AAEC 4484/ STAT(AAEC) 5484: Applied Economic Forecasting

Your Name Here

Homework #2 - Spring 2023

The purpose of this assignment is to enhance your understanding of time series and data patterns. It is intended to be rather straightforward and simple.

Instructions:

- Where necessary, please ensure that your graphs and visuals have proper titles and axes labels.
- Recall that you can use help() or ?seriesname in your consoles to get a general overview of the dataset.

Question 1: Basic Visualization of Time Series Data

- Create time plots of the following time series: Bricks from aus_production, Hare from pelt, Closing prices for AMZN from gafa_stock, Demand from vic_elec.
- Please use the grid.arrange function from the gridExtra package to arrange your plots as a 2 x 2 grid.
- ii. Briefly discuss any discernible pattern(s) you might have noticed in the data.

Question 2: Assessing (Potential) Seasonality

A cool feature of R is the ability to draw a "random" sample of your data. In the code chunk that follows, we will pull a random draw of the aus_retail data set by selecting a Series ID according to our seed. You will need to change the seed to ensure that we both get a different draw of the data.

```
set.seed(12345678) #Change this to your preferred seed
myseries <- aus_retail %>%
  filter(`Series ID` == sample(aus_retail$`Series ID`,1))
```

- i. Use the autoplot(), gg_season(), gg_subseries(), and ACF() %>% autoplot() functions to explore possible seasonality in your chosen time series.
- Please use the grid.arrange() function from the gridExtra package to arrange your plots. You are free to organize them however you wish.
- It might be useful to change the lag_max (how about to 3 years of data?) in the ACF to ensure that you can see a fair bit of the pattern in the correlogram.
- ii. What can you say about the series? Are there any seasonal patterns? Trends?

Question 3: White Noise

The aus_livestock series contains data on monthly "Meat production in Australia for human consumption".

- i. Using the filter() function, extract the number of pigs slaughtered between 1990 & 1999, inclusively. Store this variable as pigs. Along with other necessary conditions, I would like to see you use the year() and %in% commands to extract the relevant time periods.
- ii. Produce the autoplot() of this series and its correlogram. Comment on any pattern noticed in both. Does this series look like white noise?
- iii. Now, using the difference() and mutate() functions, create a new column, diff that computes the quarterly changes (lag = 1) in your pigs series. Store this new data as d.pigs. I would suggest using the head() command on your d.pigs series in the console to ensure that the diff column looks as you would expect.

```
#Remember to remove eval = FALSE to ensure your code runs.
#This code should get you started
____ <- ___ %>% ____ (___ = difference(____, lag = 1))
```

- iii. Produce an autoplot() of diff along with the associated correlogram. Does the first differenced data now look like white noise? Did differencing remove any potential seasonality in the data? Recall that a simple yes will not suffice. You will need to explain your conclusion.
- iv. Return to the pigs series and again, use the difference() and mutate() functions, to create a new column called diff12 that computes the quarterly changes (lag = 12) in your pigs series. Store this new data as d12.pigs. I would suggest using the head(n = 14) command on your d12.pigs series in the console to ensure that the diff12 column looks as you would expect.
- v. Produce an autoplot of diff12 along with the associated correlogram.
- a. Does the data, differenced at the monthly frequency, now look like white noise? Recall that a simple yes will not suffice. You will need to explain your conclusion.
- b. Did differencing remove any potential seasonality or trend in the data? If not, can you think of a way that you would solve this? You are not required to do this. Just an intelligent answer based on the plots will suffice.