NYPD Shooting Incident Project

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This dataset is a list of every shooting incident that occurred in NYC from 2006 through the end of 2022. This dataset can be found at https://data.cityofnewyork.us/Public-Safety/NYPD-Shooting-Incident-Data-Historic-/833y-fsy8. It is provided by the NYPD and available for use by the public. Each record represents a shooting in NYC and includes information about the event, location, and the date/time. Also, information about the gender, age, and race of the victim and suspect is included. I'm going to do an analysis to see how the number of shootings changes over the years. I'm also going to see if the number of shootings that occurred differed based on the month and the time of day. Finally, I look at which Borough is the most dangerous by looking at which Borough had the most shootings and the highest shooting rate.

I installed these packages for this analysis.

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

First, I will read in the csv file for the data from the website.

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

Now I will read in the data and see what it looks like.

```
shooting_data <- read_csv(url_in)</pre>
```

After looking at shooting_data, I am going to get rid of data that I don't need including INCIDENT_KEY, LOC_OF_OCCUR_DESC, PRECINCT, JURISDICTION_CODE, LOC_CLASSFCTN_DESC, LOCATION_DESC, X_COORD_CD, Y_COORD_CD, Latitude, Longitude, and Lon_Lat. I am going to rename the variables that I am keeping. The STATISTICAL_MURDER_FLAG (renamed Flag) is a True/False based on whether the shooting resulted in the victim's death which would be counted as a murder.

```
Perpetrator_Race = PERP_RACE,
Victim_Age = VIC_AGE_GROUP,
Victim_Sex = VIC_SEX,
Victim_Race = VIC_RACE)
```

I will also change the Date to date data type and the Perpetrator and Victim Age, Sex, and Race, and the Borough to the factor data type.

```
shooting_data$Date <- mdy(shooting_data$Date)
shooting_data$Perpetrator_Age <- as.factor(shooting_data$Perpetrator_Age)
shooting_data$Perpetrator_Sex <- as.factor(shooting_data$Perpetrator_Sex)
shooting_data$Perpetrator_Race <- as.factor(shooting_data$Perpetrator_Race)
shooting_data$Victim_Age <- as.factor(shooting_data$Victim_Age)
shooting_data$Victim_Sex <- as.factor(shooting_data$Victim_Sex)
shooting_data$Victim_Race <- as.factor(shooting_data$Victim_Race)
shooting_data$Boro <- as.factor(shooting_data$Boro)</pre>
```

Here is a summary of the data:

```
summary(shooting_data)
```

```
##
       Flag
                         Time
                                             Date
                                                                          Boro
##
    Mode :logical
                     Length: 27312
                                               :2006-01-01
                                                              BRONX
                                        Min.
                                                                            : 7937
                     Class1:hms
##
    FALSE: 22046
                                        1st Qu.:2009-07-18
                                                              BROOKLYN
                                                                            :10933
    TRUE :5266
                     Class2:difftime
                                        Median :2013-04-29
                                                                             : 3572
##
                                                              MANHATTAN
##
                     Mode :numeric
                                        Mean
                                               :2014-01-06
                                                              QUEENS
                                                                            : 4094
##
                                        3rd Qu.:2018-10-15
                                                              STATEN ISLAND:
                                                                              776
##
                                        Max.
                                                :2022-12-31
##
    Perpetrator_Age Perpetrator_Sex
                                            Perpetrator_Race
##
                                                                Victim_Age
   18-24 :6222
                     (null):
                              640
                                                     :11432
##
                                      BLACK
                                                              <18
                                                                      : 2839
##
    25-44
           :5687
                     F
                              424
                                      WHITE HISPANIC: 2341
                                                              1022
    UNKNOWN:3148
                                                              18-24 :10086
##
                     Μ
                           :15439
                                      UNKNOWN
                                                     : 1836
##
    <18
           :1591
                     U
                           : 1499
                                      BLACK HISPANIC: 1314
                                                              25-44
                                                                     :12281
    (null): 640
                     NA's : 9310
                                      (null)
                                                                      : 1863
##
                                                        640
                                                              45-64
    (Other): 680
##
                                      (Other)
                                                        439
                                                              65+
                                                                         181
                                                                      :
                                      NA's
##
   NA's
           :9344
                                                     : 9310
                                                              UNKNOWN:
                                                                          61
##
    Victim_Sex
                                         Victim_Race
                AMERICAN INDIAN/ALASKAN NATIVE:
##
    F: 2615
##
    M:24686
                ASIAN / PACIFIC ISLANDER
                                                   404
##
    U:
         11
               BLACK
                                                :19439
                                                : 2646
               BLACK HISPANIC
##
##
               UNKNOWN
                                                    66
##
               WHITE.
                                                   698
##
               WHITE HISPANIC
                                                : 4049
```

