NY Shootings Analysis Post Great Recession - Covid

Thales

2023-05-10

In this document, I will delve into the shootings that occurred in New York City from the post-Great Recession era (starting in 2010) up until the first two years of the Covid-19 Pandemic (2020-2022). Our goal is to analyze governmental datasets and identify potential factors that may have influenced these incidents. By doing so, we hope to gain a better understanding of the root causes of these shootings.

Fetch Reports

##

##

##

summarize

```
library(tidyverse)
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr
         1.1.2
                   v readr
                              2.1.4
## v forcats
            1.0.0
                    v stringr
                              1.5.0
## v ggplot2
            3.4.2
                              3.2.1
                   v tibble
## v lubridate 1.9.2
                    v tidyr
                              1.3.0
## v purrr
            1.0.1
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                 masks stats::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
library(plyr)
## ------
## You have loaded plyr after dplyr - this is likely to cause problems.
## If you need functions from both plyr and dplyr, please load plyr first, then dplyr:
## library(plyr); library(dplyr)
## ------
##
## Attaching package: 'plyr'
```

arrange, count, desc, failwith, id, mutate, rename, summarise,

The following objects are masked from 'package:dplyr':

```
## The following object is masked from 'package:purrr':
##
##
      compact
nypd report url <- "https://data.cityofnewyork.us/api/views/833y-fsy8/rows.csv?accessType=DOWNLOAD"</pre>
ny pop report url <- "https://data.ny.gov/api/views/krt9-ym2k/rows.csv?accessType=DOWNLOAD&sorting=true
raw_nypd_report = read_csv(nypd_report_url)
## Rows: 27312 Columns: 21
## -- Column specification -----
## Delimiter: ","
       (12): OCCUR_DATE, BORO, LOC_OF_OCCUR_DESC, LOC_CLASSFCTN_DESC, LOCATION...
        (7): INCIDENT_KEY, PRECINCT, JURISDICTION_CODE, X_COORD_CD, Y_COORD_CD...
## dbl
        (1): STATISTICAL_MURDER_FLAG
## time (1): OCCUR_TIME
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
raw_ny_pop_report <- read_csv(ny_pop_report_url)</pre>
## Rows: 3717 Columns: 5
## -- Column specification -------
## Delimiter: ","
## chr (2): Geography, Program Type
## dbl (3): FIPS Code, Year, Population
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

Apply necessary tranformations for further Analysis

```
Geography == "Queens County" ~ "QUEENS",
    Geography == "Richmond County" ~ "STATEN ISLAND",
    )
) %>%
select(-"FIPS Code", -"Program Type", -"Geography") %>%
arrange(Year)
```

Summarize Data for periods of interest

