

NYPD Shooting Incident Data Report

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The NYPD Shooting Incident Data (Historic) csv file is used in this data report. It is obtained from the data catalog of the U.S Government's open data initiative (data.gov).

```
#Start by using the Tidyverse library
library(tidyverse)
```

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

```
## v ggplot2 3.3.4      v purrr  0.3.4
## v tibble  3.1.2      v dplyr  1.0.7
## v tidyr   1.1.3      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()
```

```
#Obtain NYPD Shooting Incident CSV file from data.gov
```

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

The csv file that we will import contains a list of all shooting incidents that took place in NYC from 2006 to November 10, 2020. The event, location, and time of each shooting incident are also included along with demographic information pertaining to the suspects and victims.

```
#Read in CSV file
```

```
nypd_data <- read_csv(url_link)
```

We will first inspect and view the csv file.

```
#View the dataset
```

```
nypd_data
```

```
## # A tibble: 23,568 x 19
```

	INCIDENT_KEY	OCCUR_DATE	OCCUR_TIME	BORO	PRECINCT	JURISDICTION_CODE
	<dbl>	<chr>	<time>	<chr>	<dbl>	<dbl>
## 1	201575314	08/23/2019	22:10	QUEENS	103	0
## 2	205748546	11/27/2019	15:54	BRONX	40	0
## 3	193118596	02/02/2019	19:40	MANHATTAN	23	0
## 4	204192600	10/24/2019	00:52	STATEN ISLAND	121	0

```
## 5      201483468 08/22/2019 18:03      BRONX              46              0
## 6      198255460 06/07/2019 17:50      BROOKLYN            73              0
## 7      194570529 03/11/2019 16:30      BROOKLYN            81              0
## 8      203211777 10/03/2019 01:45      BROOKLYN            67              0
## 9      193694863 02/17/2019 03:00      QUEENS              114             2
## 10     199582060 07/10/2019 02:56      BROOKLYN            69              0
## # ... with 23,558 more rows, and 13 more variables: LOCATION_DESC <chr>,
## #   STATISTICAL_MURDER_FLAG <lgl>, PERP_AGE_GROUP <chr>, PERP_SEX <chr>,
## #   PERP_RACE <chr>, VIC_AGE_GROUP <chr>, VIC_SEX <chr>, VIC_RACE <chr>,
## #   X_COORD_CD <dbl>, Y_COORD_CD <dbl>, Latitude <dbl>, Longitude <dbl>,
## #   Lon_Lat <chr>
```

Then, we will obtain the summary statistics of the data.

```
summary(nypd_data)
```

```
##      INCIDENT_KEY      OCCUR_DATE      OCCUR_TIME      BORO
## Min.   : 9953245      Length:23568      Length:23568      Length:23568
## 1st Qu.: 55317014      Class :character      Class1:hms        Class :character
## Median : 83365370      Mode  :character      Class2:difftime   Mode  :character
## Mean   :102218616                      Mode  :numeric
## 3rd Qu.:150772442
## Max.   :222473262
##
##      PRECINCT      JURISDICTION_CODE      LOCATION_DESC      STATISTICAL_MURDER_FLAG
## Min.   : 1.00      Min.   :0.0000      Length:23568      Mode :logical
## 1st Qu.: 44.00      1st Qu.:0.0000      Class :character      FALSE:19080
## Median : 69.00      Median :0.0000      Mode  :character      TRUE :4488
## Mean   : 66.21      Mean   :0.3323
## 3rd Qu.: 81.00      3rd Qu.:0.0000
## Max.   :123.00      Max.   :2.0000
##
##      NA's :2
##      PERP_AGE_GROUP      PERP_SEX      PERP_RACE      VIC_AGE_GROUP
## Length:23568      Length:23568      Length:23568      Length:23568
## 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:23568      Length:23568      Min.   : 914928      Min.   :125757
## Class :character      Class :character      1st Qu.: 999900      1st Qu.:182565
## Mode  :character      Mode  :character      Median :1007645      Median :193482
##
##      Mean   :1009363      Mean   :207312
##      3rd Qu.:1016807      3rd Qu.:239163
##      Max.   :1066815      Max.   :271128
##
##
##      Latitude      Longitude      Lon_Lat
## Min.   :40.51      Min.   : -74.25      Length:23568
## 1st Qu.:40.67      1st Qu.: -73.94      Class :character
## Median :40.70      Median : -73.92      Mode  :character
## Mean   :40.74      Mean   : -73.91
```

```
## 3rd Qu.:40.82 3rd Qu.: -73.88
## Max. :40.91 Max. : -73.70
##
```

There are 19 variables in this data set. We would like to explore the relationships between the variables. More specifically, we will determine the frequency of the shooting incidents by location, analyze the demographics of both parties, and investigate whether the victim's death is dependent on the borough. To answer these questions, we would like to include all columns except INCIDENT_KEY, X_COORD_CD, Y_COORD_CD, Latitude, Longitude, and Lon_Lat. Since the OCCUR_DATE column has characters, we will transform those dates into a date type variable using the lubridate library.

```
#Use lubridate library
library(lubridate)
```

```
##
## Attaching package: 'lubridate'

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

