

NYPD Historical Shooting Analysis

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Description of Data

The NYPD Shooting Incident Historical data lists every shooting incident that occurred in NYC going back to 2006 through 2021. This data is manually extracted every quarter and reviewed by the Office of Management Analysis and Planning before being posted on the NYPD website. Each record represents a shooting incident in NYC and includes information about the event, the location and time of occurrence. In addition, information related to suspect and victim demographics is also included.

Below is a description of all columns in the dataset:

1. INCIDENT_KEY: Randomly generated persistent ID for each incident
2. OCCUR DATE: Exact date of the shooting incident
3. OCCUR TIME: Exact time of the shooting incident
4. BORO: Borough where the shooting incident occurred
5. PRECINCT: Precinct where the shooting incident occurred
6. JURISDICTION_CODE: Jurisdiction where the shooting incident occurred. Jurisdiction codes 0(Patrol), 1(Transit), and 2(Housing) represent NYPD whilst codes 3 and more represent non NYPD jurisdictions.
7. LOCATION_DESC: Location of the shooting incident
8. STATISTICAL_MURDER_FLAG: Shooting resulted in the victim's death which would be counted as a murder
9. PERP_AGE_GROUP: Perpetrator's age within a category
10. PERP SEX: Perpetrator's sex within a category
11. PERP RACE: Perpetrator's race description
12. VIC_AGE_GROUP: Victim's age within a category
13. VIC SEX: Victim's sex description
14. VIC RACE: Victim's race description
15. X_COORD_CD: Midblock X-coordinate for New York State Plane Coordinate System, Long Island Zone, NAD83, units feet (FIPS 3104)
16. Y_COORD_CD: Midblock Y-coordinate for New York State Plane Coordinate System, Long Island Zone, NAD83, units feet (FIPS 3104)

Objective

For this analysis I answer the following questions:

1. What day of the week do shootings occur the most?
2. How many shootings are there per year? What is the trend over time?
3. What borough has the most shooting incidents? Does the borough with the most shootings change year-to-year?

Finally, I developed a linear regression model to predict murders as a function of shootings.

Import Libraries and Data

I started by importing the necessary libraries and data.

```
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.1 --

## v ggplot2 3.3.5      v purrr  0.3.4
## v tibble  3.1.6      v dplyr  1.0.7
## v tidyr   1.1.4      v stringr 1.4.0
## v readr   1.4.0      v forcats 0.5.1

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()

library(lubridate)

##
## Attaching package: 'lubridate'

## The following objects are masked from 'package:base':
##
##   date, intersect, setdiff, union

library(tidyquant)

## Loading required package: PerformanceAnalytics

## Loading required package: xts

## Loading required package: zoo

##
## Attaching package: 'zoo'

## The following objects are masked from 'package:base':
##
##   as.Date, as.Date.numeric

##
## Attaching package: 'xts'

## The following objects are masked from 'package:dplyr':
##
##   first, last

##
## Attaching package: 'PerformanceAnalytics'
```

```
## The following object is masked from 'package:graphics':
##
##     legend

## Loading required package: quantmod

## Loading required package: TTR

## Registered S3 method overwritten by 'quantmod':
##   method      from
##   as.zoo.data.frame zoo

## == Need to Learn tidyquant? =====
## Business Science offers a 1-hour course - Learning Lab #9: Performance Analysis & Portfolio Optimization
## </> Learn more at: https://university.business-science.io/p/learning-labs-pro </>
```

```
NYPD_shooting_tbl <- read_csv("NYPD_Shooting_Incident_Data__Historic_.csv")
```

```
## Warning in gzfile(file, mode): cannot open compressed file 'C:/Users/qvall/
## AppData/Local/Temp/RtmpcPqhBT\filea9e04be951c7', probable reason 'No such file
## or directory'
```

```
##
## -- Column specification -----
## cols(
##   INCIDENT_KEY = col_double(),
##   OCCUR_DATE = col_character(),
##   OCCUR_TIME = col_time(format = ""),
##   BORO = col_character(),
##   PRECINCT = col_double(),
##   JURISDICTION_CODE = col_double(),
##   LOCATION_DESC = col_character(),
##   STATISTICAL_MURDER_FLAG = col_logical(),
##   PERP_AGE_GROUP = col_character(),
##   PERP_SEX = col_character(),
##   PERP_RACE = col_character(),
##   VIC_AGE_GROUP = col_character(),
##   VIC_SEX = col_character(),
##   VIC_RACE = col_character(),
##   X_COORD_CD = col_double(),
##   Y_COORD_CD = col_double(),
##   Latitude = col_double(),
##   Longitude = col_double(),
##   Lon_Lat = col_character()
## )
```

Examine Data

I then examined the data for any import issues or missing values. There were no issues reading in the file. However, there were five columns with missing data.