```
incidents_10_19 <- report %>%
   filter(Year >= 2010 & Year < 2020) %>%
    group_by(Year) %>%
    dplyr::summarize(total_cases = n())
incidents_10_22 <- report \%>%
   filter(Year >= 2010 & Year < 2023) %>%
   group_by(Year) %>%
   dplyr::summarize(total_cases = n())
incidents_per_boro_10_19 <- report %>%
   filter(lubridate::year(OCCUR_DATE) < 2020) %>%
   group_by(BORO, Year) %>%
   dplyr::summarize(`Total Cases` = n())
## 'summarise()' has grouped output by 'BORO'. You can override using the
## '.groups' argument.
incidents_per_boro_10_22 <- report %>%
   filter(lubridate::year(OCCUR_DATE) >= 2010) %>%
   group_by(BORO, Year) %>%
   dplyr::summarize(`Total Cases` = n())
## 'summarise()' has grouped output by 'BORO'. You can override using the
## '.groups' argument.
pop_report_10_22 <- merge(</pre>
                      incidents_per_boro_10_22,
                      ny_pop_report,
                      by=c("BORO","Year"),
                      all.x=TRUE
                      )
pop_report_10_19 <- merge(</pre>
                      incidents_per_boro_10_19,
                      ny_pop_report,
                      by=c("BORO","Year"),
                      all.x=TRUE
```

Merge summarised data with dataset containg population and adding calculated columns

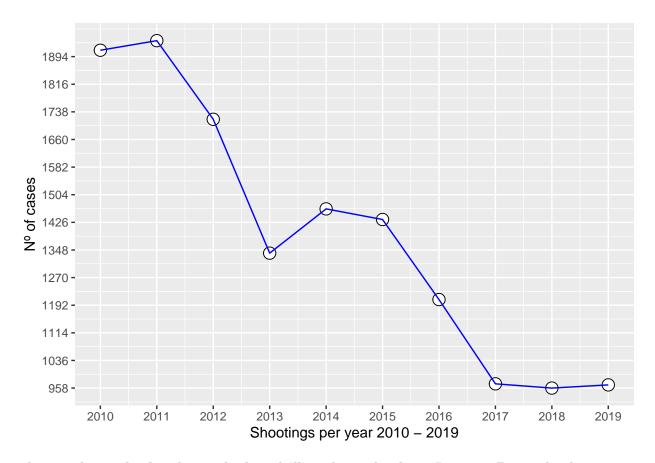
```
deaths_per_boro_10_19 <- report %>%
                  filter(Year < 2020 & STATISTICAL_MURDER_FLAG == TRUE) %>%
                  group_by(BORO, Year) %>%
                  dplyr::summarize(Deaths = n()) %>%
                  arrange(BORO, Year)
## 'summarise()' has grouped output by 'BORO'. You can override using the
## '.groups' argument.
deaths_per_boro_10_22 <- report %>%
                  filter(Year >= 2010 & STATISTICAL_MURDER_FLAG == TRUE) %>%
                  group_by(BORO, Year) %>%
                  dplyr::summarize(Deaths = n()) %>%
                  arrange(BORO, Year)
## 'summarise()' has grouped output by 'BORO'. You can override using the
## '.groups' argument.
full_trend_report_10_19 <- merge(</pre>
                            pop report 10 19,
                            deaths_per_boro_10_19,
                            by=c("BORO","Year"),
                            all.x=TRUE
full_trend_report_10_22 <- merge(</pre>
                            pop_report_10_22,
                            deaths_per_boro_10_22,
                            by=c("BORO","Year"),
                            all.x=TRUE
                            )
pop_report_10_19 <- full_trend_report_10_19 %>%
    mutate("Incidents per 100k residents " = (`Total Cases`/Population)*10.0**4) %%
    mutate("Deaths per 100k residents " = (Deaths/Population)*10.0**4)
pop_report_10_19 <- pop_report_10_19 %>%
  mutate('Incidents per 100k residents ' = round('Incidents per 100k residents ', digits = 2)) %%
  mutate(`Deaths per 100k residents ` = round(`Deaths per 100k residents `, digits = 2))
pop_report_10_22 <- full_trend_report_10_22 %>%
    mutate("Incidents per 100k residents " = (`Total Cases`/Population)*10.0**4) %>%
    mutate("Deaths per 100k residents " = (Deaths/Population)*10.0**4)
pop_report_10_22 <- pop_report_10_22 %>%
```

```
mutate(`Incidents per 100k residents ` = round(`Incidents per 100k residents `, digits = 2)) %>%
mutate(`Deaths per 100k residents ` = round(`Deaths per 100k residents `, digits = 2))
```

Shootings trend 2010 - 2019

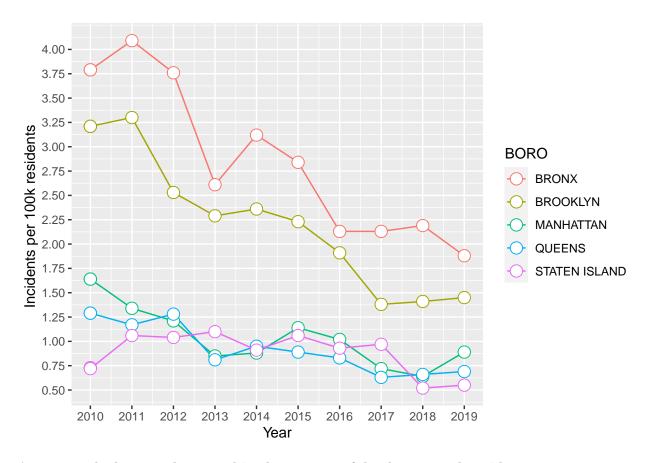
We can observe a downward trend in the number of shootings in the period after The Great Recession Until 2019 (before the Covid Pandemic Crisis)