```
#Clean up data by removing some columns and transform OCCUR_DATE column to contain date type values
nypd <- nypd_data %>%
  select(OCCUR_DATE:VIC_RACE) %>%
  mutate(OCCUR_DATE=mdy(OCCUR_DATE))

nypd
```

```
## # A tibble: 23,568 x 13
##   OCCUR_DATE OCCUR_TIME BORO      PRECINCT JURISDICTION_CODE LOCATION_DESC
##   <date>      <time>    <chr>      <dbl>      <dbl> <chr>
## 1 2019-08-23 22:10    QUEENS      103          0 <NA>
## 2 2019-11-27 15:54    BRONX        40          0 <NA>
## 3 2019-02-02 19:40    MANHATTAN    23          0 <NA>
## 4 2019-10-24 00:52    STATEN I~    121         0 PVT HOUSE
## 5 2019-08-22 18:03    BRONX        46          0 <NA>
## 6 2019-06-07 17:50    BROOKLYN     73          0 <NA>
## 7 2019-03-11 16:30    BROOKLYN     81          0 <NA>
## 8 2019-10-03 01:45    BROOKLYN     67          0 MULTI DWELL - APT~
## 9 2019-02-17 03:00    QUEENS      114          2 MULTI DWELL - PUB~
## 10 2019-07-10 02:56    BROOKLYN     69          0 <NA>
## # ... with 23,558 more rows, and 7 more variables:
## #   STATISTICAL_MURDER_FLAG <lgl>, PERP_AGE_GROUP <chr>, PERP_SEX <chr>,
## #   PERP_RACE <chr>, VIC_AGE_GROUP <chr>, VIC_SEX <chr>, VIC_RACE <chr>
```

The summary statistics of the transformed dataset is examined.

```
summary(nypd)
```

```
##   OCCUR_DATE      OCCUR_TIME      BORO      PRECINCT
##  Min.   :2006-01-01 Length:23568 Length:23568 Min.   : 1.00
```

```
## 1st Qu.:2008-12-30    Class1:hms          Class :character    1st Qu.: 44.00
## Median :2012-02-26    Class2:difftime      Mode  :character    Median : 69.00
## Mean   :2012-10-03    Mode   :numeric      Mean   : 66.21
## 3rd Qu.:2016-02-28                                3rd Qu.: 81.00
## Max.   :2020-12-31                                Max.   :123.00
##
## JURISDICTION_CODE LOCATION_DESC      STATISTICAL_MURDER_FLAG
## Min.   :0.0000      Length:23568      Mode :logical
## 1st Qu.:0.0000      Class :character    FALSE:19080
## Median :0.0000      Mode  :character    TRUE :4488
## Mean   :0.3323
## 3rd Qu.:0.0000
## Max.   :2.0000
## NA's   :2
## PERP_AGE_GROUP      PERP_SEX          PERP_RACE          VIC_AGE_GROUP
## Length:23568        Length:23568        Length:23568        Length:23568
## Class :character    Class :character    Class :character    Class :character
## Mode  :character    Mode  :character    Mode  :character    Mode  :character
##
##
##
##
## VIC_SEX              VIC_RACE
## Length:23568          Length:23568
## Class :character      Class :character
## Mode  :character      Mode  :character
##
##
##
##
```

We will now identify the class for all the column variables in the transformed dataset.

```
str(nypd)
```

```
## tibble [23,568 x 13] (S3: tbl_df/tbl/data.frame)
## $ OCCUR_DATE      : Date[1:23568], format: "2019-08-23" "2019-11-27" ...
## $ OCCUR_TIME      : 'hms' num [1:23568] 22:10:00 15:54:00 19:40:00 00:52:00 ...
## ..- attr(*, "units")= chr "secs"
## $ BORO            : chr [1:23568] "QUEENS" "BRONX" "MANHATTAN" "STATEN ISLAND" ...
## $ PRECINCT        : num [1:23568] 103 40 23 121 46 73 81 67 114 69 ...
## $ JURISDICTION_CODE : num [1:23568] 0 0 0 0 0 0 0 0 2 0 ...
## $ LOCATION_DESC    : chr [1:23568] NA NA NA "PVT HOUSE" ...
## $ STATISTICAL_MURDER_FLAG: logi [1:23568] FALSE FALSE FALSE TRUE FALSE FALSE ...
## $ PERP_AGE_GROUP    : chr [1:23568] NA "<18" "18-24" "25-44" ...
## $ PERP_SEX          : chr [1:23568] NA "M" "M" "M" ...
## $ PERP_RACE         : chr [1:23568] NA "BLACK" "WHITE HISPANIC" "BLACK" ...
## $ VIC_AGE_GROUP     : chr [1:23568] "25-44" "25-44" "18-24" "25-44" ...
## $ VIC_SEX           : chr [1:23568] "M" "F" "M" "F" ...
## $ VIC_RACE          : chr [1:23568] "BLACK" "BLACK" "BLACK HISPANIC" "BLACK" ...
```

Eight out of thirteen variables have characters as the class of the variables. We will convert those column variables with character class into factors.

```
nypd_c <- as.data.frame(unclass(nypd))
nypd_c[sapply(nypd_c, is.character)] <- lapply(nypd_c[sapply(nypd_c, is.character)], as.factor)
```

We verify that all column variables previously with the character class are now converted into the factor class.

