")
```

Theoretical ACF



```
# plot the sample acf with theoretical acf overlayed on top  
acf(sim, lag.max = 10)  
points(acf_theory[2:11])
```

Series sim



Question 2

(2a)

We can rewrite the process as

$$(1 - \frac{7}{10}B)X_t = (1 - \frac{1}{10}B)Z_t$$

where $(1 - \frac{7}{10}B) = \phi(B)$ and $(1 - \frac{1}{10}B) = \theta(B)$. Then the roots b_1 and b_2 of $\phi(B)$ and $\theta(B)$ are $\frac{10}{7}$ and 10, both of which have modulus greater than unity, so the process is stationary and invertible.

(2b)

We first identify the polynomial $\psi(B) = \frac{\theta(B)}{\phi(B)} = \frac{1 - \frac{1}{10}B}{1 - \frac{7}{10}B}$ which can be expanded to the geometric sum

$$\begin{aligned}
 \psi(B) &= (1 - \frac{1}{10}B)[1 + \frac{7}{10}B + (\frac{7}{10})^2B^2 + \dots] \\
 &= [1 + \frac{7}{10}B + (\frac{7}{10})^2B^2 + \dots] - \frac{1}{10}[B + \frac{7}{10}B^2 + (\frac{7}{10})^2B^3 + \dots] \\
 &= 1 + \sum_{j=1}^{\infty} [0.7^j - 0.1(0.7)^{j-1}]B^j \\
 &= 1 + \sum_{j=1}^{\infty} (0.7)^{j-1}(0.6)B^j
 \end{aligned}$$

Then the model can be expressed as a pure MA process $X_t = \psi(B)Z_t$ or

$$X_t = [1 + \sum_{j=1}^{\infty} (0.7)^{j-1} (0.6) B^j] Z_t.$$

(2c)

We first identify the polynomial $\pi(B) = \frac{\phi(B)}{\theta(B)} = \frac{1-0.7B}{1-0.1B}$ which can be expanded to the geometric sum

$$\begin{aligned} \pi(B) &= (1 - 0.7B)(1 + 0.1B + 0.1^2 B^2 + \dots) \\ &= 1 + 0.1B + 0.1^2 B^2 + \dots - 0.7(B + 0.1B^2 + 0.1^2 B^3 + \dots) \\ &= 1 + \sum_{i=1}^{\infty} [0.1^i - 0.7(0.1)^{i-1}] B^i \\ &= 1 + \sum_{i=1}^{\infty} (0.1)^{i-1} (-0.6) B^i \end{aligned}$$

Then the model can be expressed as a pure AR process $Z_t = \pi(B)X_t$ or

$$Z_t = [1 + \sum_{i=1}^{\infty} (0.1)^{i-1} (-0.6) B^i] X_t.$$

(2d)

To derive the acf will need $Var(X_t)$.

$$\begin{aligned} Var(X_t) &= Var(Z_t) + 0.6^2 Var(Z_{t-1} + 0.7Z_{t-2} + 0.7^2 Z_{t-3} + \dots) \\ &= \sigma^2 + 0.6^2 (\sigma^2 + 0.7^2 \sigma^2 + 0.7^4 \sigma^2 + \dots) \\ &= \sigma^2 + \frac{0.6^2 \sigma^2}{1 - 0.7^2} \\ &= \sigma^2 (1 + \frac{0.6^2}{1 - 0.7^2}). \end{aligned}$$

Now we find $Cov(X_t, X_{t+1})$. Recall $X_t = \psi(B)Z_t = \sum_{i=0}^{\infty} \psi_i B^i Z_t = \sum_{i=0}^{\infty} \psi_i Z_{t-i}$ where $\psi_i = 0.6(0.7)^{i-1}$ for $j = 1, 2, \dots$ and $\psi_0 = 1$.

