## NYPD Shooting Incident Data Report

2023-02-14

```
library(tidyverse)
library(lubridate)
```

### Getting the data

We first get the raw data from the following link: "https://data.cityofnewyork.us/api/views/833y-fsy8/rows.csv?accessType=DOWNLOAD"

```
urlData <- "https://data.cityofnewyork.us/api/views/833y-fsy8/rows.csv?accessType=DOWNLOAD"
nypd_raw <- read_csv(urlData)</pre>
```

### Tidying and Transforming the Data

Several columns can be taken into account to draw any type of analysis. For my analysis, I will look at the shooting trend on a yearly basis while keeping the different NYC boroughs into consideration. Because we are only focusing on year of event and borough, we can remove most of the unnecessary columns for this analysis.

```
nypd_tr<- nypd_raw %>%
  mutate(Year = as.integer(substr(OCCUR_DATE,nchar(OCCUR_DATE)-3,nchar(OCCUR_DATE)))) %>%
  select(Year, BORO)
summarized_byBORO <- nypd_tr %>%
   group_by(Year,BORO) %>%
   summarize(shootings = n()) %>%
   pivot wider(names from = BORO, values from = shootings)%>%
   ungroup() %>%
   mutate(TOTAL = rowSums(across(-Year)))
nypd_perHour <- nypd_raw %>%
     mutate(Hour = hour(OCCUR TIME) ) %>%
     select(Hour) %>%
     group_by(Hour)%>%
     summarize(Shootings = n()) %>%
     ungroup()
nypd_perHour_prop <- nypd_perHour %>%
     mutate(Prop_Shootings = Shootings/sum(Shootings)*100)
```

#### Visualizations and Analysis

```
grouped_nypd_tr <- nypd_tr %>%
  group_by(Year,BORO) %>%
  summarize(shootings = n())

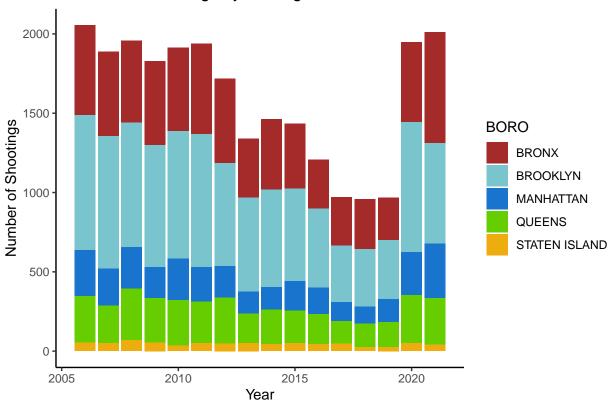
nypd_wide <- nypd_tr %>%
  group_by(Year,BORO) %>%
  summarize(shootings = n()) %>%
  pivot_wider(names_from = BORO,values_from = shootings)

nypd_wide
```

```
## # A tibble: 16 x 6
## # Groups: Year [16]
##
       Year BRONX BROOKLYN MANHATTAN QUEENS 'STATEN ISLAND'
##
      <int> <int>
                     <int>
                               <int> <int>
                                                      <int>
##
  1 2006
              568
                       850
                                 288
                                        296
                                                         53
## 2 2007
                                 233
              533
                       833
                                        238
                                                         50
## 3 2008
              520
                       785
                                 259
                                        326
                                                         69
## 4 2009
                       770
                                        278
                                                         55
              529
                                 196
## 5 2010
              525
                       805
                                 260
                                        288
                                                         34
## 6 2011
              571
                       839
                                 215
                                        264
                                                         50
  7 2012
                       651
                                 196
                                        290
                                                         49
##
              531
## 8 2013
              371
                       593
                                 138
                                        185
                                                         52
## 9 2014
              446
                       614
                                 143
                                        218
                                                         43
## 10 2015
                       583
                                 187
                                        205
                                                         50
              409
## 11 2016
                       498
                                                         44
              308
                                 167
                                        191
## 12 2017
              306
                       357
                                        144
                                                         46
                                 117
## 13 2018
              313
                       365
                                 105
                                        150
                                                         25
## 14 2019
              267
                       372
                                 146
                                        156
                                                         26
## 15 2020
              504
                       819
                                 272
                                        303
                                                         50
## 16 2021
              701
                       631
                                 343
                                        296
                                                         40
```