```
str(nypd_c)
```

```
## 'data.frame': 23568 obs. of 13 variables:
## $ OCCUR_DATE : Date, format: "2019-08-23" "2019-11-27" ...
## $ OCCUR_TIME : 'hms' num 22:10:00 15:54:00 19:40:00 00:52:00 ...
## ..- attr(*, "units")= chr "secs"
## $ BORO : Factor w/ 5 levels "BRONX","BROOKLYN",...: 4 1 3 5 1 2 2 2 4 2 ...
## $ PRECINCT : num 103 40 23 121 46 73 81 67 114 69 ...
## $ JURISDICTION_CODE : num 0 0 0 0 0 0 0 0 2 0 ...
## $ LOCATION_DESC : Factor w/ 39 levels "ATM","BANK","BAR/NIGHT CLUB",...: NA NA NA 28 NA NA ...
## $ STATISTICAL_MURDER_FLAG: logi FALSE FALSE FALSE TRUE FALSE FALSE ...
## $ PERP_AGE_GROUP : Factor w/ 9 levels "<18","1020","18-24",...: NA 1 3 5 5 6 3 NA 3 5 ...
## $ PERP_SEX : Factor w/ 3 levels "F","M","U": NA 2 2 2 2 2 2 NA 2 2 ...
## $ PERP_RACE : Factor w/ 7 levels "AMERICAN INDIAN/ALASKAN NATIVE",...: NA 3 7 3 4 7 3 NA ...
## $ VIC_AGE_GROUP : Factor w/ 6 levels "<18","18-24",...: 3 3 2 3 2 3 3 3 3 3 ...
## $ VIC_SEX : Factor w/ 3 levels "F","M","U": 2 1 2 1 2 2 2 2 2 2 ...
## $ VIC_RACE : Factor w/ 7 levels "AMERICAN INDIAN/ALASKAN NATIVE",...: 3 3 4 3 3 3 3 3 3 3 ...
```

The summary statistics of the new data frame is gathered once again.

```
summary(nypd_c)
```

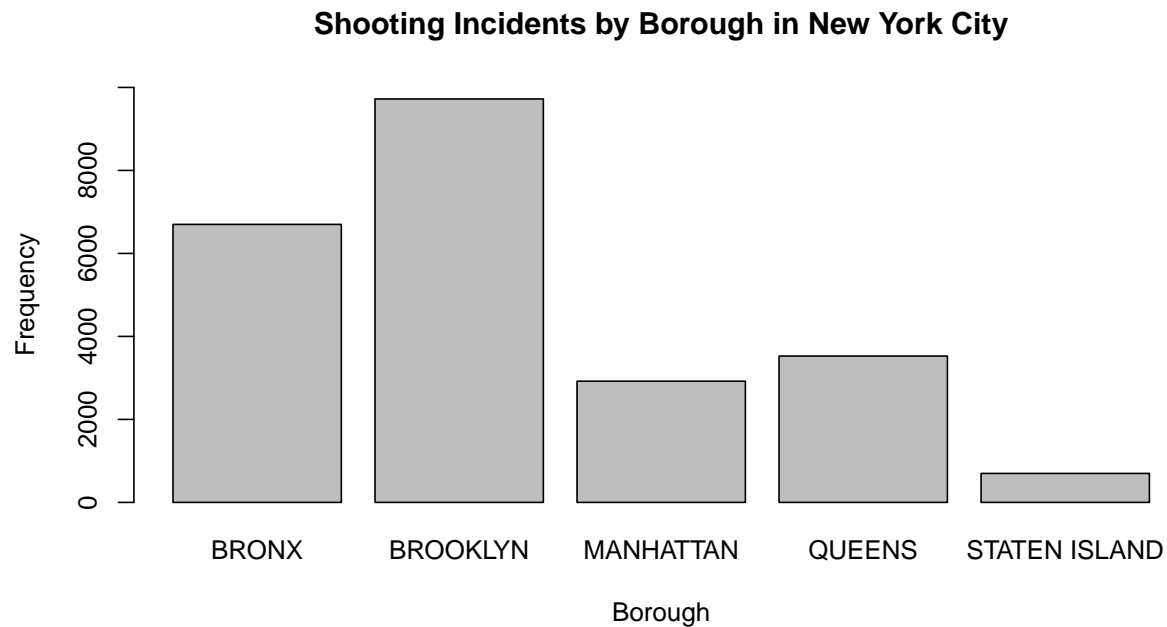
```
## OCCUR_DATE OCCUR_TIME BORO PRECINCT
## Min. :2006-01-01 Length:23568 BRONX :6700 Min. : 1.00
## 1st Qu.:2008-12-30 Class1:hms BROOKLYN :9722 1st Qu.: 44.00
## Median :2012-02-26 Class2:difftime MANHATTAN :2921 Median : 69.00
## Mean :2012-10-03 Mode :numeric QUEENS :3527 Mean : 66.21
## 3rd Qu.:2016-02-28 STATEN ISLAND: 698 3rd Qu.: 81.00
## Max. :2020-12-31 Max. :123.00
##
## JURISDICTION_CODE LOCATION_DESC STATISTICAL_MURDER_FLAG
## Min. :0.0000 MULTI DWELL - PUBLIC HOUS: 4230 Mode :logical
## 1st Qu.:0.0000 MULTI DWELL - APT BUILD : 2551 FALSE:19080
## Median :0.0000 PVT HOUSE : 858 TRUE :4488
## Mean :0.3323 GROCERY/BODEGA : 572
## 3rd Qu.:0.0000 BAR/NIGHT CLUB : 558
## Max. :2.0000 (Other) : 1218
## NA's :2 NA's :13581
## PERP_AGE_GROUP PERP_SEX PERP_RACE VIC_AGE_GROUP VIC_SEX
## 18-24 :5448 F : 334 BLACK :9855 <18 : 2525 F: 2195
## 25-44 :4613 M :13305 WHITE HISPANIC:1961 18-24 : 9000 M:21353
## UNKNOWN:3156 U : 1504 UNKNOWN :1869 25-44 :10287 U: 20
## <18 :1354 NA's: 8425 BLACK HISPANIC:1081 45-64 : 1536
## 45-64 : 481 WHITE : 255 65+ : 155
## (Other): 57 (Other) : 122 UNKNOWN: 65
## NA's :8459 NA's :8425
```