- JURISDICTION_CODE
- LOCATION_DESC
- PERP_AGE_GROUP
- PERP_SEX
- PERP_RACE

There are the same number of observations (9,310) missing from the PERP_SEX and PERP_RACE columns. The first bias that came to mind was that these are probably shooting incidents when a suspect was unidentifiable resulting in an unsolved case. However, I don't have the evidence to prove that. Additionally, those fields are not needed for this analysis.

```
# Check for problems reading in the file
readr::problems(NYPD_shooting_tbl)
```

```
## [1] row      col      expected actual
## <0 rows> (or 0-length row.names)
```

```
# High-level view
NYPD_shooting_tbl %>% glimpse()
```

```
## Rows: 25,596
## Columns: 19
## $ INCIDENT_KEY      <dbl> 24050482, 77673979, 226950018, 237710987, 2247~
## $ OCCUR_DATE        <chr> "08/27/2006", "03/11/2011", "04/14/2021", "12/~
## $ OCCUR_TIME        <time> 05:35:00, 12:03:00, 21:08:00, 19:30:00, 00:18~
## $ BORO              <chr> "BRONX", "QUEENS", "BRONX", "BRONX", "MANHATTA~
## $ PRECINCT          <dbl> 52, 106, 42, 52, 34, 75, 32, 26, 41, 67, 43, 6~
## $ JURISDICTION_CODE <dbl> 0, 0, 0, 0, 0, 0, 0, 2, 2, 0, 0, 0, 0, 0, 0~
## $ LOCATION_DESC     <chr> NA, NA, "COMMERCIAL BLDG", NA, NA, NA, NA, "MU~
## $ STATISTICAL_MURDER_FLAG <lgl> TRUE, FALSE, TRUE, FALSE, FALSE, TRUE, FALSE, ~
## $ PERP_AGE_GROUP    <chr> NA, NA, NA, NA, NA, "25-44", "25-44", NA, "25--
## $ PERP_SEX          <chr> NA, NA, NA, NA, NA, "M", "M", NA, "M", NA, NA,~
## $ PERP_RACE         <chr> NA, NA, NA, NA, NA, "BLACK HISPANIC", "BLACK",~
## $ VIC_AGE_GROUP     <chr> "25-44", "65+", "18-24", "25-44", "25-44", "25~
## $ VIC_SEX          <chr> "F", "M", "M", "M", "M", "M", "M", "M", "M", "~
## $ VIC_RACE         <chr> "BLACK HISPANIC", "WHITE", "BLACK", "BLACK", "~
## $ X_COORD_CD       <dbl> 1017542, 1027543, 1009489, 1017440, 1005426, 1~
## $ Y_COORD_CD       <dbl> 255918.9, 186095.0, 243050.0, 256046.0, 254690~
## $ Latitude         <dbl> 40.86906, 40.67737, 40.83376, 40.86941, 40.865~
## $ Longitude        <dbl> -73.87963, -73.84392, -73.90880, -73.88000, -7~
## $ Lon_Lat          <chr> "POINT (-73.87963173099996 40.86905819000003)"~
```

```
# Check for empty fields
find_NAs <- colSums(is.na(NYPD_shooting_tbl))
find_NAs
```

```
##      INCIDENT_KEY      OCCUR_DATE      OCCUR_TIME
##      0                0                0
##      BORO              PRECINCT        JURISDICTION_CODE
##      0                0                2
##      LOCATION_DESC STATISTICAL_MURDER_FLAG PERP_AGE_GROUP
##      14977          0                9344
```

```
##          PERP_SEX          PERP_RACE          VIC_AGE_GROUP
##          9310          9310          0
##          VIC_SEX          VIC_RACE          X_COORD_CD
##          0          0          0
##          Y_COORD_CD          Latitude          Longitude
##          0          0          0
##          Lon_Lat
##          0
```

