Lab Quiz

SID: 470408326

4th April 2022

R version

The R version used is 4.1.01

Question 1

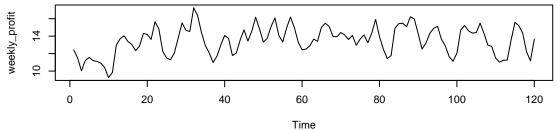
Question 2

We replace data with diff(data)

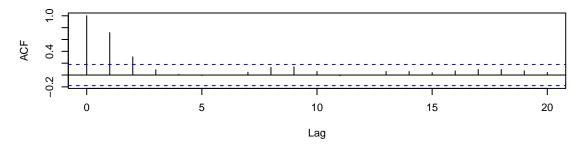
```
data = c(12.1, 13.2, 13.8, 13.5, 14.3, 15.2, 12.4, 14.5, 12.4, 11.7, 13.4, 12.8,
    11.6, 13.2, 11.1, 10.9)
m = 4
weights = c(1/(2 * m), rep(1/m, m - 1), 1/(2 * m))
filter(diff(data), weights)
## Time Series:
## Start = 1
## End = 15
## Frequency = 1
                     NA 0.5250 0.0750 -0.0500 -0.1125 -0.6750 -0.3125 -0.0875
## [1]
             NA
## [10] -0.3125  0.0875 -0.1000 -0.5250
The first two and last two de-trended values are: 0.5250, 0.0750, -0.1, -0.525
weekly_profit = c(12.43, 11.51, 10.04, 11.21, 11.56, 11.2, 11.11, 10.92, 10.4, 9.25,
    9.85, 12.97, 13.71, 14.04, 13.39, 13.05, 12.34, 12.86, 14.31, 14.2, 13.63, 15.64,
    14.83, 12.26, 11.51, 11.29, 12.07, 13.8, 15.53, 14.66, 14.54, 17.25, 16.37, 14.43,
```

```
12.95, 12.09, 10.98, 11.7, 12.99, 14.08, 13.75, 11.78, 12.05, 13.55, 14.72, 13.44, 14.56, 16.17, 14.84, 13.32, 13.75, 15.12, 16.09, 14.07, 13.33, 14.93, 16.2, 15, 13.32, 12.45, 12.51, 12.89, 13.66, 13.42, 15.02, 15.48, 15.11, 13.96, 13.93, 14.42, 14.14, 13.62, 14.09, 12.96, 13.66, 14.14, 13.24, 14.33, 15.92, 13.96, 12.52, 11.43, 11.77, 14.91, 15.44, 15.48, 15.08, 16.23, 15.97, 14.28, 12.52, 13.19, 14.32, 14.9, 15.13, 13.65, 12.88, 11.65, 11.13, 12.04, 14.72, 15.22, 14.61, 14.35, 14.41, 15.49, 14.32, 12.97, 12.83, 11.5, 11.02, 11.24, 11.26, 13.54, 15.57, 15.21, 14.34, 12.21, 11.17, 13.65)

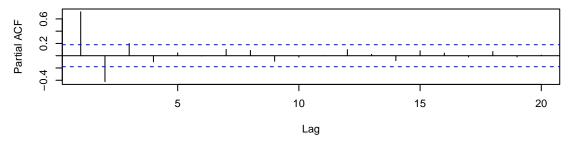
weekly_profit_mean = mean(weekly_profit)
weekly_profit_var = var(weekly_profit)
par(mfrow = c(3, 1))
ts.plot(weekly_profit)
rhos = acf(weekly_profit)
pis = pacf(weekly_profit)
```



Series weekly_profit



Series weekly_profit



We see that the time series seems stationary with constant mean 13.5 and variance 2.5955832. This is supported by a quickly decaying ACF plot.

Question 4

We stored the ACF and PACF values of the time series from before:

rhos

```
##
## Autocorrelations of series 'weekly_profit', by lag
##
##
                                3
                                                               7
                                                                                     10
                        2
                                               5
                                                       6
                1
##
    1.000
            0.716
                   0.306
                           0.089
                                   0.011 - 0.007
                                                  0.000
                                                          0.048
                                                                  0.129
                                                                          0.138
                                                                                 0.060
##
               12
                       13
                               14
                                      15
                                              16
                                                      17
                                                              18
                                                                      19
                                                                             20
       11
  -0.013
           0.000
                   0.061
                           0.060
                                   0.039
                                          0.068
                                                  0.093
                                                          0.094
                                                                  0.071
```

```
##
## Partial autocorrelations of series 'weekly_profit', by lag
##
##
                                             6
                                                                                 11
                  0.200 -0.100
                                                                             0.001
##
    0.716 - 0.422
                                 0.047 -0.003
                                                0.102
                                                       0.086 -0.091 -0.023
       12
              13
                      14
                             15
                                     16
                                            17
                                                   18
                                                           19
                                                                  20
   0.099 0.022 -0.080 0.081
                                 0.042 -0.020
                                                0.068 -0.020
```