```
##                VIC_RACE
## AMERICAN INDIAN/ALASKAN NATIVE:    9
## ASIAN / PACIFIC ISLANDER          : 320
## BLACK                             :16846
## BLACK HISPANIC                     : 2244
## UNKNOWN                           :  102
## WHITE                             :  615
## WHITE HISPANIC                     : 3432
```

From the summary, we discover that the JURISDICTION_CODE, LOCATION_DESC, PERP_AGE_GROUP, PERP_SEX, PERP_RACE all contain missing values or NAs. When we examine those variables, we will remove the missing data for our analysis. Similarly, “unknown” values are recorded in the PERP_AGE_GROUP, VIC_AGE_GROUP, VIC_SEX, and VIC_RACE column variables.

First, we want to visualize the frequency distribution of shooting incidents by borough with a bar chart.

```
freq_boro <- table(nypd_c$BORO)
barplot(freq_boro, xlab = "Borough", ylab = "Frequency", main = "Shooting Incidents by Borough in New York City")
```



The bar chart shows that out of all the boroughs recorded in New York City, Brooklyn has the highest frequency of shooting incidents, followed by Bronx, Queens, Manhattan, and Staten Island. This may suggest that Brooklyn is less safe compared to other boroughs in New York City.

Next, we want to examine the demographic information for both the perpetrators and victims. We will subset the data set further by looking at the relevant columns pertaining to both parties and removing cases with missing values.

```
nypd_demog <- nypd_c %>%
  select(PERP_AGE_GROUP:VIC_RACE)
nypd_narm <- na.omit(nypd_demog)
```

The summary function is run on the new dataframe to ensure that all missing datas are removed before analysis.

```
summary(nypd_narm)
```

```
## PERP_AGE_GROUP PERP_SEX PERP_RACE VIC_AGE_GROUP
## 18-24 :5448 F: 334 AMERICAN INDIAN/ALASKAN NATIVE: 2 <18 :1788
## 25-44 :4613 M:13305 ASIAN / PACIFIC ISLANDER : 120 18-24 :5714
## UNKNOWN:3156 U: 1470 BLACK :9855 25-44 :6400
## <18 :1354 BLACK HISPANIC :1081 45-64 :1033
## 45-64 : 481 UNKNOWN :1835 65+ : 117
## 65+ : 54 WHITE : 255 UNKNOWN: 57
## (Other): 3 WHITE HISPANIC :1961
## VIC_SEX VIC_RACE
## F: 1576 AMERICAN INDIAN/ALASKAN NATIVE: 7
## M:13521 ASIAN / PACIFIC ISLANDER : 235
## U: 12 BLACK :10325
## BLACK HISPANIC : 1490
## UNKNOWN : 68
## WHITE : 477
## WHITE HISPANIC : 2507
```

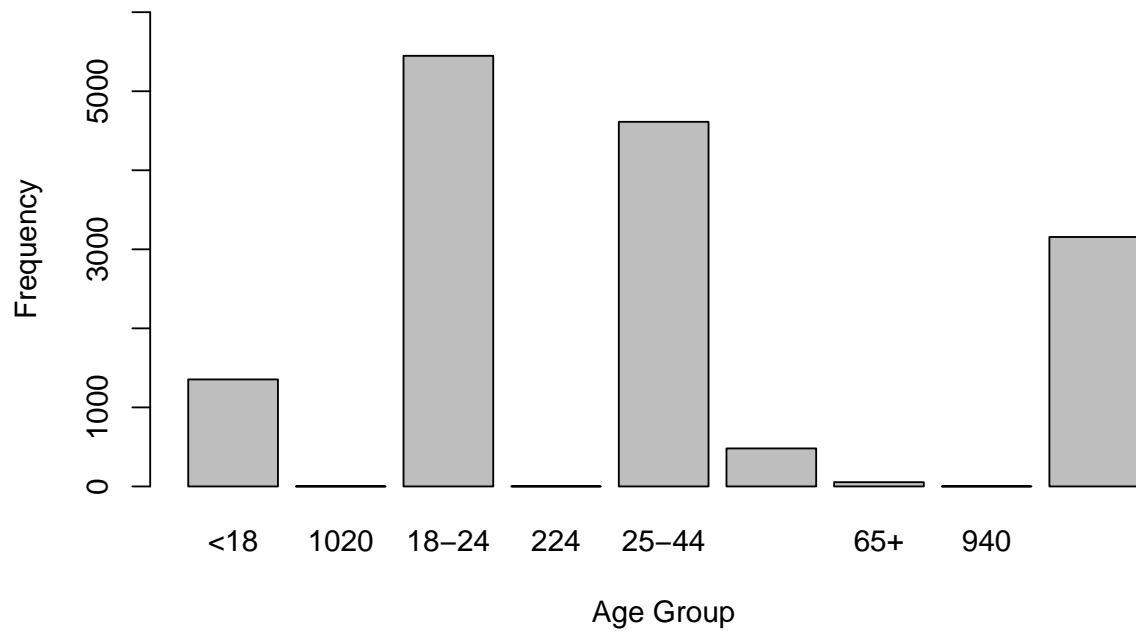
Now that the missing values are removed, we can proceed with obtaining the demographic information for the perpetrators as well as victims starting with age.