Tidy and Transform

```
# convert occur_date from character to date field
NYPD_shooting_tbl <- NYPD_shooting_tbl %>%
  mutate(OCCUR_DATE= as.Date(OCCUR_DATE, format = "%m/%d/%Y")) %>%
  mutate(year_of_shooting = year(OCCUR_DATE)) %>%

# Remove columns with missing data and select fields needed to answer my questions
select(-JURISDICTION_CODE, -LOCATION_DESC, -PERP_SEX,
       -PERP_AGE_GROUP, -PERP_RACE) %>% glimpse()
```

```
## Rows: 25,596
## Columns: 15
## $ INCIDENT_KEY          <dbl> 24050482, 77673979, 226950018, 237710987, 2247~
## $ OCCUR_DATE            <date> 2006-08-27, 2011-03-11, 2021-04-14, 2021-12-1~
## $ OCCUR_TIME            <time> 05:35:00, 12:03:00, 21:08:00, 19:30:00, 00:18~
## $ BORO                  <chr> "BRONX", "QUEENS", "BRONX", "BRONX", "MANHATTA~
## $ PRECINCT              <dbl> 52, 106, 42, 52, 34, 75, 32, 26, 41, 67, 43, 6~
## $ STATISTICAL_MURDER_FLAG <lgl> TRUE, FALSE, TRUE, FALSE, FALSE, TRUE, FALSE, ~
## $ VIC_AGE_GROUP         <chr> "25-44", "65+", "18-24", "25-44", "25-44", "25~
## $ VIC_SEX               <chr> "F", "M", "M", "M", "M", "M", "M", "M", "M", "~
## $ VIC_RACE              <chr> "BLACK HISPANIC", "WHITE", "BLACK", "BLACK", "~
## $ X_COORD_CD            <dbl> 1017542, 1027543, 1009489, 1017440, 1005426, 1~
## $ Y_COORD_CD            <dbl> 255918.9, 186095.0, 243050.0, 256046.0, 254690~
## $ Latitude              <dbl> 40.86906, 40.67737, 40.83376, 40.86941, 40.865~
## $ Longitude             <dbl> -73.87963, -73.84392, -73.90880, -73.88000, -7~
## $ Lon_Lat               <chr> "POINT (-73.87963173099996 40.86905819000003)"~
## $ year_of_shooting      <dbl> 2006, 2011, 2021, 2021, 2021, 2021, 2021, 2021~
```

```
# Make sure all missing data has been resolved
find_NAs <- colSums(is.na(NYPD_shooting_tbl))
find_NAs
```

```
##          INCIDENT_KEY          OCCUR_DATE          OCCUR_TIME
##          0          0          0
##          BORO          PRECINCT STATISTICAL_MURDER_FLAG
##          0          0          0
##          VIC_AGE_GROUP          VIC_SEX          VIC_RACE
##          0          0          0
##          X_COORD_CD          Y_COORD_CD          Latitude
##          0          0          0
##          Longitude          Lon_Lat          year_of_shooting
##          0          0          0
```

Analysis and Visualization

What day of the week do shootings occur the most?