```
ggplot(data = incidents_10_19, aes(x = Year, y = total_cases)) +
 geom_point( size=4, shape=21, fill="white") + geom_line(color = "blue") +
 xlab("Year") +
 ylab("N^{\circ} of cases") +
 ggtitle("Shootings per year 2010 - 2019") %>%
  scale_x_continuous(
   breaks=seq(
      min(incidents_10_19$Year),
      max(incidents_10_19$Year),
      by=1
    ) +
  scale_y_continuous(
    breaks=seq(
      min(incidents_10_19$total_cases)
      ,max(incidents_10_19$total_cases)
     by=78
     )
```

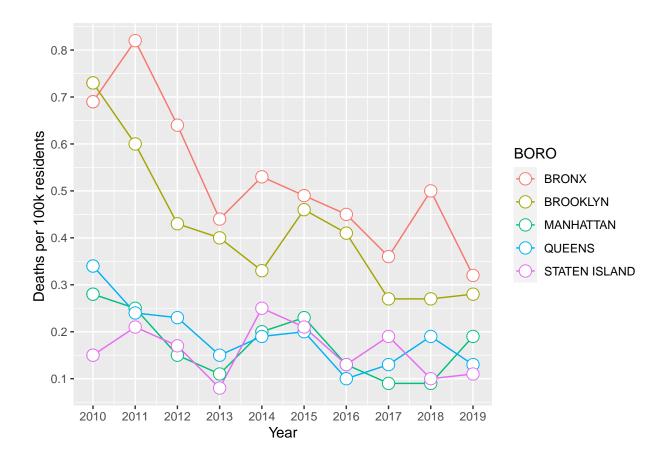


The trend can also be observed when drilling down the data. In every Borough, the amount of incidents per 100k residents decreased

```
ggplot(data=pop_report_10_19,aes(x=Year,y=`Incidents per 100k residents `,group=BORO,colour=BORO)) +
    geom_line() +
    geom_point( size=4, shape=21, fill="white") +
    scale_x_continuous(
        breaks=seq(
            min(pop_report_10_19$Year),
            max(pop_report_10_19$Year),
            by=1
            )
            ) +
        scale_y_continuous(
        breaks=seq(
            0,
            max(pop_report_10_19$`Incidents per 100k residents `),
            by=0.25
            )
        )
}
```



As expected, the same happened in the amount of deaths per 100k residents

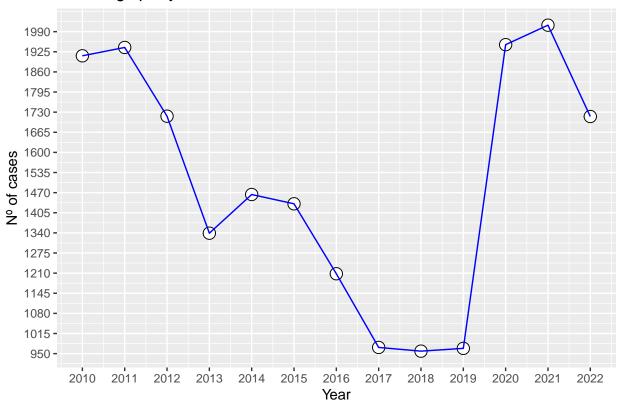


Upward trend in shootings starting in 2020

During the Pandemic, we can observe a spike in the number of cases