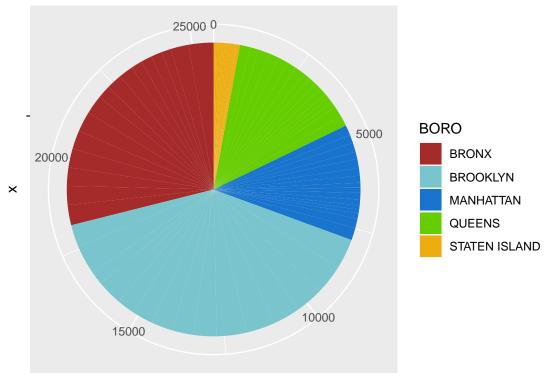
```
ggplot(grouped_nypd_tr, aes(x=Year, y=shootings, fill=BORO)) +
    geom_bar(stat="identity") +
    labs(title="Number of Shootings by Borough and Year") +
    xlab("Year") + ylab("Number of Shootings") +
    scale_fill_manual(values=c("brown", "cadetblue3", "dodgerblue3", "chartreuse3", "darkgoldenrod2"))
    theme_classic()
```

## Number of Shootings by Borough and Year



```
grouped_nypd_tr %>% ggplot(aes(x="", y=shootings, fill=BORO)) +
    geom_bar(stat="identity", width=1) +
    coord_polar("y", start=0) +
    labs(title="Total Shootings per borough from 2006 to 2021") +
    scale_fill_manual(values=c("brown", "cadetblue3", "dodgerblue3", "chartreuse3", "darkgoldenrod2"))
```

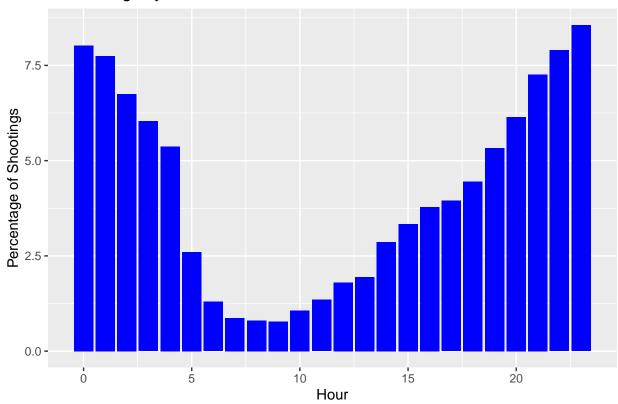
Total Shootings per borough from 2006 to 2021



shootings

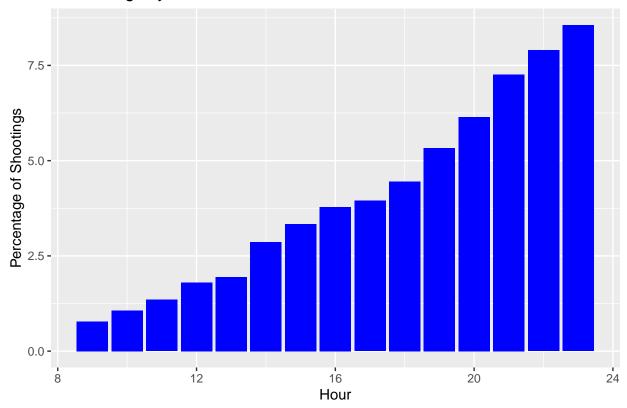
```
ggplot(nypd_perHour_prop , aes(x=Hour, y = Prop_Shootings))+
    geom_bar(stat="identity", fill = "blue")+
    labs(title="%Shootings by Hour", x="Hour", y="Percentage of Shootings")
```