Most shootings happened on the weekend with Sunday and Saturday accounting for 20 and 19 percent, respectively, of all shootings throughout the week.

```
day_shooting_tbl <- NYPD_shooting_tbl %>%
  select(OCCUR_DATE, year_of_shooting) %>%
  mutate(day_of_week = wday(OCCUR_DATE, label = TRUE, abbr = TRUE)) %>%
  mutate(count = 1)

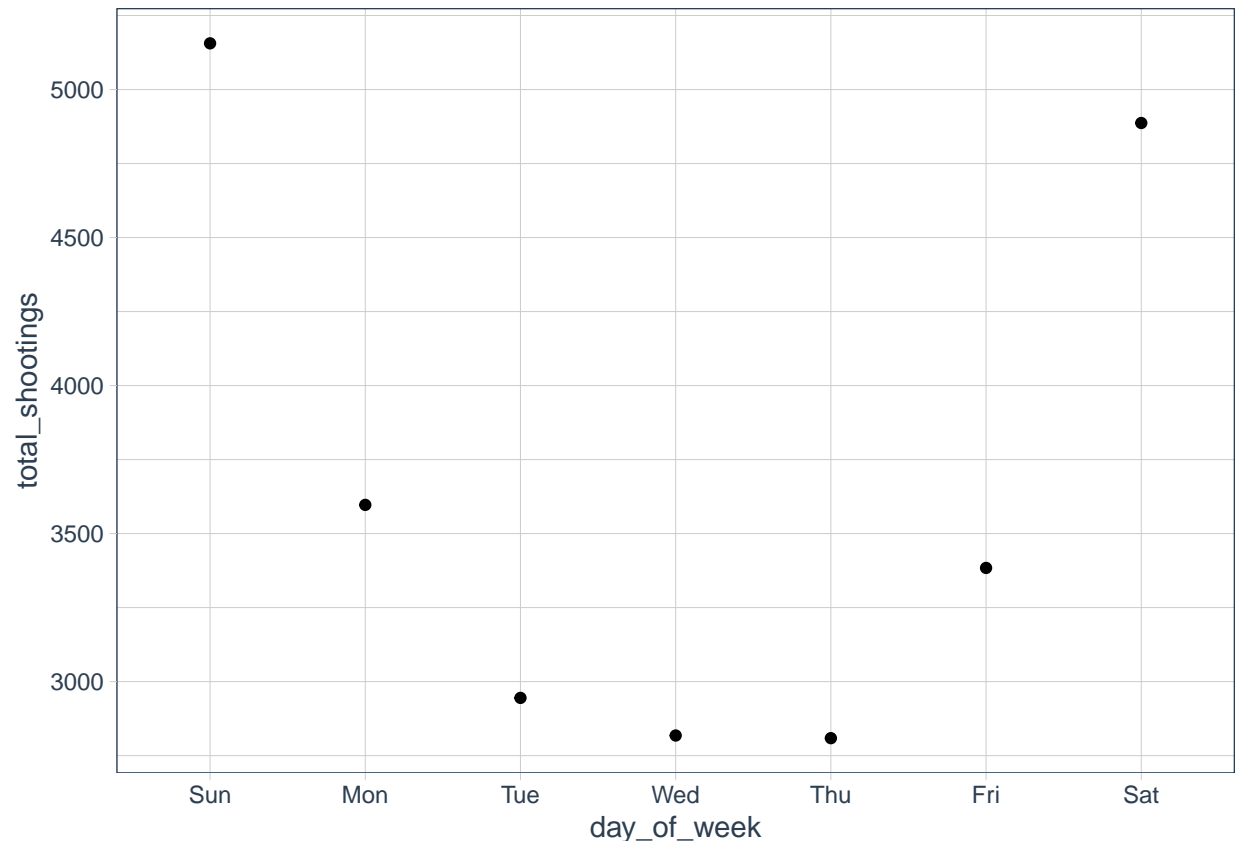
day_shooting_summary_tbl <- day_shooting_tbl %>%
  group_by(day_of_week) %>%
  summarize(total_shootings = sum(count)) %>%
  arrange(desc(total_shootings)) %>%
  mutate(pct = total_shootings / sum(total_shootings)) %>%
  mutate(pct = scales::percent(pct, accuracy = 1.)) %>%
  ungroup()

day_shooting_summary_tbl
```

```
## # A tibble: 7 x 3
##   day_of_week total_shootings pct
##   <ord>          <dbl> <chr>
## 1 Sun              5156 20%
## 2 Sat              4887 19%
## 3 Mon              3597 14%
## 4 Fri              3384 13%
## 5 Tue              2945 12%
## 6 Wed              2818 11%
## 7 Thu              2809 11%
```

```
day_shooting_plot <- day_shooting_summary_tbl %>%
  ggplot(aes(x = day_of_week, y = total_shootings)) +
  geom_point() +
  theme_tq()

day_shooting_plot
```



How many shootings are there per year? What is the trend over time?

The number of shootings per year ranged from 967 to 2,055 over the 15-year period. Shootings had been steadily declining from 2006, with big drops in 2011 through 2013 and 2015 through 2017. However, shootings increased sharply after 2019, and were back to 2006 levels by 2021.