```
p_age <- table(nypd_narm$PERP_AGE_GROUP)
p_age
```

```
##
## <18 1020 18-24 224 25-44 45-64 65+ 940 UNKNOWN
## 1354 1 5448 1 4613 481 54 1 3156
```

```
barplot(p_age, xlab = "Age Group", ylab = "Frequency", main = "Age Group Frequency Distribution of Perp
```

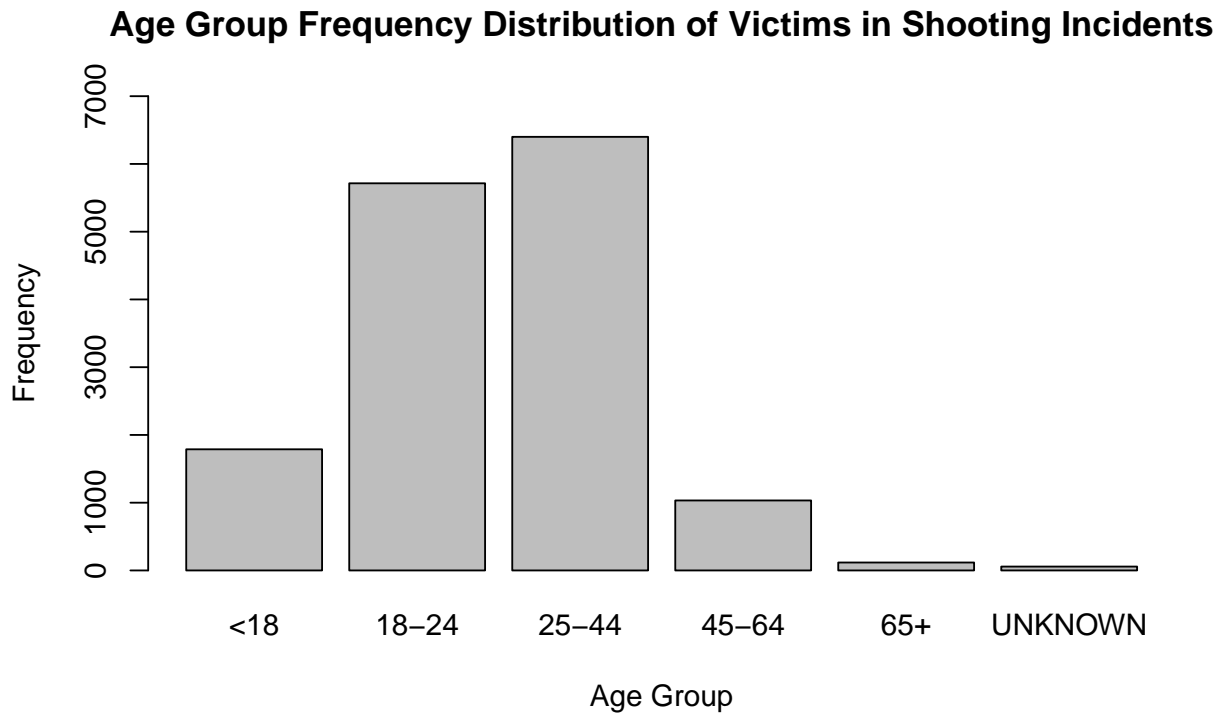
Age Group Frequency Distribution of Perpetrators in Shooting Incidents