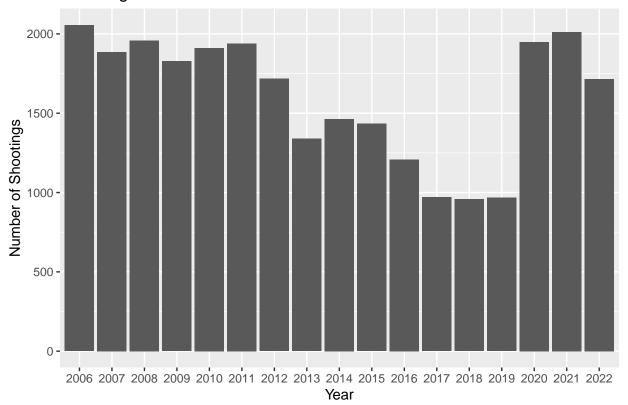
There is quite a bit of missing data in the perpetrator age, sex, and race categories, likely because in these cases it is not known who the perpetrator is. There is also some missing data in the victim age, sex, and race categories. I also noticed there was one victim age that was 1022, which I'm assuming is a typo. I'm deciding not to use these variables in any analysis since there is so much missing data.

Visualizations

I was interested to see how the number of shootings changes as time went on, so I grouped the data by year and then plotted the shootings per year.

```
shooting_data <- shooting_data %>% mutate(year = year(Date))
shootings_per_year <- shooting_data %>% group_by(year) %>% summarize(incidents_per_year = n())
ggplot(data = shootings_per_year, mapping = aes(x = factor(year), y = incidents_per_year)) +
    geom_bar(stat="identity") +
    xlab("Year") + ylab("Number of Shootings") + ggtitle("Shootings Per Year")
```

Shootings Per Year



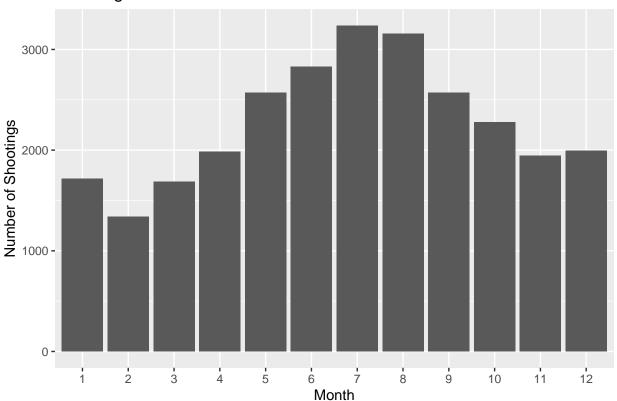
This graph showing the number of shootings per year is the opposite of what I thought it would be. It shows the number of shootings per year trending down until 2020, where the number of shootings is nearly as many as in 2006. I would have expected that criminal activity would decrease when the COVID lockdowns went into effect.

I was also interested to see how many shootings occurred each month, so I grouped the data by month and then plotted the total number of shootings per month.

```
shooting_data <- shooting_data %>% mutate(month = month(Date))
shootings_per_month <- shooting_data %>% group_by(month) %>% summarize(incidents_per_month = n())
ggplot(data = shootings_per_month, mapping = aes(x = factor(month), y = incidents_per_month)) +
```

```
geom_bar(stat = "identity") +
xlab("Month") + ylab("Number of Shootings") + ggtitle("Shootings Per Month Across All Years")
```

Shootings Per Month Across All Years

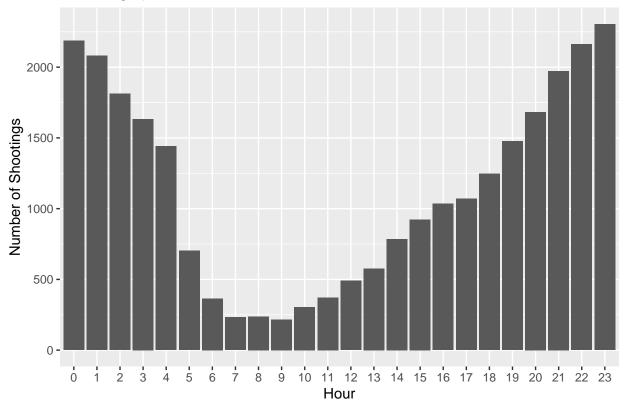


This second graph shows that across all the years of data, the most shootings occurred in the summer, and the fewest in the winter.