After seeing this trend, I was curious whether the COVID-19 pandemic had something to do with the increased shootings. However, I didn't go further with this line of thinking because that would have required gathering additional data sources.

```
shooting_by_yr_tbl <- NYPD_shooting_tbl %>%
  select(OCCUR_DATE, year_of_shooting) %>%
  mutate(count = 1) %>%

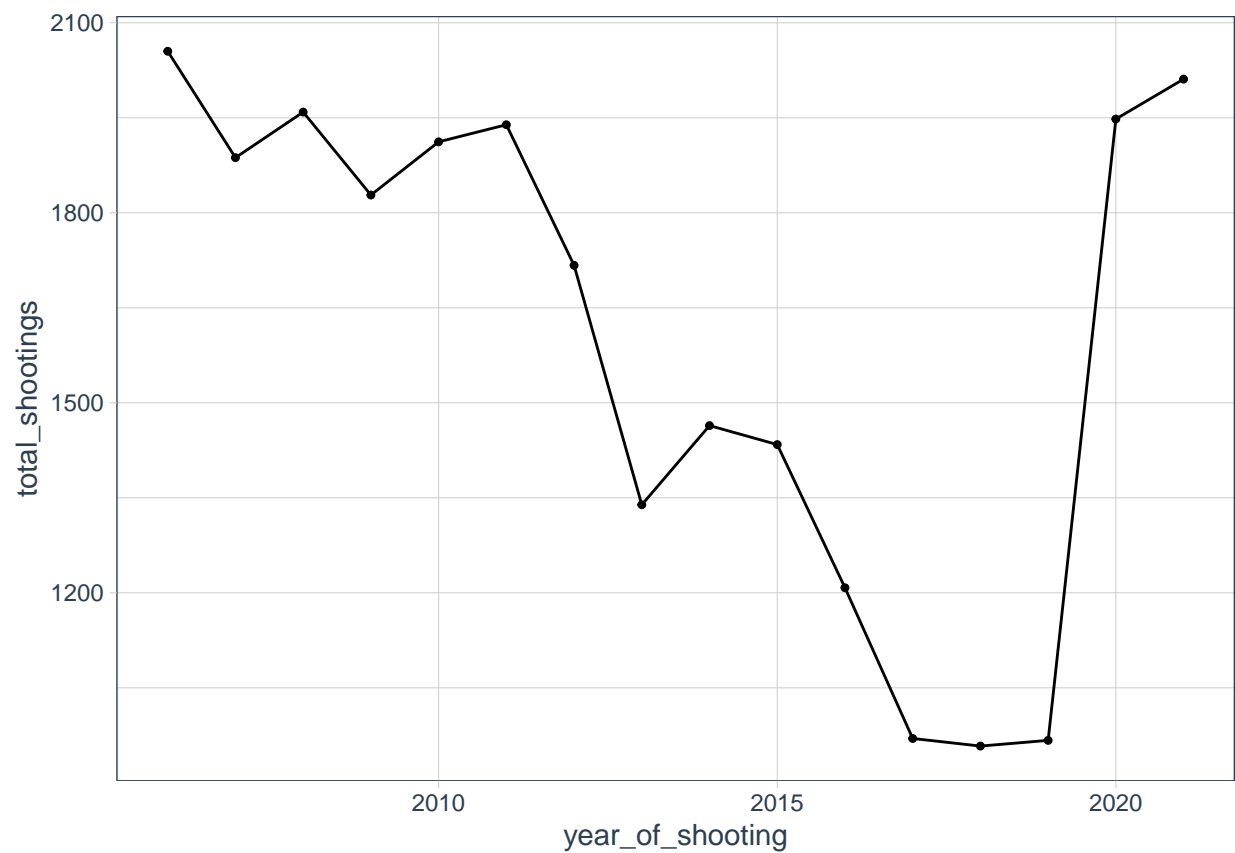
  group_by(year_of_shooting) %>%
  summarize(total_shootings = sum(count)) %>%
  ungroup()

shooting_by_yr_plot <- shooting_by_yr_tbl %>%
  ggplot(aes(x = year_of_shooting, y = total_shootings)) +
  geom_line() +
  geom_point(shape = 20) +
  theme_tq()

shooting_by_yr_tbl
```

```
## # A tibble: 16 x 2
##   year_of_shooting total_shootings
##   <dbl>           <dbl>
## 1      2006           2055
## 2      2007           1887
## 3      2008           1959
## 4      2009           1828
## 5      2010           1912
## 6      2011           1939
## 7      2012           1717
## 8      2013           1339
## 9      2014           1464
## 10     2015           1434
## 11     2016           1208
## 12     2017           970
## 13     2018           958
## 14     2019           967
## 15     2020           1948
## 16     2021           2011
```

shooting_by_yr_plot



What borough has the most shooting incidents? Does the borough with the most shootings change year-to-year?