$$\begin{aligned} Cov(X_t, X_{t+1}) &= E(X_t X_{t+1}) \\ &= E(\sum_{i=0}^{\infty} \psi_i Z_{t-i} \sum_{j=0}^{\infty} \psi_j Z_{t-j+1}) \end{aligned}$$

Since $E(Z_{t-a} Z_{t-b}) = \sigma^2$ when $a = b$ and 0 otherwise, we have

$$\begin{aligned} Cov(X_t, X_{t+1}) &= \sigma^2 (\sum_{i=0}^{\infty} \psi_i \psi_{i+1}) \\ &= \sigma^2 (0.6 + \sum_{i=1}^{\infty} \psi_i \psi_{i+1}) \end{aligned}$$

$$\begin{aligned}
&= \sigma^2(0.6 + \sum_{i=1}^{\infty} 0.6^2(0.7)^{2i+1}) \\
&= \sigma^2(0.6 + \frac{0.6^2(0.7)}{1-0.7^2})
\end{aligned}$$

Next we will need $Cov(X_t, X_{t+h})$.

$$\begin{aligned}
Cov(X_t, X_{t+h}) &= E(X_t X_{t+h}) \\
&= E(\sum_{i=0}^{\infty} \psi_i Z_{t-i} \sum_{j=0}^{\infty} \psi_j Z_{t-j+h})
\end{aligned}$$

Since $E(Z_{t-a} Z_{t-b}) = \sigma^2$ when $a = b$ and 0 otherwise, we have

$$\begin{aligned}
Cov(X_t, X_{t+h}) &= \sigma^2(\sum_{i=0}^{\infty} \psi_i \psi_{i+h}) \\
&= \sigma^2[0.6(0.7^{|h|-1}) + 0.6^2(0.7^{|h|}) + 0.6^2(0.7^{|h|+2}) + 0.6^2(0.7^{|h|+4}) + \dots] \\
&= \sigma^2[0.6(0.7^{|h|-1}) + \frac{0.6^2(0.7^{|h|})}{1-0.7^2}] \\
&= 0.7^{|h|-1} Cov(X_t, X_{t+1}).
\end{aligned}$$

Then

$$\begin{aligned}
\rho(1) &= \frac{Cov(X_t, X_{t+1})}{Var(X_t)} = \frac{\sigma^2(0.6 + \frac{0.6^2(0.7)}{1-0.7^2})}{\sigma^2((1 + \frac{0.6^2}{1-0.7^2}))} \\
&= 0.6413793, \\
\rho(h) &= \frac{Cov(X_t, X_{t+h})}{Var(X_t)} = \frac{0.7^{|h|-1} Cov(X_t, X_{t+1})}{Var(X_t)} \\
&= 0.7^{|h|-1} \rho(1) \\
&= 0.7^{|h|-1} * 0.6413793
\end{aligned}$$

for $h > 0$, and $\rho(0) = \frac{Var(X_t)}{Var(X_t)} = 1$.

Question 3

Let $\{X_t\}$ be a SARIMA(2, 1, 0) \times (0, 1, 2)₁₂ process:

$$(1 - \alpha_1 B^1 - \alpha_2 B^2) W_t = (1 + \beta_1 B^{12} + \beta_2 B^{24}) Z_t$$

where $W_t = (1 - B^{12})[(1 - B)X_t]$. We will use this fact later but will keep the W_t term for now. Distributing the terms gives

$$W_t - \alpha_1 W_{t-1} - \alpha_2 W_{t-2} = Z_t + \beta_1 Z_{t-12} + \beta_2 Z_{t-24}$$

and now writing W_t in terms of X_t gives

$$X_t - X_{t-1} - X_{t-12} + X_{t-13} - \alpha_1(X_{t-1} - X_{t-2} - X_{t-13} + X_{t-14}) - \alpha_2(X_{t-2} - X_{t-3} - X_{t-14} + X_{t-15}) = Z_t + \beta_1 Z_{t-12} + \beta_2 Z_{t-24}.$$

Without a need to simplify further we can see that this is an ARMA(15, 24) process.