MA615 Final Project

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MBTA Data EDA

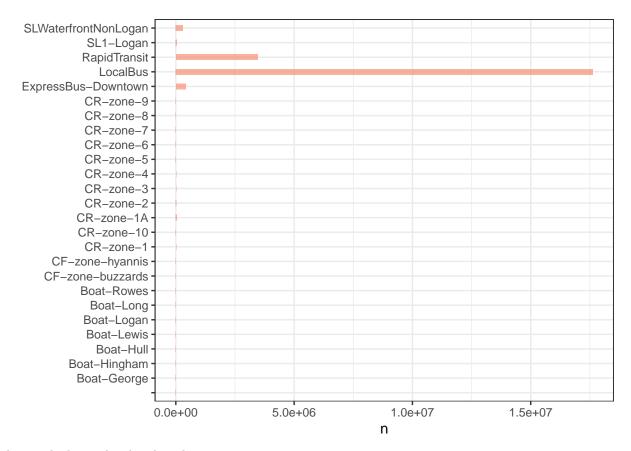
This report includes the data from MBTA in last 12 months with all categories of transit (https://cdn.mbta.com/archive/archived_feeds.txt), and I picked one week from each month. The dataset is large, to run code faster and drop the useless data, I will do the data cleaning first.

In this report, I'm gonna graph the distribution of the stops. Besides, I found that typicality is worthy of analysis. According to the reference(https://github.com/mbta/gtfs-documentation/blob/master/reference/gtfs.md#calendar_attributestxt), in this dataset, current valid values are: 0 (or empty): Not defined; 1: Typical service with perhaps minor modifications; 2: Extra service supplements typical schedules; 3: Reduced holiday service is provided by typical Saturday or Sunday schedule; 4: Major changes in service due to a planned disruption, such as construction; 5: Major reductions in service for weather events or other atypical situations.

Read Data and Data Cleaning

Count stops

```
newdata_df %>%
  ggplot( aes(x=zone_id, y=n)) +
    geom_bar(stat="identity", fill="#f68060", alpha=.6, width=.4) +
    coord_flip() +
    xlab("") +
    theme_bw()
```

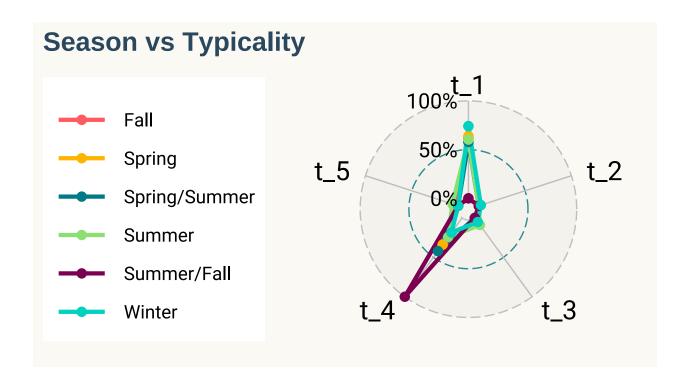


The graph shows that bus has the most stops.

Season vs Typicality

```
#Clean Data
new_cadata <- ca_data %>%
  count(rating_description, service_schedule_typicality, sort = TRUE) %>%
  pivot_wider(names_from = service_schedule_typicality, values_from = n)
new_cadata[is.na(new_cadata)] <- 0</pre>
new_cadata <- new_cadata %>%
  mutate(t_1 = 1)/(1+2+3+3+4+5),
         t_2 = 2^{-2}/(1^{+2}+3^{+3}+4^{+5}),
         t_3 = 3^{(1)} + 2^{+3} + 4^{+5}
         t_4 = 4^{(1)}/(1^{+2}+3^{+3}+4^{+5}),
         t_5 = 5^{(1+2+3+3+5)}
new_cadata <- new_cadata[,c(1,7:11)]</pre>
library(ggradar)
library(palmerpenguins)
library(tidyverse)
library(scales)
library(showtext)
font_add_google("Roboto", "roboto")
```

```
showtext_auto()
#plot1
#code citation: gg gallery
plot_1 <- new_cadata %>%
  ggradar(
   font.radar = "roboto",
    grid.label.size = 7,
   axis.label.size = 7,
   group.point.size = 3
  )+
  labs(title = "Season vs Typicality") +
  theme(
     plot.background = element_rect(fill = "#fbf9f4", color = "#fbf9f4"),
     panel.background = element_rect(fill = "#fbf9f4", color = "#fbf9f4"),
     plot.title.position = "plot",
      plot.title = element_text(
        family = "lobstertwo",
        size = 20,
        face = "bold",
        color = "#2a475e"
    )
  )
plot_1
```

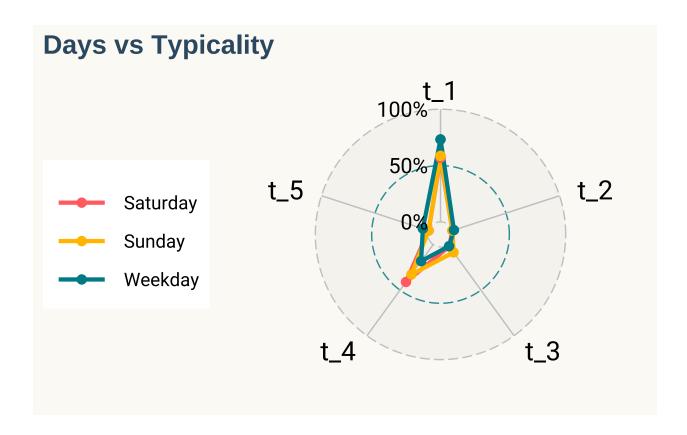


According to this graph, we can know that in most seasons, the most typical service reported is typical

service with perhaps minor modifications, but in summer/fall, major changes in service due to a planned disruption becomes the most, such as construction. Therefore, it can be concluded as MBTA generally tend to be constructed in summer/fall.

Days vs Typicality

```
plot_2 <- new_data_days %>%
  ggradar(
   font.radar = "roboto",
   grid.label.size = 7,
   axis.label.size = 7,
   group.point.size = 3) +
  labs(title = "Days vs Typicality") +
  theme(
      plot.background = element_rect(fill = "#fbf9f4", color = "#fbf9f4"),
      panel.background = element_rect(fill = "#fbf9f4", color = "#fbf9f4"),
      plot.title.position = "plot",
      plot.title = element_text(
       family = "lobstertwo",
       size = 20,
        face = "bold",
        color = "#2a475e"
   )
 )
plot_2
```



This graph shows that weekdays or weekends do not influence the typical service that much, weekdays have the less type4(Major changes in service due to a planned disruption, such as construction), MBTA might consider that the construction cannot influence the traffic in weekdays, so this kind of issue might be moved to saturdays and sundays.