NYPD_Shooting

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First, read in the NYPD Shooting incident data. The CSV file is located in ./data (use setwd to set your working directory to the same folder as this Rmd file is located) and the original file can be downloaded at https://catalog.data.gov/dataset/nypd-shooting-incident-data-historic.

You will need tidyverse and lubridate install.packages("tidyverse") install.packages("lubridate") library(tidyverse) library(lubridate)

```
shooting_data <- read_csv("./data/NYPD_Shooting_Incident_Data__Historic_.csv")</pre>
```

Now eliminate INCIDENT_KEY and all columns after VIC_RACE and convert OCCUR_DATE to a date data type

```
shooting_data <- shooting_data %>%
select(OCCUR_DATE:VIC_RACE) %>%
mutate(OCCUR_DATE = mdy(OCCUR_DATE))
```

Show summary of the data

summary(shooting_data)

```
OCCUR TIME
##
      OCCUR DATE
                                                 BORO
                                                                   PRECINCT
##
           :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
##
           :2020-12-31
                                                                        :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
           :2.0000
   Max.
   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
##
##
##
##
```

The visualizations I will be using do not require any filtering of missing values, but if it did I could do it with:

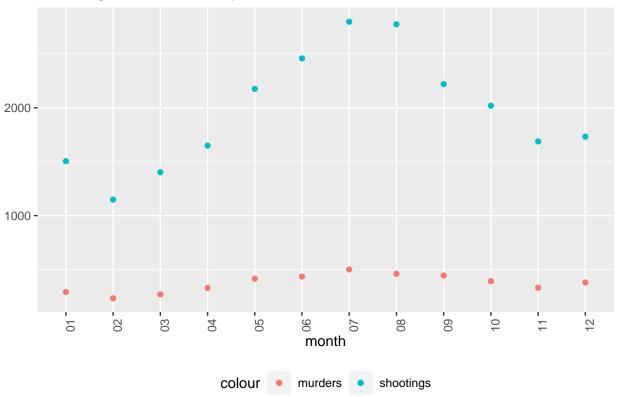
```
shooting_data_no_missing <- shooting_data %>%
filter(PERP_AGE_GROUP != "NA" & PERP_AGE_GROUP != "UNKNOWN" & PERP_SEX != "NA" &
    PERP_RACE != "NA" & VIC_AGE_GROUP != "NA" & VIC_AGE_GROUP != "UNKNOWN" &
    VIC_SEX != "NA" & VIC_RACE != "NA")
```

Group the data by month for both murders and shootings for the first visualization

```
month_group <- shooting_data %>% group_by(month=format(floor_date(OCCUR_DATE, "month"), format="%m")) %
  summarize(murders = sum(STATISTICAL_MURDER_FLAG, na.rm = TRUE), shootings = length(STATISTICAL_MURDER
  ungroup()
```

First visualization - Shootings and Murders by Month





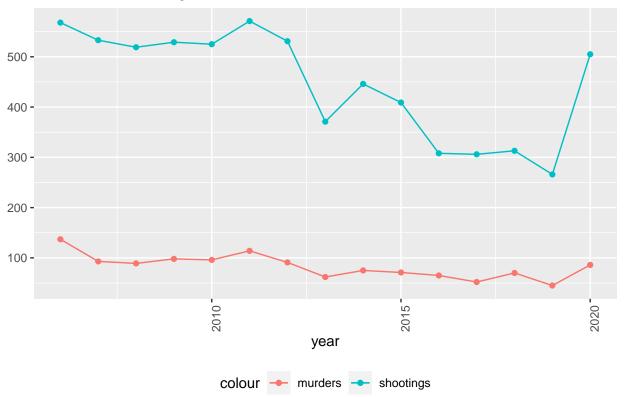
Group the data by borough and year for murder and shootings for the second visualization

```
boro_group <- shooting_data %>% group_by(BORO, year=year(OCCUR_DATE)) %>%
   summarize(murders = sum(STATISTICAL_MURDER_FLAG, na.rm = TRUE), shootings = length(STATISTICAL_MURDER
   select(BORO, year, murders, shootings) %>%
   ungroup()
```

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

Second visualization: Murders and shootings by year for Bronx





Find the three boroughs with the highest murder counts since 2010

```
boro_group %>% filter(year >= 2010) %>% group_by(BORO) %>% summarize(murders = sum(murders)) %>% slice_markets
```

Bias identification: At first I was very interested in seeing how race and age might play out in these shooting incidents, but then realized how fraught with biases both of these were, both my own and in the race identifications available in the data as well as the very broad age groupings that were used.

So to avoid these biases both my own and in the data, I looked only at murders and shootings as they relate to time, either month of the year or year over year. The exception to this is the analysis of the boroughs with the highest number of murders. One might think that Manhattan is a safer place from this, but instead it could be that most murders happen in the evenings and Manhattan has more businesses than residences. To find out if this could be biasing the results would require further research and data.