Now I'm curious about what time of day the most shootings occur, so I'm going to graph that.

```
shooting_data <- shooting_data %>% mutate(hour = hour(Time))
shootings_per_hour <- shooting_data %>% group_by(hour) %>% summarize(incidents_per_hour = n())
ggplot(data = shootings_per_hour, mapping = aes(x = factor(hour), y = incidents_per_hour)) +
    geom_bar(stat = "identity") +
    xlab("Hour") + ylab("Number of Shootings") + ggtitle("Shootings per Hour Across All Years")
```

Shootings per Hour Across All Years



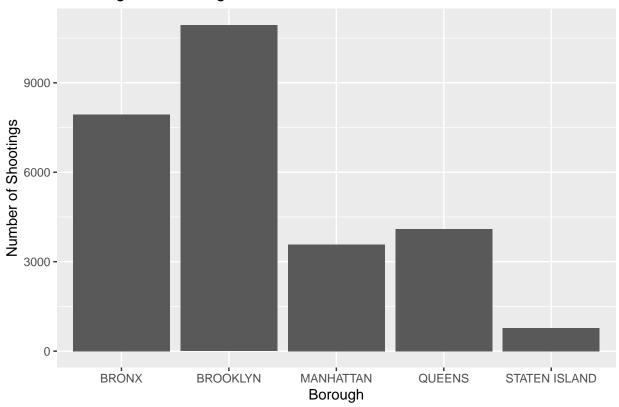
This shows that most incidents happen late at night, between the hours of 10 PM and 2 AM.

I want to plot the number of shootings that have occurred since 2006 for each Borough so I grouped the data and counted the shootings in each.

```
boro_shootings <- shooting_data %>% group_by(Boro) %>% summarize(incidents = n())

ggplot(data = boro_shootings, mapping = aes(x = Boro, y = incidents)) +
    geom_bar(stat = "identity") +
    xlab("Borough") + ylab("Number of Shootings") + ggtitle("Shootings Per Borough")
```

Shootings Per Borough



Brooklyn had the highest number of incidents, followed by the Bronx and Queens.

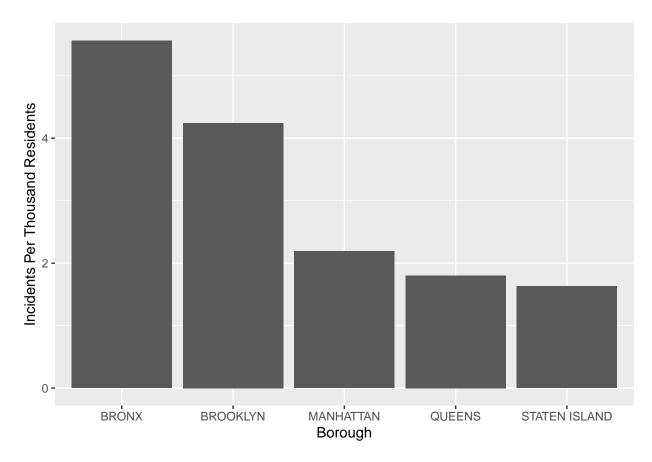
Next I added the population for each Borough.

```
shooting_data <- shooting_data %>%
mutate(Population = case_when(
   Boro == "BROOKLYN" ~ 2576771,
   Boro == "QUEENS" ~ 2270976,
   Boro == "BRONX" ~ 1427056,
   Boro == "MANHATTAN" ~ 1629153,
   Boro == "STATEN ISLAND" ~ 475596))
```

I also wanted to add the number of incidents per 1000 residents using the population for each Borough and visualize this in order to see which Borough is the most dangerous.

```
rate <- shooting_data %>%
  group_by(Boro) %>% summarize(total = n(),
  population = unique(Population)) %>%
  mutate(deaths_per_thou = total / population * 1000)

ggplot(data = rate, mapping = aes(x = Boro, y = deaths_per_thou)) +
  geom_bar(stat = "identity") +
  xlab("Borough") + ylab("Incidents Per Thousand Residents")
```



This visualization shows that even though Brooklyn had the largest number of incidents, the Bronx has the highest shooting rate per thousand residents.

Conclusion and Bias

From the analysis, we can see that the total number of shootings was on a steady decline until 2020, where the number of shootings jumped to almost the level it was at in 2006, which had the highest number of shootings of any year in this dataset. The highest number of shootings occurred in the summer months. The most shootings also occurred in the middle of the night, between the hours of 10 pm and 2 am. Brooklyn had the most shootings overall, and Staten Island had the fewest. Brooklyn also has the highest population of all the Boroughs, and Staten Island had the smallest population. However, calculating the shooting rate per 1000 residents showed us that the Bronx actually had the highest shooting rate.

This topic can bring out a lot of bias. This can start from the moment the data was collected and happen every step of the way up to and including the analysis of the data. There was quite a bit of missing data in the perpetrator age, sex, and race categories, likely because in these cases it is not known who the perpetrator is. I decided not to do any analysis involving this data because any analysis may not have been accurate since so much was missing. I was surprised that there are significantly more incidents involving male victims than female victims. Not knowing much about NYC, I was surprised that Brooklyn had the largest number of incidents overall, I would have assumed it would be the Bronx or Queens. This was an interesting dataset to look at to see that some of my first assumptions were incorrect.