

Assignment 1

Social Forecasting

Due: March 29, 2024

Instructions

- When submitting your answers, please 1) show your code and 2) make sure your presentation is clear and easy to follow. Rmarkdown will make it easy to produce a suitable report, but you are welcome to use another approach if you prefer.

Answer Task 1

```
1 # Install packages
2 install.packages("tidyverse")
3 install.packages("forecast")
4 install.packages("zoo")
5
6 # Load the libraries
7 library(tidyverse)
8 library(forecast)
9 library(zoo)
10
11 #TASK 1
12 # Import the dataset
13 rd <- read_csv("/Users/iseli/Downloads/AFG_IRL_refugeeData.csv")
```

Answer Task 2

```
1 # Extract the column "refugees under UNHCR mandate" and create the time
  series object
2 rdts <- ts(rd[,6], start=1999, end=2021, frequency=1)
3
4 # Print the time series object
5 print(rdts)
```

Answer Task 3

```
1 # Create the plot using autoplot
2 plot_rdt <- autoplot(rdt) +
3   labs(title = "Time Series of Refugees under UNHCR Mandate",
4         x = "Year",
5         y = "Number of Refugees") +
6   theme_minimal()
7
8 # Print the plot -FIRST PLOT-
9 print(plot_rdt)
```

1 Forecasts vs Actual Data

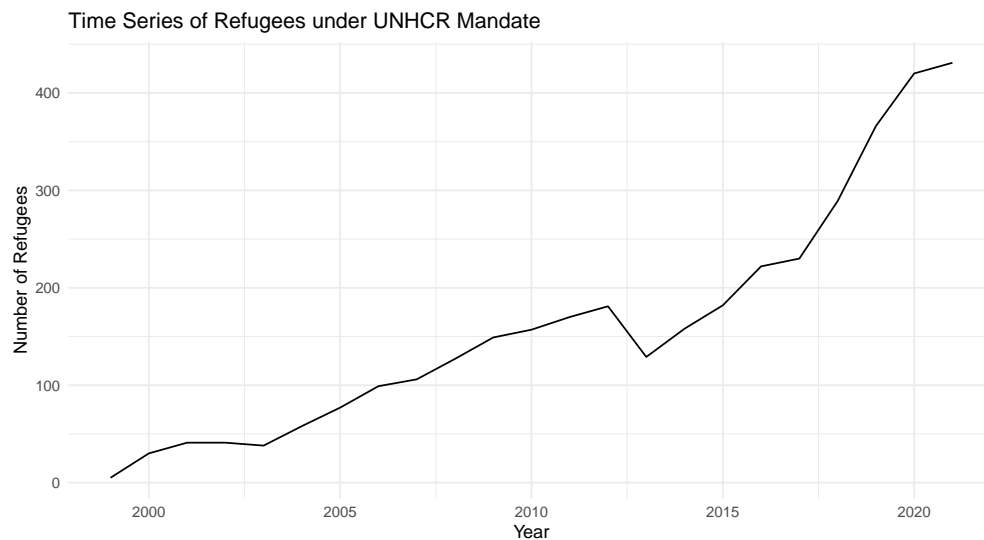


Figure 1: Forecasts vs Actual Data (2016-2021)

Answer Task 4

```
1 # Subset the time series to include only data from 1999 to 2015
2 rdt_subset <- window(rdt, start=1999, end=2015)
3
4 # Calculate the mean of the subsetted time series
5 mean_value <- mean(rdt_subset)
6
7 # Create a time series object of the mean value for the forecast years
8   2016 to 2021
9 forecast_values <- rep(mean_value, times=6)
```

```

9
10 # Create a ts object for the forecasted values
11 forecast_ts <- ts(forecast_values, start=2016, end=2021, frequency=1)
12
13 # Print the forecasted values
14 print(forecast_ts)
15
16 # Just making sure forecast package is already loaded
17 if (!requireNamespace("forecast", quietly = TRUE)) {
18   install.packages("forecast")
19 }
20 library(forecast)
21
22 # Create the naive forecast
23 naive_forecast <- naive(rdts_subset, h=6)
24
25 # Print the forecasted values
26 print(naive_forecast)
27 naive_forecast <- forecast::naive(rdts_subset, h=6)$mean
28
29 # Just double checking that packages are already loaded
30 if (!requireNamespace("forecast", quietly = TRUE)) {
31   install.packages("forecast")
32 }
33 library(forecast)
34
35 # rwf() function with drift=TRUE to create the drift forecast
36 drift_forecast <- rwf(rdts_subset, h=6, drift=TRUE)
37
38 # Print the forecasted values
39 print(drift_forecast)
40 drift_forecast <- forecast::rwf(rdts_subset, h=6, drift=TRUE)$mean
41
42 # Extract the actual data for the years 2016 to 2021
43 actual_data <- window(rdts, start=2016, end=2021)
44
45 #Encountering many errors, -I will redo some codes:
46 # Subset the time series to include only data from 1999 to 2015
47 rdts_subset <- window(rdts, start=1999, end=2015)
48
49 # Calculate the average forecast
50 average_value <- mean(rdts_subset)
51 average_forecast <- rep(average_value, times=6)
52 average_forecast_ts <- ts(average_forecast, start=2016, end=2021,
53   frequency=1)
54
55 # Calculate the naive forecast
56 naive_forecast <- forecast::naive(rdts_subset, h=6)
57 naive_forecast_values <- naive_forecast$mean
58
59 # Calculate the drift forecast