```
v_age <- table(nypd_narm$VIC_AGE_GROUP)
v_age
```

```
##
##    <18    18-24    25-44    45-64    65+ UNKNOWN
##    1788    5714    6400    1033    117     57
```

```
barplot(v_age, xlab = "Age Group", ylab = "Frequency", main = "Age Group Frequency Distribution of Vict")
```

For perpetrators in the shooting incidents, there are almost 5500 that belong in the 18-24 age group. On the other hand, the majority of victims are from the 25-44 age group. There is a drop in counts in the lower end (below 18) and higher end (65+) of the age groups for both parties. More analysis is needed to determine the cause in the disproportion. There are also 3 outliers in the variable coding the perpetrators' age. Since they are beyond the normal lifespan, it is likely these are typos in data entry.

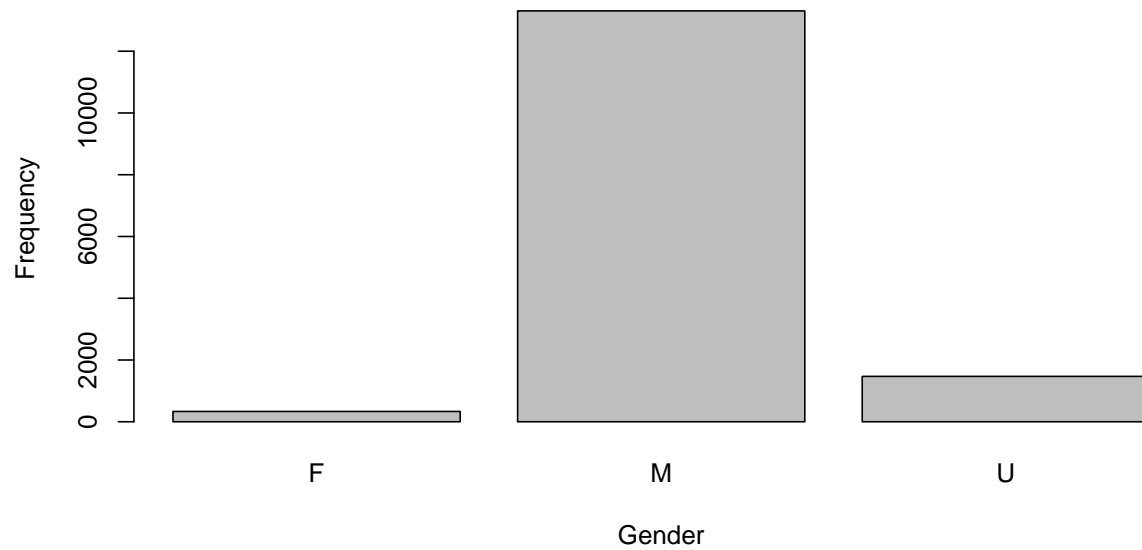
Next, we will compare the frequency distribution of gender for both the suspects and victims.

```
p_sex <- table(nypd_narm$PERP_SEX)
p_sex
```

```
##
##      F      M      U
##  334 13305  1470
```

```
barplot(p_sex,xlab = "Gender", ylab = "Frequency", main = "Frequency Distribution of Suspects' Gender in Shooting Incidents")
```

Frequency Distribution of Suspects' Gender in Shooting Incidents

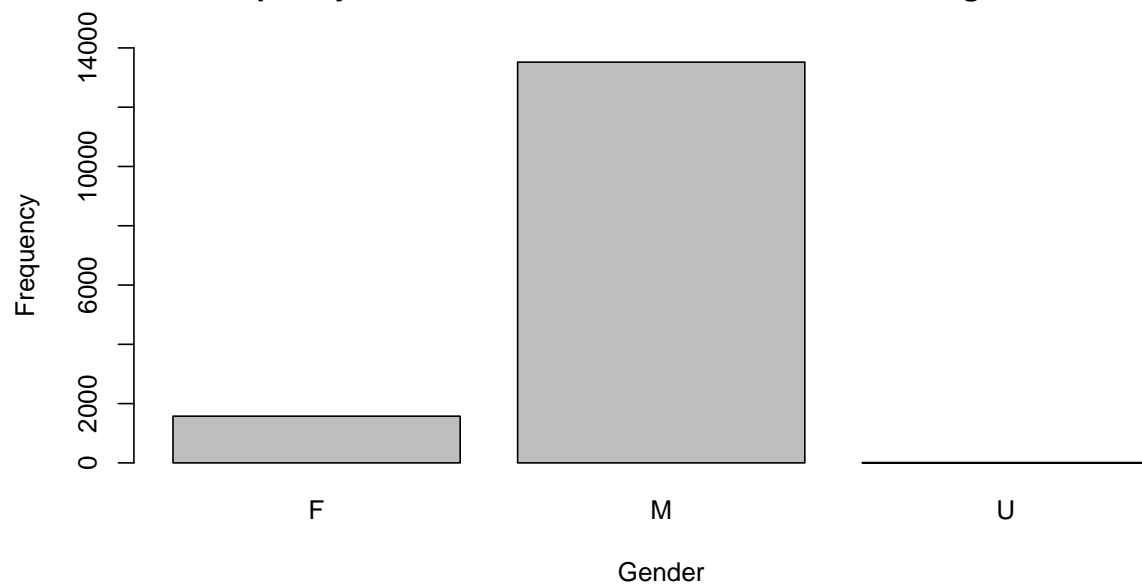


