

NYPD Shooting Analysis

Anna Behlke

2026-02-08

Introduction

This project analyzes the NYPD Shooting Incident dataset to examine trends over time, differences across boroughs, victim demographics, and factors associated with fatal outcomes.

Data Preparation

Packages used: tidyverse, lubridate, hms, scales, broom
Cleaned data: mutate occur_date, occur_time, year, create binary for fatal, filter by is not !is.na for occur_date, BORO, VIC_AGE_GROUP, VIC_SEX

```
shootings <- read_csv("NYPD_Shooting_Incident_Data_Historic_.csv")

## # Rows: 29744 Columns: 21
## -- Column specification -----
## Delimiter: ","
## chr (12): OCCUR_DATE, BORO, LOC_OF_OCCUR_DESC, LOC_CLASSFCTN_DESC, LOCATION...
## dbl (5): INCIDENT_KEY, PRECINCT, JURISDICTION_CODE, Latitude, Longitude
## num (2): X_COORD_CD, Y_COORD_CD
## lgl (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.

shootings_clean <- shootings %>%
  mutate(
    occur_date = mdy(OCCUR_DATE),
    occur_time = as_hms(OCCUR_TIME),
    year = year(occur_date),
    fatal = if_else(STATISTICAL_MURDER_FLAG == TRUE, 1, 0)
  ) %>%
  filter(
    !is.na(occur_date),
    !is.na(BORO),
    !is.na(VIC_AGE_GROUP),
    !is.na(VIC_SEX)
  )

summary(shootings_clean)

##      INCIDENT_KEY          OCCUR_DATE          OCCUR_TIME
## Min.   : 9953245   Length:29744   Min.   :00:00:00.000000
## 1st Qu.: 67321140  Class :character  1st Qu.:03:30:45.000000
## Median :109291972  Mode  :character  Median :15:15:00.000000
```

```

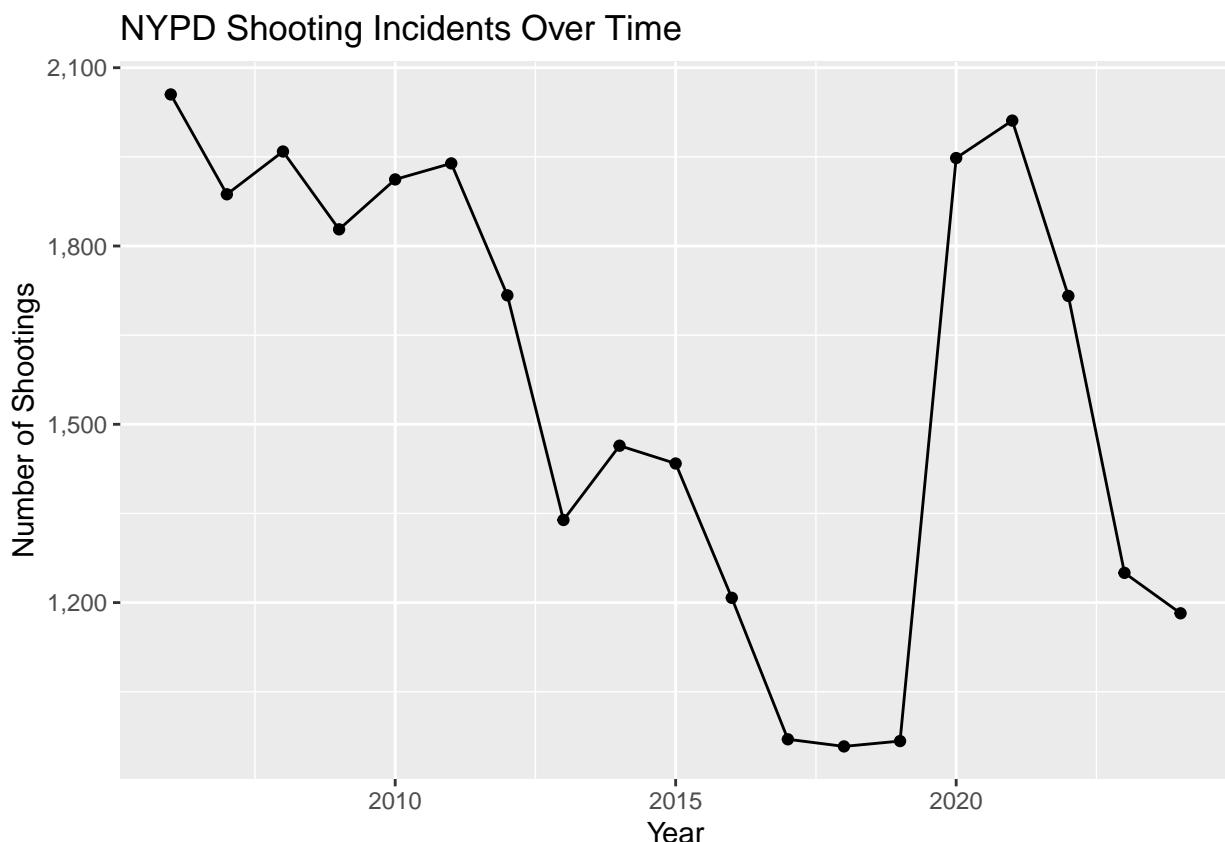
##  Mean   :133850951          Mean   :12:46:10.874798
##  3rd Qu.:214741917          3rd Qu.:20:44:00.000000
##  Max.   :299462478          Max.   :23:59:00.000000
##
##           BORO            LOC_OF_OCCUR_DESC      PRECINCT      JURISDICTION_CODE
##  Length:29744        Length:29744        Min.   : 1.00    Min.   :0.0000
##  Class :character    Class :character    1st Qu.: 44.00   1st Qu.:0.0000
##  Mode  :character    Mode  :character    Median : 67.00   Median :0.0000
##                                Mean   : 65.23   Mean   :0.3181
##                                3rd Qu.: 81.00   3rd Qu.:0.0000
##                                Max.   :123.00   Max.   :2.0000
##                                NA's   :2
##           LOC_CLASSFCTN_DESC LOCATION_DESC      STATISTICAL_MURDER_FLAG
##  Length:29744        Length:29744        Mode  :logical
##  Class :character    Class :character    FALSE:23979
##  Mode  :character    Mode  :character    TRUE :5765
##
##           PERP_AGE_GROUP     PERP_SEX        PERP_RACE        VIC_AGE_GROUP
##  Length:29744        Length:29744        Length:29744        Length:29744
##  Class :character    Class :character    Class :character    Class :character
##  Mode  :character    Mode  :character    Mode  :character    Mode  :character
##
##           VIC_SEX            VIC_RACE        X_COORD_CD        Y_COORD_CD
##  Length:29744        Length:29744        Min.   : 914928   Min.   :125757
##  Class :character    Class :character    1st Qu.:1000094   1st Qu.:183042
##  Mode  :character    Mode  :character    Median :1007826   Median :195506
##                                Mean   :1009442   Mean   :208722
##                                3rd Qu.:1016739   3rd Qu.:239980
##                                Max.   :1066815   Max.   :271128
##
##           Latitude       Longitude      Lon_Lat        occur_date
##  Min.   :40.51       Min.   :-74.25      Length:29744      Min.   :2006-01-01
##  1st Qu.:40.67       1st Qu.:-73.94      Class :character  1st Qu.:2009-10-29
##  Median :40.70       Median :-73.91      Mode  :character  Median :2014-03-25
##  Mean   :40.74       Mean   :-73.91      NA's  :97          Mean   :2014-10-31
##  3rd Qu.:40.83       3rd Qu.:-73.88      NA's  :97          3rd Qu.:2020-06-29
##  Max.   :40.91       Max.   :-73.70      NA's  :97          Max.   :2024-12-31
##           occur_time          year        fatal
##  Min.   :00:00:00.000000  Min.   :2006   Min.   :0.0000
##  1st Qu.:03:30:45.000000  1st Qu.:2009   1st Qu.:0.0000
##  Median :15:15:00.000000  Median :2014   Median :0.0000
##  Mean   :12:46:10.874798  Mean   :2014   Mean   :0.1938
##  3rd Qu.:20:44:00.000000  3rd Qu.:2020   3rd Qu.:0.0000
##  Max.   :23:59:00.000000  Max.   :2024   Max.   :1.0000
##