```
ggplot(data = incidents_10_22, aes(x = Year, y = total_cases)) +
  geom_point(size=4, shape=21, fill="white") + geom_line(color = "blue") +
  xlab("Year") +
  ylab("N^{\circ} of cases") +
  ggtitle("Shootings per year 2010 - 2022") +
  scale_x_continuous(
    breaks=seq(
      min(incidents_10_22$Year),
      max(incidents_10_22$Year),
      by=1
      )
    ) +
  scale_y_continuous(
    breaks=seq(
      950,
      max(incidents_10_22$total_cases),
      by = 65
      )
```

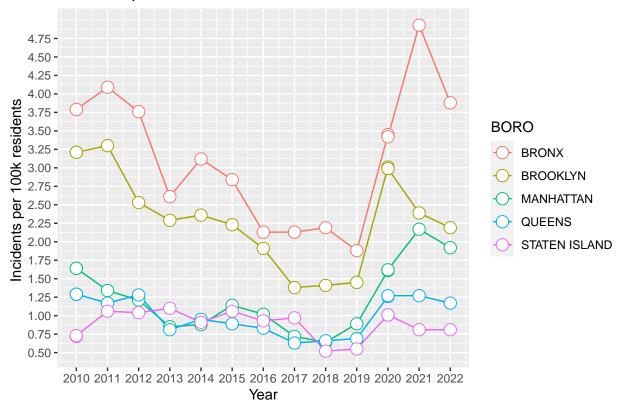
Shootings per year 2010 – 2022



Also observed in every Borough in NYC

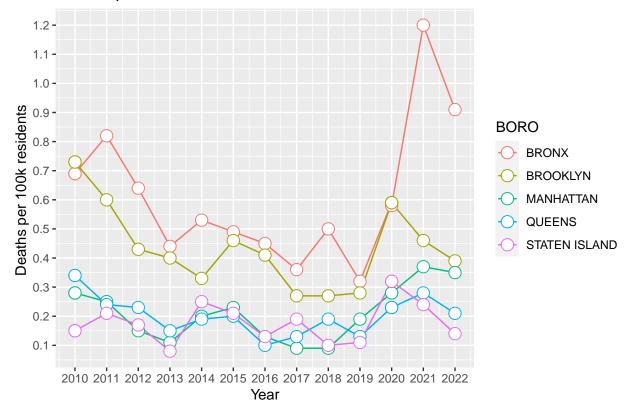
```
ggplot(data=pop_report_10_22,aes(x=Year,y=`Incidents per 100k residents `,group=BORO,colour=BORO)) +
   geom_line() +
   geom_point( size=4, shape=21, fill="white") +
   ggtitle("Incidents per 100k residents 2010 - 2022") +
   scale_x_continuous(
     breaks=seq(
       min(pop_report_10_22$Year),
       max(pop_report_10_22$Year),
       by=1
        )
      ) +
   scale_y_continuous(
      breaks=seq(
        max(pop_report_10_22$`Incidents per 100k residents `),
        by=0.25
        )
      )
```

Incidents per 100k residents 2010 – 2022



```
ggplot(data=pop_report_10_22, aes(x=Year, y=`Deaths per 100k residents `, group = BORO, colour = BORO))
  geom_line() +
  geom_point( size=4, shape=21, fill="white") +
  ggtitle("Deaths per 100k residents 2010 - 2022") +
  scale_x_continuous(breaks = seq(min(pop_report_10_22$Year), max(pop_report_10_22$Year), by = 1)) +
  scale_y_continuous(breaks = seq(0, max(pop_report_10_22$`Deaths per 100k residents `), by = 0.1))
```

Deaths per 100k residents 2010 – 2022



The unemployment rate could be a factor that contributes to the increase in gun violence. We'll import a New York State Dataset and analyze if there is a relationship for further discussion

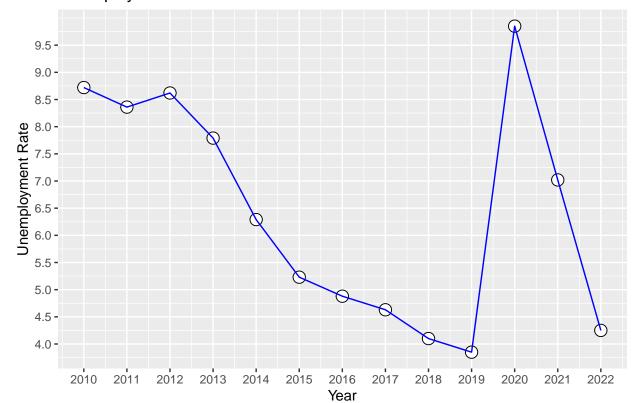
Importing NY unemployment dataset

```
summarise_at(vars(NYUR), list(`Unemployment Rate` = mean)) %>%
mutate(`Unemployment Rate` = round(`Unemployment Rate`, digits = 2))
```

Analyzing unemployment rate over the same period

