# Problem 3 (6 credits)

#### HW2

Danny Moncada (monca016) February 22, 2020

```
suppressWarnings(suppressPackageStartupMessages({
   library(TSA)
   library(forecast)
   library(ggplot2)
   library(dplyr)
}))
```

## Boston Crime Data Analysis

### Question 1

Please pull out the crime frequency data we got from Homework 1 - Problem 3 - Question 4. You may re-plot the time series to refresh yourself about the pattern.

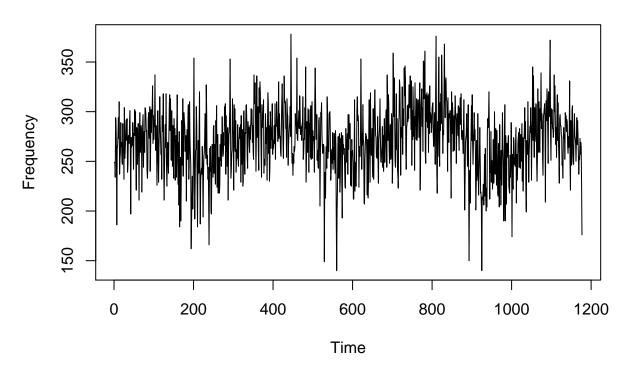
```
setwd("~/MSBA 2020 All Files/Spring 2020/MSBA 6430 - Advanced Issues in Business Analytics/HW2")
crime=read.table("crime.txt",header=T)

N=dim(crime)[1]
crime_aggr=aggregate(rep(1,N),list(year=crime[,1],month=crime[,2],day=crime[,3]),sum)
crime_aggr=crime_aggr[order(crime_aggr[,1],crime_aggr[,2],crime_aggr[,3]),]

Y = crime_aggr[,4]

ts.plot(Y, ylab = "Frequency", main = "Boston Crime From June 2015 to September 2018")
```

# **Boston Crime From June 2015 to September 2018**



### a) (1 credit)

First, let's fit an auto.arima() to find out a good ARIMA model for the data. Again, notice that, auto.arima() provides a "good" model but not necessarily the optimal. We will learn more concrete model selection techniques in Lecture 6.

#### Hints:

• use auto.arima() function

```
arima_fit <- auto.arima(Y)
arima_fit</pre>
```

```
## Series: Y
## ARIMA(1,0,3) with non-zero mean
##
##
   Coefficients:
##
            ar1
                               ma2
                                        ma3
                                                 mean
                      ma1
         0.9888
##
                           -0.2542
                                     0.0446
                                             270.6409
         0.0054
                   0.0298
                            0.0347
                                     0.0292
                                               5.5130
##
## sigma^2 estimated as 880.5: log likelihood=-5658.27
## AIC=11328.53
                  AICc=11328.61
                                   BIC=11358.96
```

#### b) (2 credits)

What's the model? For example

$$(Y_t - 10) = 0.4 \cdot (Y_{t-1} - 10) + e_t - 0.8 \cdot e_{t-1}$$

#### Hints:

- The mean value comes with every  $Y_t$ . In the example above, the mean value is 10.
- R assumes positive sign for MA models. In the example above, R would show -0.8, rather than +0.8 for the MA(1) coefficient

Please write down the model below:

$$Y_t = 0.99 \cdot (Y_{t-1} - 270.64) + e_t - 0.71 \cdot e_{t-1} - 0.25 \cdot e_{t-2} + 0.04 \cdot e_{t-3} + 270.64$$

#### c) (1 credit)

Are any of the coefficients significant?

#### Hints:

• A coefficient is significant if its magnitude is (roughly) at least twice as large as its standard error.

```
## ar1 ma1 ma2 ma3 mean

## 0.9888 -0.7142 -0.2542 0.0446 270.6409

##s.e. 0.0054 0.0298 0.0347 0.0292

"The first, second, and third coefficients are significant."
```

## [1] "The first, second, and third coefficients are significant."

```
"The standard error*2 for ar1 is 0.0108, while the coefficient is 0.9888."
```

## [1] "The standard error\*2 for ar1 is 0.0108, while the coefficient is 0.9888."

```
"The standard error*2 for ma1 is 0.0596, while the coefficient is 0.7142."
```

## [1] "The standard error\*2 for mal is 0.0596, while the coefficient is 0.7142."

```
"The standard error*2 for ma2 is 0.0694, while the coefficient is 0.2542."
```

## [1] "The standard error\*2 for ma2 is 0.0694, while the coefficient is 0.2542."

#### d) (2 credits)

Please superimpose the fitted values on the original crime frequency time series. Does the model sufficiently explain the data?

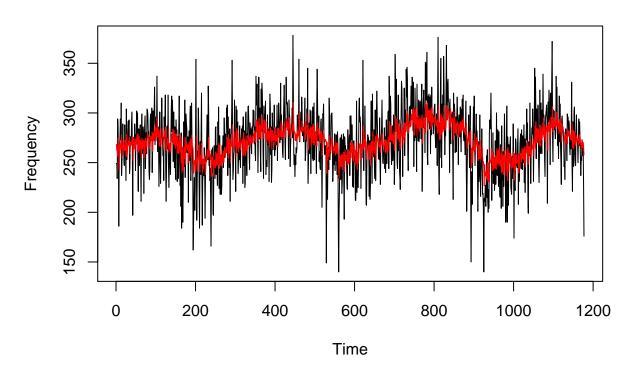
#### Hints:

• The fitted values can be calculated by the original time series - arima\_fit\$residuals

```
fitted_vals = Y - arima_fit$residuals

par(mfrow=c(1,1))
ts.plot(Y, ylab = "Frequency", main = "Boston Crime From June 2015 to September 2018")
lines(fitted_vals,col="red")
```

# **Boston Crime From June 2015 to September 2018**



"The auto arima model DOES sufficiently explain the data - it generally follows the process laid out by

## [1] "The auto arima model DOES sufficiently explain the data - it generally follows the process laid
"It doesn't capture the fringe points but not every model can be perfect."

## [1] "It doesn't capture the fringe points but not every model can be perfect."