```

Exploratory Data Analysis

Shootings Over Time

```
shootings_clean %>%
  count(year) %>%
  ggplot(aes(year, n)) +
  geom_line() +
  geom_point() +
  scale_y_continuous(labels = comma) +
  labs(
    title = "NYPD Shooting Incidents Over Time",
    x = "Year",
    y = "Number of Shootings"
  )
```

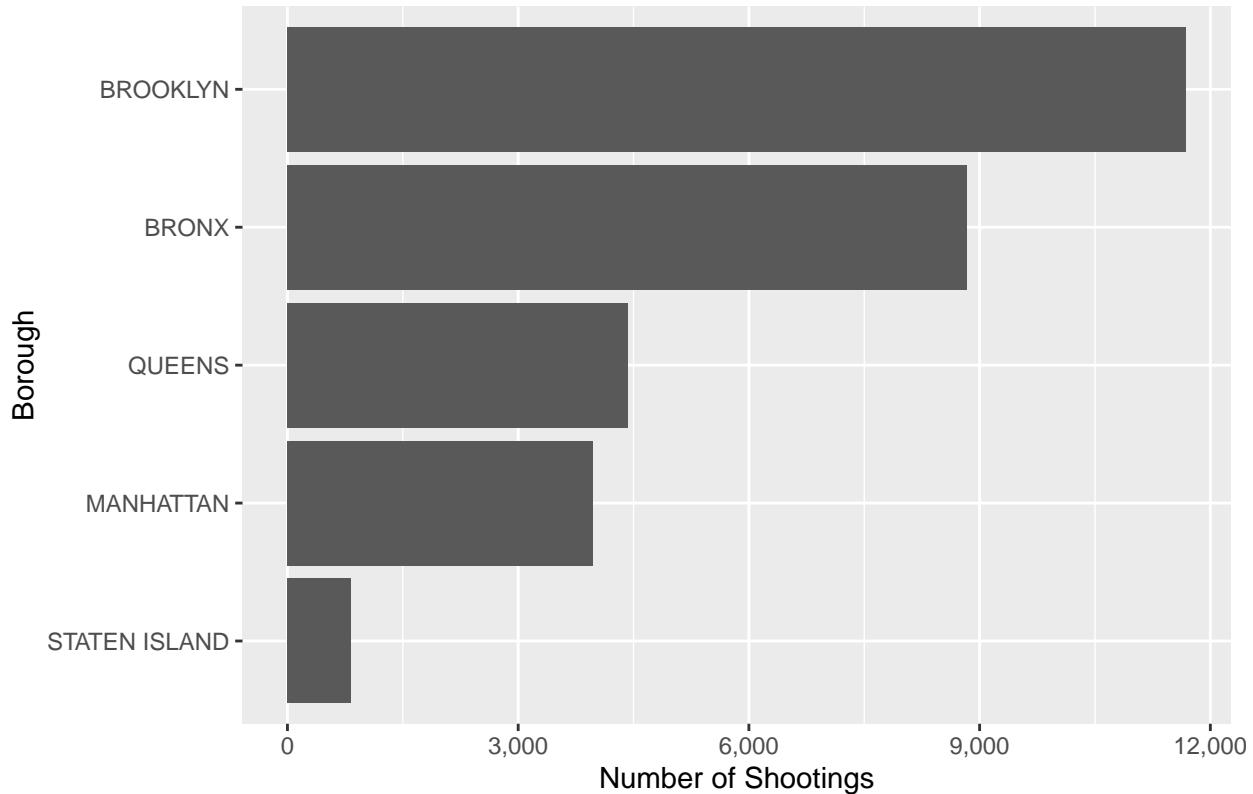


Shootings by Borough

```
shootings_clean %>%
  count(BORO) %>%
  ggplot(aes(reorder(BORO, n), n)) +
  geom_col() +
  coord_flip() +
  scale_y_continuous(labels = comma) +
  labs(
    title = "Shooting Incidents by Borough",
    x = "Borough",
```

```
y = "Number of Shootings"  
)
```