Forty percent of all shootings over the 15-year period occurred in Brooklyn. I was curious if this was the case for all years and found that most shooting incidents occurred in Brooklyn except for the year 2021. Most shootings occurred in the Bronx in 2021. Lastly, I created box plots for each borough to see the distribution of shootings by borough.

```
boro_shooting_tbl <- NYPD_shooting_tbl %>%
  select(BORO, PRECINCT, year_of_shooting) %>%
  mutate(count = 1)

boro_shooting_summary_tbl <- boro_shooting_tbl %>%
  group_by(BORO) %>%
  summarize(total_shootings = sum(count)) %>%
  arrange(desc(total_shootings)) %>%
  mutate(pct = total_shootings / sum(total_shootings)) %>%
  mutate(pct = scales::percent(pct, accuracy = 1.)) %>%
  ungroup()

boro_shooting_by_yr_tbl <- boro_shooting_tbl %>%
  select(BORO, year_of_shooting, count) %>%
  group_by(year_of_shooting, BORO) %>%
  summarize(total_shootings = sum(count)) %>%
  ungroup()
```

'summarise()' has grouped output by 'year_of_shooting'. You can override using the '.groups' argument

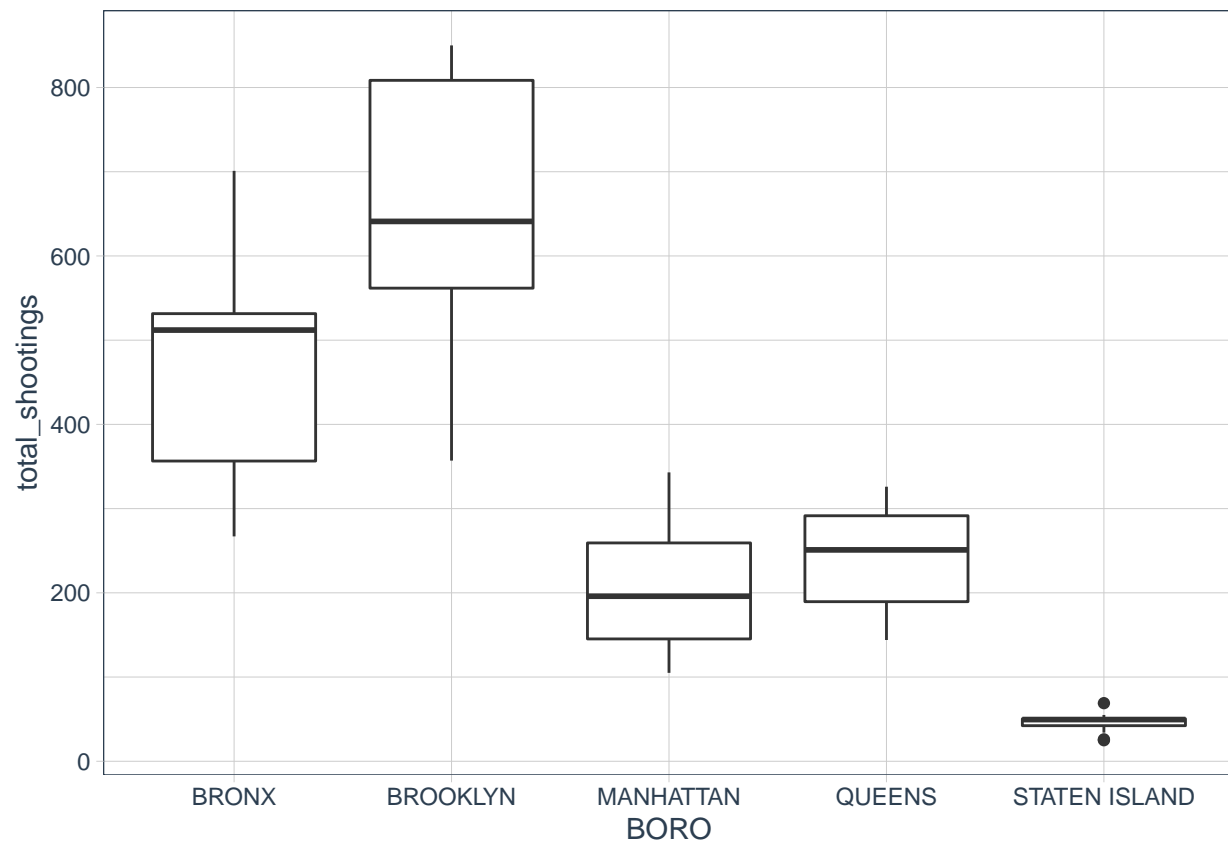
```
boro_shooting_by_yr_spread <- boro_shooting_by_yr_tbl %>%
  spread(key = BORO, value = total_shootings)

boro_shooting_by_yr_spread$row_max <- apply(boro_shooting_by_yr_spread[, -1], 1, max)
boro_shooting_by_yr_spread
```