```
ggplot(data = unempleyment_yearly_summary %% filter(Year >= 2010),aes(x=Year,y =`Unemployment Rate`))
  geom_point( size=4, shape=21, fill="white") + geom_line(color = "blue") +
  xlab("Year") +
  ylab("Unemployment Rate") +
  ggtitle("Unemployment Rate 2010 - 2022") +
  scale_x_continuous(
   breaks=seq(
     min(unempleyment_yearly_summary$Year),
     max(unempleyment_yearly_summary$Year),
     by=1
      )
   ) +
  scale_y_continuous(
   breaks=seq(
     max(unempleyment_yearly_summary$`Unemployment Rate`),
      by=0.5
      )
```

Unemployment Rate 2010 - 2022

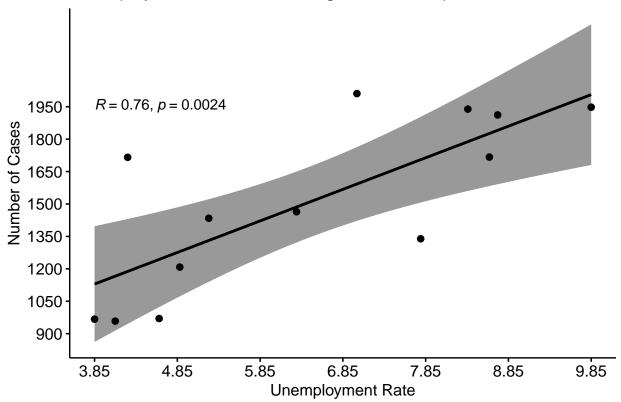


Merging Unenmployment and Shooting Incidents Datasets

Pearson correlation Analysis

```
library("ggpubr")
##
## Attaching package: 'ggpubr'
## The following object is masked from 'package:plyr':
##
##
       mutate
ggscatter(shootings_unemployment_set, x = "Unemployment Rate", y = "total_cases",
          add = "reg.line", conf.int = TRUE,
          cor.coef = TRUE, cor.method = "pearson",
          xlab = "Unemployment Rate", ylab = "Number of Cases") +
          ggtitle("Unemployment Rate & Shootings Relationship") +
          scale_x_continuous(
            breaks=seq(
              min(shootings_unemployment_set$`Unemployment Rate`),
              max(shootings_unemployment_set$`Unemployment Rate`),
              by=1
            ) +
          scale_y_continuous(
            breaks=seq(
              max(shootings_unemployment_set$total_cases),
              by=150
              )
```





With a correlation of 0.76 and a P value of 0.0024, the data suggests that the unemployment could be related to the Number of Shootings in NYC.

Linear regression Analysis

col = "red".

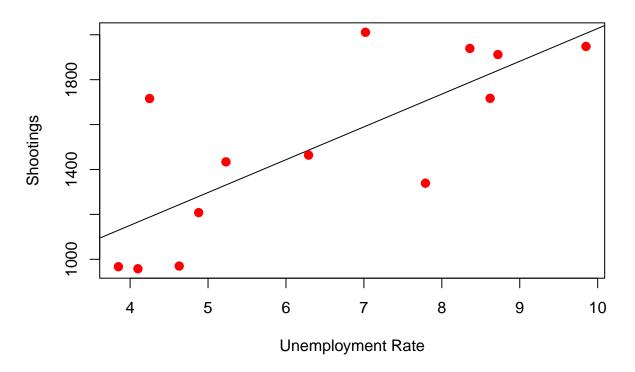
abline(

main = "Unemployment Rate & Shootings Regression",

```
relation <- lm(shootings_unemployment_set$`Unemployment Rate`~shootings_unemployment_set$total_cases)
relation
##
## Call:
   lm(formula = shootings_unemployment_set$'Unemployment Rate' ~
       shootings_unemployment_set$total_cases)
##
##
##
   Coefficients:
##
                               (Intercept)
                                             {\tt shootings\_unemployment\_set\$total\_cases}
                                  0.431318
                                                                            0.003982
##
# Plot the chart.
plot(
  shootings_unemployment_set$`Unemployment Rate`,
  shootings_unemployment_set$total_cases,
```

```
lm(
     shootings_unemployment_set$total_cases~shootings_unemployment_set$`Unemployment Rate`
     ))
,cex = 1.3,
pch = 16,
xlab = "Unemployment Rate",
ylab = "Shootings"
)
```

Unemployment Rate & Shootings Regression



The linear regression plot also suggests that variations in the Unemployment Rate could be a factor that contributes to variations in shooting incidents in NYC, as observed in the spike during 2020-2022.

Possible Bias Sources

- Unemployment rates were based on average value for New York State over 2010-2022, therefore including rates for all Boroughs/Counties in the state
- \bullet Boroughs and Counties were assumed to be interchangeable (eg: "Kings County" \sim "Brooklyn") for population mapping
- A more in depth analysis would be necessary, including other factors like income distribution, poverty levels, and correlation with other criminal activities in order to better understand the observed trend