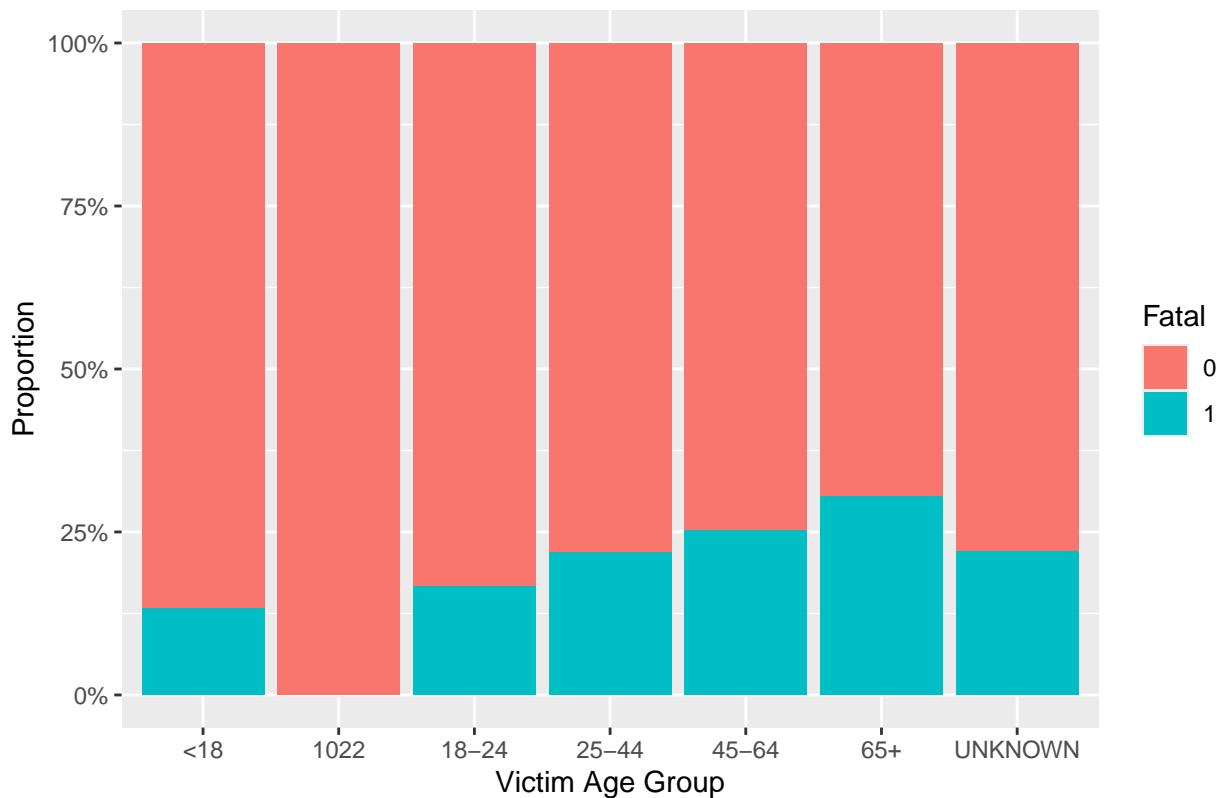
Shooting Incidents by Borough



Fatal vs Non-Fatal by Victim Age Group

```
shootings_clean %>%  
  ggplot(aes(VIC_AGE_GROUP, fill = factor(fatal))) +  
  geom_bar(position = "fill") +  
  scale_y_continuous(labels = percent) +  
  labs(  
    title = "Fatal vs Non-Fatal Shootings by Victim Age Group",  
    x = "Victim Age Group",  
    y = "Proportion",  
    fill = "Fatal"  
)
```

