# NYPD Shooting Incident Data Report

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### NYPD Data-Load and Tidy

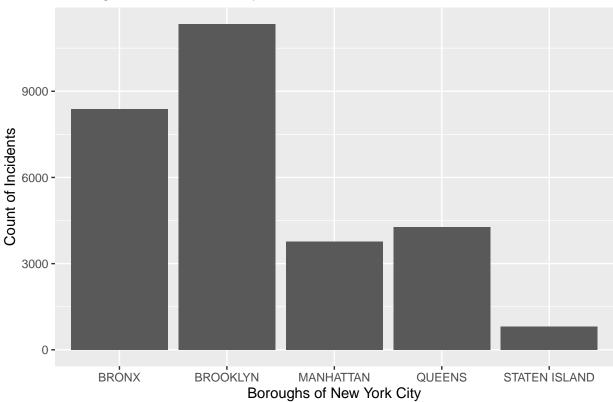
Below is a summary of the NYPD Shooting Data set. We begin by loading in the dataset and selecting only the relevant columns for our analysis. We then summarize the dataset

```
OCCUR_DATE
                                            OCCUR_TIME
                                                                  BORO
##
     INCIDENT KEY
##
                        Length: 28562
                                           Length:28562
                                                              Length: 28562
         : 9953245
   1st Qu.: 65439914
                                           Class1:hms
                        Class : character
                                                              Class : character
                                           Class2:difftime
## Median : 92711254
                        Mode : character
                                                              Mode :character
## Mean
          :127405824
                                           Mode :numeric
## 3rd Qu.:203131993
## Max.
           :279758069
## STATISTICAL_MURDER_FLAG PERP_AGE_GROUP
                                                 PERP_SEX
## Mode :logical
                            Length: 28562
                                               Length: 28562
## FALSE:23036
                            Class :character
                                               Class : character
##
   TRUE :5526
                            Mode :character
                                               Mode :character
##
##
##
##
    PERP_RACE
                       VIC_AGE_GROUP
                                            VIC_SEX
                                                                VIC_RACE
##
   Length: 28562
                       Length: 28562
                                          Length: 28562
                                                              Length: 28562
  Class :character
                       Class :character
                                          Class :character
                                                              Class : character
   Mode :character Mode :character
                                          Mode :character
                                                              Mode : character
##
##
##
```

## **Analysis and Research Questions**

1) What areas of New York City has the most number of incidents.

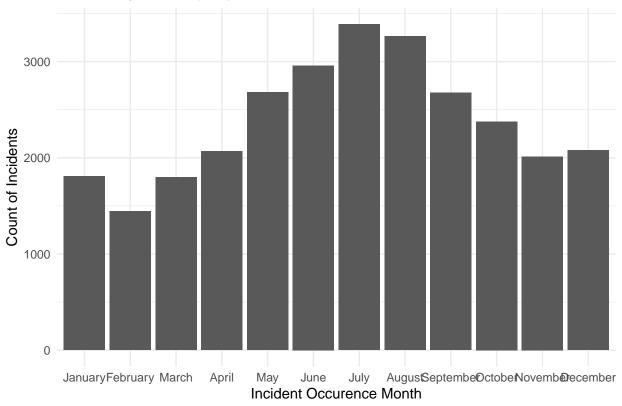
# Boroughs of New York City



From the above graph, we can clearly see that Brooklyn is the neighbourhood with the most number of incidents in New York City with over 10000 incidents. Staten Island is the neighbourhood with the least number of incidents with less than 1000 incidents being reported. Individuals in Brooklyn should be more cautious and vigilant in New York City.

2) Which months saw the most incidents/ murders





From the above graph, we can see that July and August are the months with the most number of incidents in the year and February has the least number of incidents. We are also able to see overall that summer months and warmer months tend to have more incidents that the winter months. This may be mainly due to the fact that people tend to spend a lot more time outdoors during the warmer months making them potentially more vulnerable to shooting incidents in New York City.