```

```

59 drift_forecast <- forecast::rwf(rdts_subset, h=6, drift=TRUE)
60 drift_forecast_values <- drift_forecast$mean

```

Answer Task 5

```

1 # Extract the actual data for the years 2016 to 2021 from the original
  time series object
2 actual_data <- window(rdts, start=2016)
3
4 # Combine all forecasts and actual data into a data frame for plotting
5 combined_data <- data.frame(
6   Year = rep(2016:2021, times=4),
7   Value = c(average_forecast_ts, naive_forecast_values, drift_forecast_
  values, actual_data),
8   Method = rep(c("Average", "Naive", "Drift", "Actual"), each=6)
9 )
10
11 # Load ggplot2 for plotting
12 library(ggplot2)
13
14 # Create the combined plot
15 combined_plot <- ggplot(combined_data, aes(x = Year, y = Value, color =
  Method, group = Method)) +
16   geom_line() +
17   geom_point() +
18   labs(title = "Forecasts vs Actual Data (2016–2021)",
19     x = "Year",
20     y = "Number of Refugees",
21     color = "Method") +
22   theme_minimal()
23
24 # Print the combined plot –SECOND PLOT–
25 print(combined_plot)
26
27 #Interpretation of plot
28 #The actual data line shows the real number of refugees from Afghanistan
  to Ireland for each year from 2016 to 2021
29 #By comparing the actual data with the forecasts, it's evident that there
  are variations in the actual number of refugees over these years
30 #The actual data line allows us to evaluate the accuracy of each
  forecasting method.
31 #Based on the plot, it appears that none of the forecasting methods
  perfectly match the actual data, which is expected in real-world
  scenarios due to the complexity and unpredictability of factors
  influencing refugee movements

```

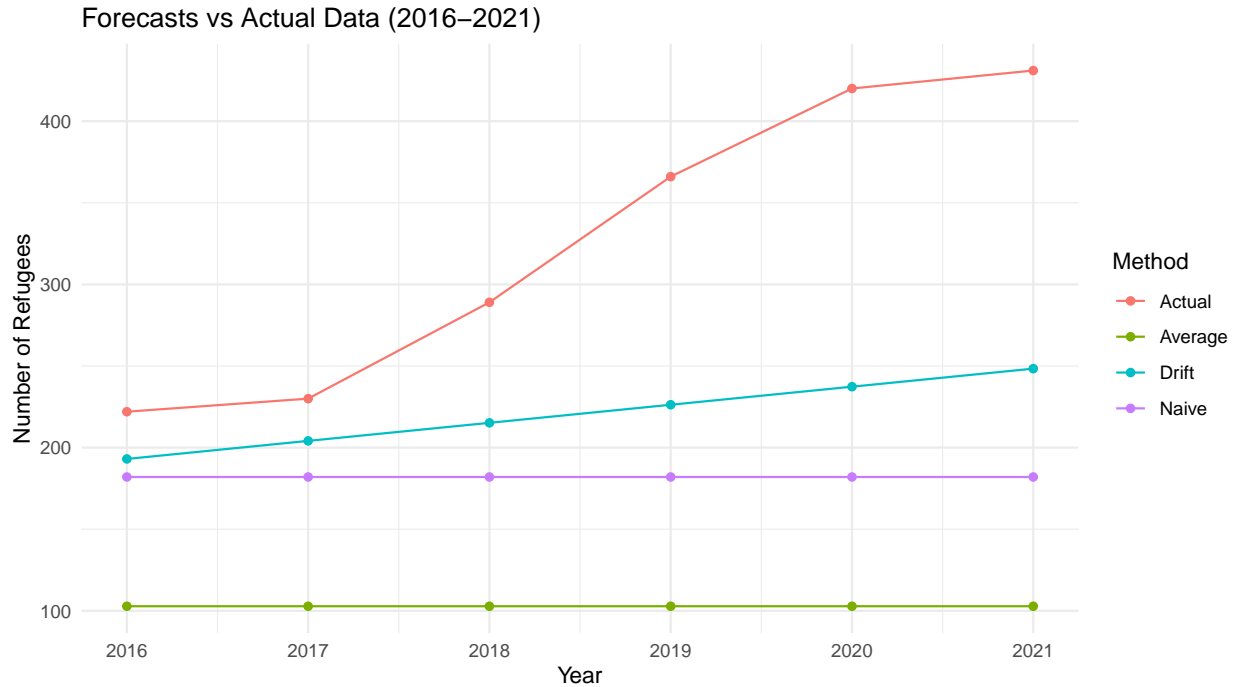


Figure 2: Forecasts vs Actual Data (2016-2021)

2 Forecasts vs Actual Data

As shown in Figure 2, the drift method provides the closest approximation to the actual data from 2016 to 2021.

Answer Task 6

- 1 #The Drift method, might offer a slightly more flexible approach compared to the static forecasts of the Average and Naive methods.
- 2 #I would not particularly suggest an improvement. The effectiveness of each method would depend on how well it aligns with the actual observed trends in the data.