Fatal vs Non-Fatal Shootings by Victim Age Group



Modeling

Create logistic regression model using the cleaned data to determine if fatality is related to victim age, sex, and borough.

```

model_data <- shootings_clean %>%
  select(fatal, VIC_AGE_GROUP, VIC_SEX, BORO) %>%
  drop_na()

fatal_model <- glm(
  fatal ~ VIC_AGE_GROUP + VIC_SEX + BORO,
  data = model_data,
  family = binomial
)

summary(fatal_model)

##
## Call:
## glm(formula = fatal ~ VIC_AGE_GROUP + VIC_SEX + BORO, family = binomial,
##      data = model_data)
##
## Coefficients:
##                               Estimate Std. Error z value Pr(>|z|)
## (Intercept)             -1.82137   0.07139 -25.511 < 2e-16 ***
## VIC_AGE_GROUP1022     -7.58894   72.46288  -0.105   0.9166
## VIC_AGE_GROUP18-24      0.26494   0.05930   4.468 7.90e-06 ***

```

```

## VIC AGE GROUP25-44      0.61013   0.05719 10.668 < 2e-16 ***
## VIC AGE GROUP45-64      0.79153   0.07306 10.834 < 2e-16 ***
## VIC AGE GROUP65+       1.04987   0.15127  6.941 3.91e-12 ***
## VIC AGE GROUPUNKNOWN   0.68748   0.30233  2.274  0.0230 *
## VIC SEXM                -0.04024  0.04953 -0.812  0.4165
## VIC SEXU                -1.08936  1.06043 -1.027  0.3043
## BOROBROOKLYN           -0.02016  0.03574 -0.564  0.5727
## BOROMANHATTAN          -0.11540  0.04939 -2.336  0.0195 *
## BOROQUEENS              -0.01714  0.04658 -0.368  0.7130
## BOROSTATEN ISLAND      0.05413   0.09065  0.597  0.5504
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 29251  on 29743  degrees of freedom
## Residual deviance: 28992  on 29731  degrees of freedom
## AIC: 29018
##
## Number of Fisher Scoring iterations: 8
tidy(fatal_model, exponentiate = TRUE, conf.int = TRUE)

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## # A tibble: 13 x 7
##   term            estimate std.error statistic p.value conf.low conf.high
##   <chr>          <dbl>     <dbl>     <dbl>    <dbl>    <dbl>     <dbl>
## 1 (Intercept)  0.162      0.0714    -25.5  1.48e-143  0.141     0.186
## 2 VIC AGE GROUP1022 0.000506  72.5      -0.105 9.17e- 1 NA        5063.
## 3 VIC AGE GROUP18-24 1.30       0.0593     4.47  7.90e- 6  1.16      1.47
## 4 VIC AGE GROUP25-44 1.84       0.0572     10.7   1.43e- 26  1.65      2.06
## 5 VIC AGE GROUP45-64 2.21       0.0731     10.8   2.38e- 27  1.91      2.55
## 6 VIC AGE GROUP65+  2.86       0.151      6.94  3.91e- 12  2.11      3.83
## 7 VIC AGE GROUPUNKNO~ 1.99       0.302      2.27  2.30e- 2   1.06      3.51
## 8 VIC SEXM         0.961      0.0495    -0.812 4.17e- 1  0.872     1.06
## 9 VIC SEXU         0.336       1.06      -1.03  3.04e- 1  0.0181    1.81
## 10 BOROBROOKLYN   0.980      0.0357    -0.564 5.73e- 1  0.914     1.05
## 11 BOROMANHATTAN  0.891      0.0494    -2.34  1.95e- 2   0.808    0.981
## 12 BOROQUEENS    0.983      0.0466    -0.368 7.13e- 1  0.897     1.08
## 13 BOROSTATEN ISLAND 1.06      0.0906    0.597  5.50e- 1  0.881     1.26

```

Bias and Limitations

This dataset includes only NYPD-reported incidents and may reflect reporting bias, missing demographic data, and changes in policing practices over time. The model is exploratory and does not imply causation.

Conclusion

The analysis shows variation in shooting incidents over time and across boroughs, with differences in fatal outcomes across victim demographics. Further analysis could include spatial or policy-related factors.