```
v_sex <- table(nypd_narm$VIC_SEX)
v_sex
```

```
##
##      F      M      U
## 1576 13521    12
```

```
barplot(v_sex,xlab = "Gender", ylab = "Frequency", main = "Frequency Distribution of Victims' Gender in Shooting Incidents")
```

Frequency Distribution of Victims' Gender in Shooting Incidents



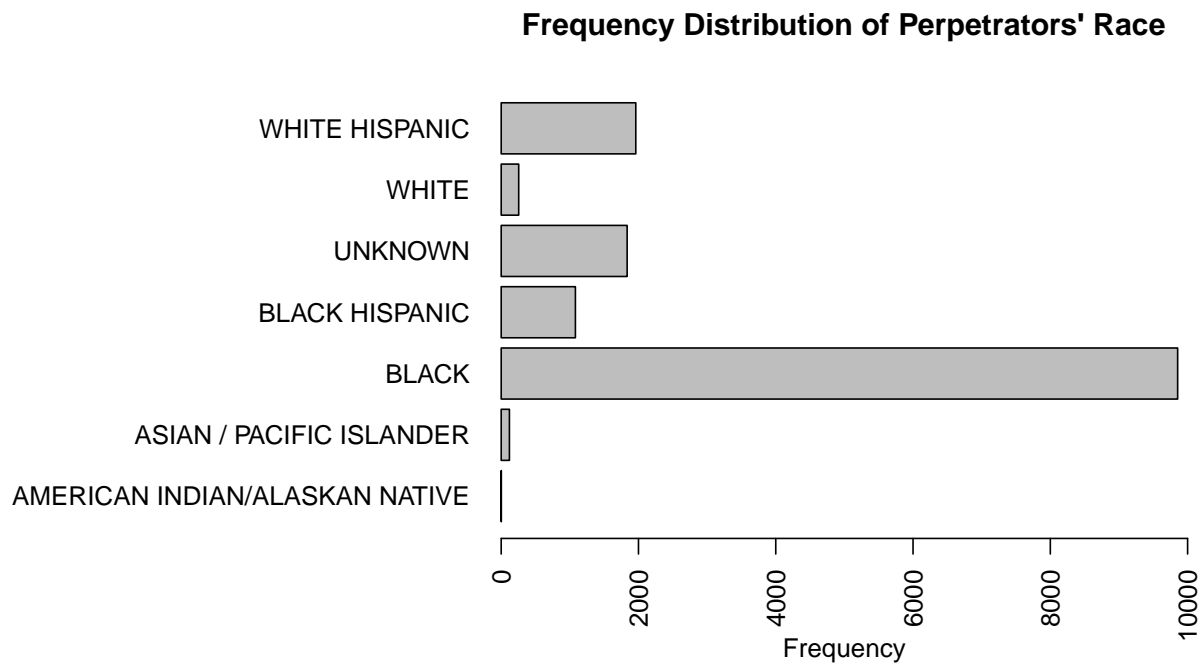
For both suspects and victims, more than half are male. This seems to suggest that males are more likely to be involved in shooting incidents in New York City.

The last category of demographic information we want to explore from the data set is race for both parties.

```
p_race <- table(nypd_narm$PERP_RACE)
p_race
```

```
##
## AMERICAN INDIAN/ALASKAN NATIVE      ASIAN / PACIFIC ISLANDER
##                                2                                120
##                                BLACK                                BLACK HISPANIC
##                                9855                               1081
##                                UNKNOWN                                WHITE
##                                1835                                255
##                                WHITE HISPANIC
##                                1961
```

```
par(mar = c(4,16,4,2))
barplot(p_race,xlab = "Frequency", main = "Frequency Distribution of Perpetrators' Race", xlim = c(0,10000))
```

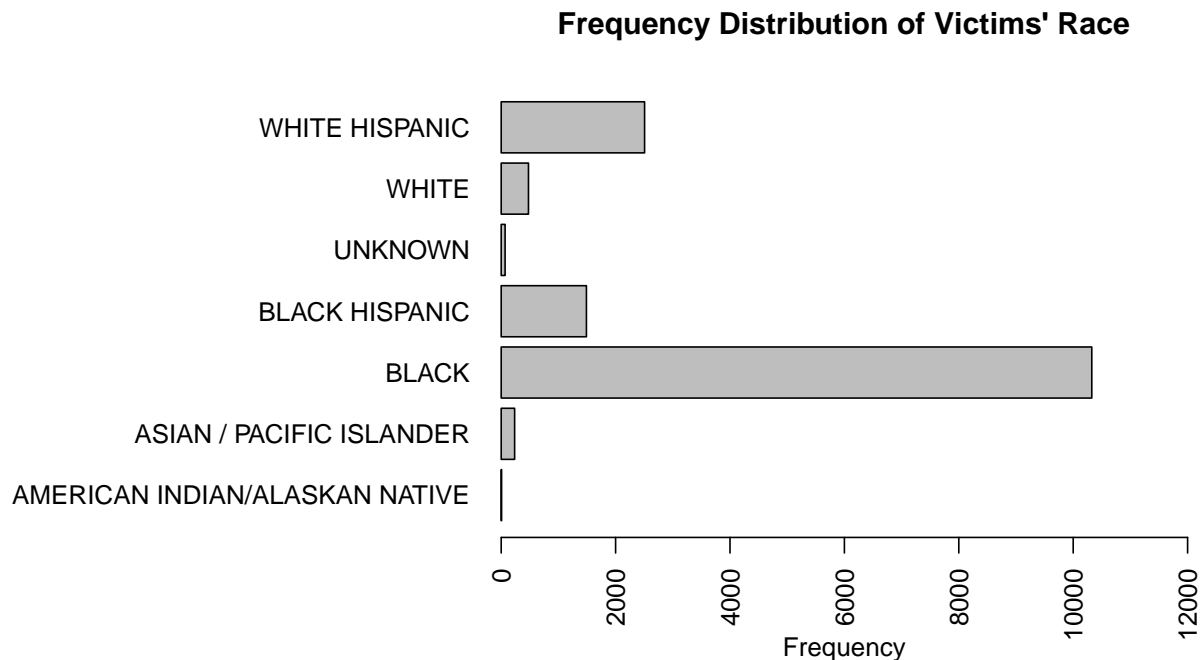