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

```
boro_shooting_by_yr_plot <- boro_shooting_by_yr_tbl %>%
  group_by(year_of_shooting) %>%
  ggplot(aes(x = BORO, y = total_shootings)) +
  geom_boxplot() +
  theme_tq()
```

```
boro_shooting_by_yr_plot
```



Modeling

For my model, I chose to predict the number of murders as a function of shootings. The model does a pretty good job at predicting murders illustrating that the number of shooting is a pretty good indicator for murders.

```
# create total_num_yrs column
NYPD_shooting_tbl <- NYPD_shooting_tbl %>%
  mutate(total_num_yrs = max(year_of_shooting) - min(year_of_shooting))

# create shooting tbl
shooting_by_yr_tbl <- NYPD_shooting_tbl %>%
  select(OCCUR_DATE, year_of_shooting, total_num_yrs) %>%
  mutate(count = 1) %>%

  group_by(year_of_shooting, total_num_yrs) %>%
```

```
summarize(total_shootings = sum(count)) %>%
ungroup()
```

'summarise()' has grouped output by 'year_of_shooting'. You can override using the '.groups' argument

```
# create deaths tibble
deaths_by_yr_tbl <- NYPD_shooting_tbl %>%
  select(OCCUR_DATE, year_of_shooting, total_num_yrs, STATISTICAL_MURDER_FLAG) %>%
  mutate(count = 1) %>%
  mutate(murder_count = case_when(
    STATISTICAL_MURDER_FLAG == TRUE ~ 1,
    STATISTICAL_MURDER_FLAG == FALSE ~ 0)) %>%

  group_by(year_of_shooting, total_num_yrs) %>%
  summarize(total_murders = sum(murder_count)) %>%
  ungroup()
```

'summarise()' has grouped output by 'year_of_shooting'. You can override using the '.groups' argument

```
# join tibbles
joined_tbl <- left_join(shooting_by_yr_tbl, deaths_by_yr_tbl,
  by = c("year_of_shooting" = "year_of_shooting"))

joined_tbl <- joined_tbl %>%
  select(-total_num_yrs.y)

joined_tbl
```