- The first 5 ACF values at lags 1 to 5 are: 0.716, 0.306, 0.089, 0.011, -0.007
- The first 5 PACF values at lags 1 to 5 are: 0.716, -0.422, 0.200, -0.100, 0.047

Question 5

We are given the model

$$X_t = 0.7X_{t-1} + 0.8X_{t-3} + Z_t - 0.7Z_{t-1} + 0.3Z_{t-2} + 0.5Z_{t-3}$$
 $Z_t \sim NID(0, 1.7^2)$

Which can be written as:

$$(1 - 0.7B - 0.8B^3)X_t = (1 - 0.7B + 0.3B^2 + 0.5B^3)Z_t$$

Hence:

- AR(3) polynomial is $a(\omega) = 1 0.7\omega 0.8\omega^3$
- MA(3) polynomial is $b(\omega) = 1 0.7\omega + 0.3\omega^2 + 0.5\omega^3$

We find the modulus of the roots using R

```
# For the AR(3) polynomial
ar_roots = polyroot(c(1, -0.7, 0, 0.8))
abs(ar_roots)

## [1] 0.9645315 1.3436223 0.9645315

ma_roots = polyroot(c(1, -0.7, 0.3, 0.5))
abs(ma_roots)
```

```
## [1] 1.027231 1.895369 1.027231
```

- The ARMA(3,3) model **is not** stationary because not all of it's roots for the AR(3) polynomial have a modulus greater than 1
- The ARMA(3,3) model **is** invertible because all of it's roots for the MA(3) polynomial are greater than 1

Question 6

We simulate 600 values

```
set.seed(142)
de = 15 + arima.sim(model = list(ma = c(0.5, 0.9)), n = 600)[201:600]
fit = arima(de, order = c(0, 0, 2))
##
## Call:
## arima(x = de, order = c(0, 0, 2))
## Coefficients:
##
             ma1
                      ma2 intercept
          0.5434 0.8761
                              15.1299
## s.e. 0.0235 0.0269
                               0.1161
## sigma^2 estimated as 0.9259: log likelihood = -553.69, aic = 1115.37
  (i) Estimates are:
  • \hat{\mu} = 15.1299
   • \hat{\beta}_1 = 0.5430
   • \hat{\beta}_2 = 0.8761
   • \hat{\sigma}^2 = 0.9259
 (ii) The AIC is: 1115.37
```

Question 7

```
fit2 = arima(de, order = c(1, 0, 2))
fit2
##
## Call:
## arima(x = de, order = c(1, 0, 2))
## Coefficients:
##
              ar1
                       ma1
                                ma2
                                      intercept
          0.0283 0.5372 0.8740
##
                                         15.1296
## s.e. 0.0569 0.0270 0.0276
                                          0.1191
## sigma^2 estimated as 0.9254: log likelihood = -553.56, aic = 1117.13
  (i) Estimates are:
   • \hat{\mu} = 15.1296
   • \hat{\beta}_1 = 0.5372
   • \hat{\beta}_2 = 0.8740
   • \hat{\sigma}^2 = 0.9254
   • \hat{\alpha}_1 = 0.0283
 (ii) The AIC is: 1117.13
```

(iii) Since the AIC is lower for model in Q6 we prefer that one.

Question 8

```
fit3 = arima(de, order = c(2, 0, 1))
fit3
##
## Call:
## arima(x = de, order = c(2, 0, 1))
##
## Coefficients:
##
                              ma1 intercept
             ar1
                      ar2
                                     15.1201
         -0.0409 0.5148 0.4788
##
        0.0801 0.0512 0.0823
                                       0.1548
## s.e.
##
## sigma^2 estimated as 1.225: log likelihood = -608.49, aic = 1226.99
  • ar1 se = 0.0801 and t ratio = -0.5106
  • ar2 se = 0.0512 and t ratio = 10.05
  • ma1 se = 0.0823 and t ratio = 5.83
  • intercept se = 0.1548 and t ratio = 100.07
```

Question 9

```
Box.test(resid(fit), lag = 12, type = "Ljung")

##
## Box-Ljung test
##
## data: resid(fit)
## X-squared = 4.5378, df = 12, p-value = 0.9717

X-squared = 4.5378 and p-value = 0.9717
```

Question 10

Didn't finish on time