```
v_race <- table(nypd_narm$VIC_RACE)
v_race
```

```
##
## AMERICAN INDIAN/ALASKAN NATIVE      ASIAN / PACIFIC ISLANDER
##                                7                                235
##                                BLACK                                BLACK HISPANIC
##                                10325                               1490
##                                UNKNOWN                                WHITE
```

```
##                68                477
##            WHITE HISPANIC
##                2507
```

```
par(mar = c(4,16,4,2))
barplot(v_race,xlab = "Frequency", main = "Frequency Distribution of Victims' Race", xlim = c(0,12000),
```



Out of the shooting incidents that we analyzed, both victims and perpetrators tend to be blacks. It is not evident the underlying causes for the disproportion in race.

Finally, we want to determine if the occasions of murder defined by a victim's death is dependent on the borough in New York City. We can perform the chi-square test of independence to examine whether the two categorical variables are dependent. The counts are assumed to be independent representing each separate shooting incident. We will construct a contingency table to further examine our remaining condition prior to conducting the chi-square test of independence.

```
boro_murder <- table(nYPD_c$STATISTICAL_MURDER_FLAG, nYPD_c$BORO)
boro_murder
```

```
##
##      BRONX BROOKLYN MANHATTAN QUEENS STATEN ISLAND
## FALSE  5456    7830      2409   2830        555
##  TRUE   1244    1892       512    697        143
```

Since the expected frequencies for each cell is greater than 5, all conditions are satisfied for us to conduct a chi-square test with an alpha level of 0.05.

```
chisq.test(boro_murder)
```

```
##
```

```
## Pearson's Chi-squared test
##
## data:  boro_murder
## X-squared = 8.5598, df = 4, p-value = 0.0731
```

With 4 degrees of freedom, the test statistic for the chi-square test of independence is 8.5598 with a p-value of 0.0731. Since the p-value is greater than our alpha level, we fail to reject our null hypothesis. Therefore, the two variables are independent.

We conclude that Brooklyn is the most dangerous borough in New York City with the highest frequency of shooting incidents. It is likely that the suspects and victims are similar in age who belong to the 18-24 as well as 25-44 age groups. Black males are the predominant group that are involved in these shooting incidents. However, the location as categorized by boroughs are independent from the instances of shooting incidents resulting in murder.

Possible sources of bias may stem from the fact that the analyst is not from the United States. It is hard for the analyst to determine external factors such as cultural, political, sociological that can explain the frequency distribution of the shooting incidents. Even though black males are identified to be most likely to be linked to shooting incidents in New York City, population density data is not examined in this report and can be further explored to further ascertain the relationships between the variables.

Please find below the session info for this document:

```
sessionInfo()
```

```
## R version 4.1.0 (2021-05-18)
## Platform: x86_64-w64-mingw32/x64 (64-bit)
## Running under: Windows 10 x64 (build 19042)
##
## Matrix products: default
##
## locale:
## [1] LC_COLLATE=English_Canada.1252  LC_CTYPE=English_Canada.1252
## [3] LC_MONETARY=English_Canada.1252 LC_NUMERIC=C
## [5] LC_TIME=English_Canada.1252
##
## attached base packages:
## [1] stats      graphics  grDevices  utils      datasets  methods    base
##
## other attached packages:
## [1] lubridate_1.7.10 forcats_0.5.1  stringr_1.4.0  dplyr_1.0.7
## [5] purrr_0.3.4      readr_1.4.0    tidyr_1.1.3    tibble_3.1.2
## [9] ggplot2_3.3.4    tidyverse_1.3.1
##
## loaded via a namespace (and not attached):
## [1] tidymodels_1.1.1  xfun_0.23      haven_2.4.1    colorspace_2.0-1
## [5] vctrs_0.3.8       generics_0.1.0  htmltools_0.5.1.1  yaml_2.2.1
## [9] utf8_1.2.1        rlang_0.4.11    pillar_1.6.1     glue_1.4.2
## [13] withr_2.4.2       DBI_1.1.1       dbplyr_2.1.1     modelr_0.1.8
## [17] readxl_1.3.1      lifecycle_1.0.0  munsell_0.5.0    gtable_0.3.0
## [21] cellranger_1.1.0  rvest_1.0.0     evaluate_0.14     knitr_1.33
## [25] curl_4.3.1        fansi_0.5.0     highr_0.9        broom_0.7.7
## [29] Rcpp_1.0.6        scales_1.1.1    backports_1.2.1   jsonlite_1.7.2
## [33] fs_1.5.0          hms_1.1.0       digest_0.6.27     stringi_1.6.1
## [37] grid_4.1.0        cli_2.5.0       tools_4.1.0      magrittr_2.0.1
```

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## [41] crayon_1.4.1      pkgconfig_2.0.3    ellipsis_0.3.2     xml2_1.3.2
## [45] reprex_2.0.0      assertthat_0.2.1   rmarkdown_2.9      httr_1.4.2
## [49] rstudioapi_0.13   R6_2.5.0           compiler_4.1.0
```