# %Shootings by Hour



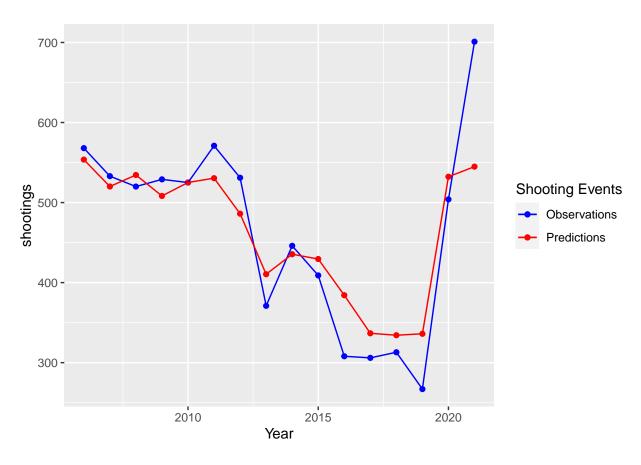
```
nypd_perHour_prop_10 <-nypd_perHour_prop[nypd_perHour_prop$Hour>8,]
ggplot(nypd_perHour_prop_10 , aes(x=Hour, y = Prop_Shootings))+
    geom_bar(stat="identity", fill = "blue")+
    labs(title="%Shootings by Hour", x="Hour", y="Percentage of Shootings")
```

### %Shootings by Hour



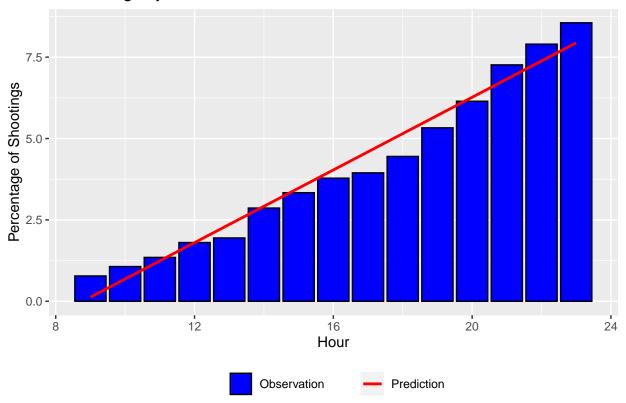
We can see clearly that some boroughs have more shooting incidents than others. The proportion of shooting incidents per borough over the years seems to be constant as well. We can also notice a steady decrease from 2006 to 2019 and a sharp increase in year 2020. The COVID pandemic and the lockdowns it caused might have played a role in this sudden increase of violence in 2020. If this is the case, we might experience a decrease from 2022 onward as it stabilizes around the shooting rates prior to year 2020. Since the proportions seem to be stable, we will try to build a model that predicts the amount of shooting events per borough on a given year based on the year total.

### **Modelling Data**



## Warning: Using 'size' aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use 'linewidth' instead.

### %Shootings by Hour



Our initial intuition was very close to the observed data. We based our model on total events and we assumed a constant proportion of events per borough. We can notice that the prediction line is following the observation line closely up until the last recorded data of 2021.

#### **Bias Identification**

By looking at the different graphs and at the data, we could argue that the boroughs of Brooklyn and Bronx are the most dangerous and that Staten Island is the safest. However, we are not taking population into account in our study. Bronx and Brooklyn are the most populous boroughs in New York City. It would make sense that, everything else being equal, these regions have the highest number of incidents. A less biased and more informative study would've been to cross reference the populations of these boroughs and calculate a per capita rate instead. Moreover, the sudden increase in 2020 could lead to several misinterpretations. This period was greatly affected by the COVID pandemic and several social and societal issues were caused by it. These issues might have contributed to the rise of violence during that period of time.