3) Building a logistic regression model to predict whether an incident is a murder or not

```
##
   glm(formula = STATISTICAL_MURDER_FLAG ~ PERP_RACE + PERP_SEX +
##
       BORO, family = binomial, data = nypd)
##
##
  Coefficients: (1 not defined because of singularities)
##
                                            Estimate Std. Error z value Pr(>|z|)
## (Intercept)
                                            -1.66962
                                                         0.08812 -18.946 < 2e-16
## PERP_RACEAMERICAN INDIAN/ALASKAN NATIVE -8.58355
                                                        84.06292
                                                                  -0.102 0.918670
## PERP_RACEASIAN / PACIFIC ISLANDER
                                                         0.30100
                                                                   4.380 1.18e-05
                                             1.31850
## PERP_RACEBLACK
                                             0.82350
                                                         0.25016
                                                                   3.292 0.000995
## PERP_RACEBLACK HISPANIC
                                                         0.25800
                                             0.73988
                                                                   2.868 0.004135
## PERP RACEUNKNOWN
                                            -0.82964
                                                         0.13101
                                                                  -6.333 2.41e-10
## PERP_RACEWHITE
                                             1.62726
                                                         0.27744
                                                                   5.865 4.48e-09
## PERP RACEWHITE HISPANIC
                                             0.99983
                                                         0.25378
                                                                   3.940 8.16e-05
                                                                  -0.478 0.632913
## PERP_SEXF
                                            -0.12340
                                                         0.25836
## PERP SEXM
                                            -0.39075
                                                         0.23501
                                                                  -1.663 0.096377
## PERP SEXU
                                                              NA
                                                                      NA
                                                                                NA
                                                  NA
```

```
## BOROBROOKLYN
                                            -0.13276
                                                        0.04564
                                                                 -2.909 0.003624
## BOROMANHATTAN
                                                        0.05861
                                                                 -2.654 0.007946
                                            -0.15556
                                                        0.05767
## BOROQUEENS
                                            -0.13138
                                                                  -2.278 0.022716
## BOROSTATEN ISLAND
                                            -0.10203
                                                        0.10166
                                                                 -1.004 0.315574
##
  (Intercept)
## PERP RACEAMERICAN INDIAN/ALASKAN NATIVE
## PERP RACEASIAN / PACIFIC ISLANDER
                                            ***
## PERP_RACEBLACK
## PERP_RACEBLACK HISPANIC
                                            **
## PERP_RACEUNKNOWN
## PERP_RACEWHITE
                                            ***
## PERP RACEWHITE HISPANIC
## PERP SEXF
## PERP_SEXM
## PERP_SEXU
## BOROBROOKLYN
## BOROMANHATTAN
## BOROQUEENS
## BOROSTATEN ISLAND
##
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
   (Dispersion parameter for binomial family taken to be 1)
##
##
##
       Null deviance: 19230
                             on 19251
                                       degrees of freedom
## Residual deviance: 18827
                             on 19238
                                        degrees of freedom
##
     (9310 observations deleted due to missingness)
## AIC: 18855
##
## Number of Fisher Scoring iterations: 9
```

Finally, we can build a logistic regression model to help predict whether an incident is going to be a murder or not. I will use the perpetrator sex, race and boroughs to determine if an incident is a murder or not. From the above model, we can see that an incident occurring in Brooklyn changes the log odds of murder by -0.13.

#### 4) Potential Sources of Bias

Potential sources of bias in analyzing NYPD shooting data include reporting bias, where certain incidents may be underreported or misclassified; selection bias, if the dataset does not capture all relevant cases; and data collection bias, influenced by how and why data is recorded. Demographic bias may arise if certain groups are over- or underrepresented due to systemic factors. Additionally, analytical bias, such as choosing specific metrics or framing results in a particular way, can impact interpretations. For example, one might feel that women might be more vulnerable to incidents or certain neighbourhoods like the Bronx might have the most incidents due to personal experiences or assumptions. Acknowledging these biases and testing them first is crucial to ensuring a fair, transparent, and data-driven analysis.