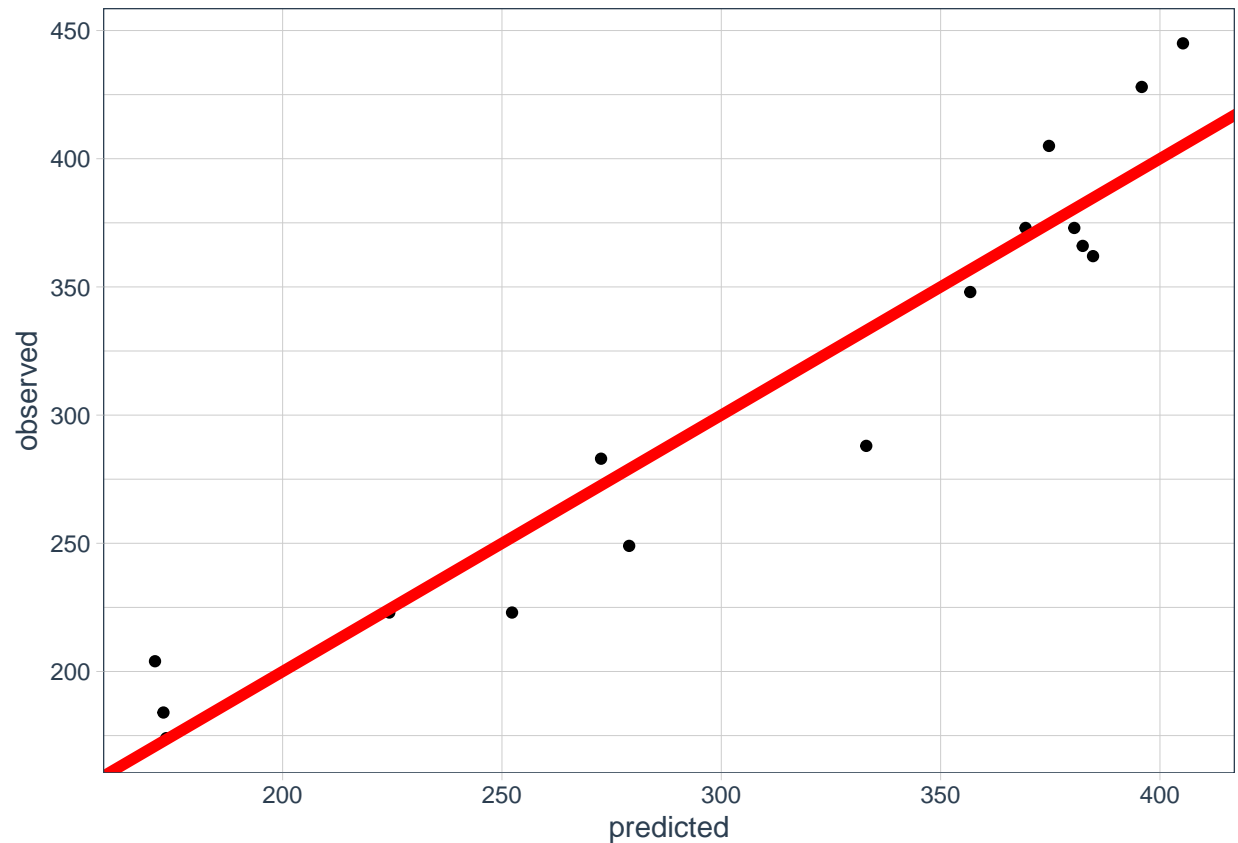
```
## # A tibble: 16 x 4
##   year_of_shooting total_num_yrs.x total_shootings total_murders
##   <dbl> <dbl> <dbl> <dbl>
## 1 2006 15 2055 445
## 2 2007 15 1887 373
## 3 2008 15 1959 362
## 4 2009 15 1828 348
## 5 2010 15 1912 405
## 6 2011 15 1939 373
## 7 2012 15 1717 288
## 8 2013 15 1339 223
## 9 2014 15 1464 249
## 10 2015 15 1434 283
## 11 2016 15 1208 223
## 12 2017 15 970 174
## 13 2018 15 958 204
## 14 2019 15 967 184
## 15 2020 15 1948 366
## 16 2021 15 2011 428
```

```
# my model: estimate linear regression
my_mod <- lm(total_murders ~ total_shootings, joined_tbl)
summary(my_mod)
```

```
##
## Call:
## lm(formula = total_murders ~ total_shootings, data = joined_tbl)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -45.047 -17.980  -0.395  15.950  39.750
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -33.73553    27.45209  -1.229   0.239
## total_shootings  0.21362     0.01667  12.817 4e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 26.16 on 14 degrees of freedom
## Multiple R-squared:  0.9215, Adjusted R-squared:  0.9159
## F-statistic: 164.3 on 1 and 14 DF,  p-value: 3.996e-09
```

```
# create data for ggplot
data_mod_tbl <- tibble(predicted = predict(my_mod),
                        observed = joined_tbl$total_murders)

# create plot
ggplot(data_mod_tbl,
       aes(x = predicted,
           y = observed)) +
  geom_point() +
  geom_abline(intercept = 0,
              slope = 1,
              color = "red",
              size = 2) +
  theme_tq()
```



Bias

A bias that I had going into the project was that borough with the highest rate of poverty in NYC would also have the highest number of shootings. I also assumed that because the cost of living in Manhattan is astronomical, that it would have the least amount of shootings. I was wrong about both!

Brooklyn had the most shootings, which has less poverty than the Bronx, and Staten Island had the least shootings, not Manhattan. However, I avoided confirmation bias just by letting the data speak for itself throughout my analysis. I did not modify any observations or cherry pick data.