Problem 12:

$$X_{t} = C + \varphi_{1} \times_{t-1} + \varphi_{2} \times_{t-2} + \varepsilon_{t}$$

$$C = 10 - \varphi_1 E[X_{t-1}] - \varphi_2 E[X_{t-2}] =$$

0

$$= 10 - \varphi_1 \cdot 10 - \varphi_2 \cdot 10$$

$$C = 10 - \varphi_{1} = 10 - \varphi_{2} = 10$$

$$P(1) = \frac{\varphi}{1 + \theta \cdot P(1)} = 0,5$$

$$p(2) = \frac{\varphi \cdot P(1) + \theta}{1 + \theta \cdot P(1)} = 0,4$$

$$\theta = \rho(2) - \varphi \cdot \rho(1)$$

# (DO 3-4000)

$$X_{t} = C + (0,5(1+0.0,5)) \cdot X_{t-1} + (0,4-(0,5(1+0.0,5)))$$

• 
$$\boldsymbol{\varepsilon}_{t-1}$$
 +  $\boldsymbol{\varepsilon}_{t}$ 

(b) 
$$X_{\xi} = \mu + \mathcal{E}_{\xi} + \theta_{1} \mathcal{E}_{\xi-1} + \theta_{2} \mathcal{E}_{\xi-2}$$

$$\rho(1) = \frac{\theta_{1}}{1+\theta_{2}} = 0.5 \qquad \rho(2) = \theta_{1} \cdot \rho(1) + \theta_{2} / (1+\theta_{2}) = 0.4$$

$$\theta_{1} = 0.5 \cdot (1+\theta_{2})$$

$$\theta_{2} = (0.99 - \theta_{1} \cdot \rho(1)) / (1-\rho(1))$$

$$\mu = 10 - \theta_{1} \mathcal{E}[\mathcal{E}_{\xi-1}] = \theta_{2} \mathcal{E}[\mathcal{E}_{\xi-2}] =$$

$$= 10 - \theta_{1} \cdot 0 = \theta_{2} \cdot 0 = 10$$

$$var(\mathcal{E}_{\chi}) = 5$$

$$X_{\xi} = 10 + \mathcal{E}_{\xi} + (0.5 \cdot (1+\theta_{2})) - \mathcal{E}_{\xi-1} +$$

$$+ ((0.99 - (0.5 \cdot (1+\theta_{2}))) / (1-0.5)) \cdot \mathcal{E}_{\xi-2}$$

$$\rho(3) = \varphi_{1} \cdot \rho(2) + \varphi_{2} \cdot \rho(1)$$

$$\rho(9) = \varphi_{1} \cdot \rho(2) + \varphi_{2} \cdot \rho(1)$$

$$\rho(9) = \varphi_{1} \cdot \rho(3) + \varphi_{2} \cdot \rho(2)$$

$$\varphi_{1} = 0.5 - \varphi_{2} \qquad \varphi_{2} = 0.99 - \varphi_{1} \cdot 0.5 \qquad \rho(1) = 0.5 \qquad \rho(2) = 0.99$$

$$\varphi = 0.5 \cdot (1 + 0.0.5) \quad \rho(2) = 0.4$$

## MA(2):

R

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$$P(3) = \theta_1 \cdot P(2) + \theta_2 \cdot P(1)$$

$$P(4) = \theta_1 P(3) + \theta_2 P(2)$$

### ts4.R

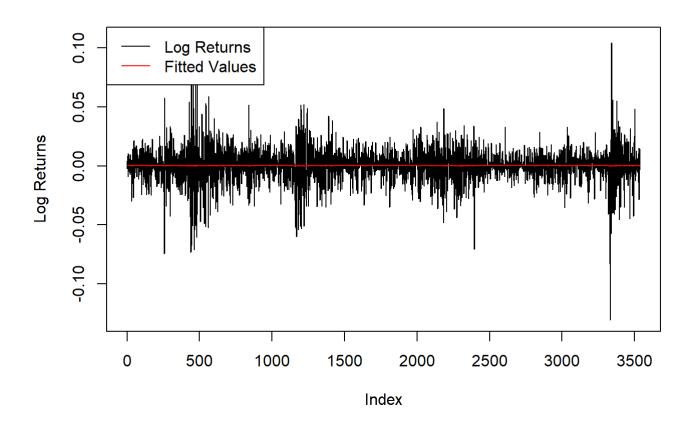
#### Acer

### 2023-05-17

```
library(dplyr)
library(stringr)
library(data.table)
library(ggplot2)
library (mondate)
library (zoo)
library (VGAM)
library (MASS)
library(car)
library(forecast)
library(conflicted)
library(tidyr)
library(scales)
library(reshape2)
library (psych)
library(rmarkdown)
conflict prefer('select', 'dplyr')
## [conflicted] Removing existing preference.
## [conflicted] Will prefer dplyr::select over any other package.
conflict prefer('filter', 'dplyr')
## [conflicted] Removing existing preference.
## [conflicted] Will prefer dplyr::filter over any other package.
conflict_prefer('lag', 'dplyr')
## [conflicted] Removing existing preference.
## [conflicted] Will prefer dplyr::lag over any other package.
setwd("C:\\Users\\Acer\\OneDrive - ADA University\\Documents\\ec")
data <- readRDS("dax.rds")</pre>
data <- as.data.frame(data)</pre>
model \leftarrow arima(data\$data, order = c(1, 0, 0))
summary(model)
```

```
##
## Call:
## arima(x = data$data, order = c(1, 0, 0))
## Coefficients:
##
            arl intercept
        0.0051
                     2e-04
##
  s.e. 0.0168
                     2e-04
##
## sigma^2 estimated as 0.0002029: log likelihood = 10018.24, aic = -20030.48
## Training set error measures:
##
                                  RMSE
                                                MAE
                                                         MPE
                                                                 MAPE
                                                                           MASE
                          ME
ACF1
## Training set 5.141891e-08 0.01424499 0.009700564 104.7648 110.2888 0.6841773 5.614084
e-05
```

```
plot(data$data, type = "l", ylab = "Log Returns")
lines(fitted(model), col = "red")
legend("topleft", legend = c("Log Returns", "Fitted Values"), col = c("black", "red"), l
ty = 1)
```



```
##
## Call:
## arima(x = data$data^2, order = c(1, 0, 0))
##
## Coefficients:
##
           arl intercept
##
        0.1159
                   2e-04
## s.e. 0.0167
                  0e+00
##
## sigma^2 estimated as 4.052e-07: log likelihood = 21011.59, aic = -42017.18
##
## Training set error measures:
##
                                  RMSE
                                                MAE
                                                        MPE
                                                               MAPE
                                                                          MASE
ACF1
## Training set -9.119537e-09 0.0006365475 0.0002426972 -1442400 1442423 0.8415227 -0.02
279896
```

```
plot(data$data^2, type = "l", ylab = "Squared Log Returns")
lines(fitted(model), col = "red")
legend("topleft", legend = c("Squared Log Returns", "Fitted Values"), col = c("black",
"red"), lty = 1)
```

