

## Week3 NY SHOOTING REPORT

Data imported from “<https://data.cityofnewyork.us/api/views/833y-fsy8/rows.csv>”

### Introduction

We have here a table of 23,585 shooting incidents in New York occurring from 2006 to 2020. Data are from government publications.

Even if both information are available, for this report I will focus on the victim rather than the perpetrator. This is a slightly biased report in that I choose to consider the victim side, but this will have no impact on the outcome of the analysis.

```
library(tidyverse)

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

## v ggplot2 3.3.5      v purrr 0.3.4
## v tibble 3.1.6       v dplyr 1.0.8
## v tidyr 1.2.0        v stringr 1.4.0
## v readr 2.1.2        v forcats 0.5.1

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()

data_collected = read_csv("https://data.cityofnewyork.us/api/views/833y-fsy8/rows.csv")

## Rows: 23585 Columns: 19

## -- Column specification -----
## Delimiter: ","
## chr  (10): OCCUR_DATE, BORO, LOCATION_DESC, PERP_AGE_GROUP, PERP_SEX, PERP_R...
## dbl  (7): INCIDENT_KEY, PRECINCT, JURISDICTION_CODE, 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.

head(data_collected)

## # A tibble: 6 x 19
##   INCIDENT_KEY OCCUR_DATE OCCUR_TIME BORO      PRECINCT JURISDICTION_CODE
##   <dbl> <chr>      <time>    <chr>      <dbl>      <dbl>
## 1 24050482 08/27/2006 05:35    BRONX      52          0
## 2 77673979 03/11/2011 12:03    QUEENS     106         0
## 3 203350417 10/06/2019 01:09    BROOKLYN   77          0
## 4 80584527 09/04/2011 03:35    BRONX      40          0
## 5 90843766 05/27/2013 21:16    QUEENS     100         0
## 6 92393427 09/01/2013 04:17    BROOKLYN   67          0
## # ... with 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>
```

```
summary(data_collected)
```

```
## INCIDENT_KEY OCCUR_DATE OCCUR_TIME BORO
## Min. : 9953245 Length:23585 Length:23585 Length:23585
## 1st Qu.: 55322804 Class :character Class1:hms Class :character
## Median : 83435362 Mode :character Class2:difftime Mode :character
## Mean :102280741 Mode :numeric
## 3rd Qu.:150911774
## Max. :230611229
##
## PRECINCT JURISDICTION_CODE LOCATION_DESC STATISTICAL_MURDER_FLAG
## Min. : 1.00 Min. :0.000 Length:23585 Mode :logical
## 1st Qu.: 44.00 1st Qu.:0.000 Class :character FALSE:19085
## Median : 69.00 Median :0.000 Mode :character TRUE :4500
## Mean : 66.21 Mean :0.333
## 3rd Qu.: 81.00 3rd Qu.:0.000
## Max. :123.00 Max. :2.000
## NA's :2
## PERP_AGE_GROUP PERP_SEX PERP_RACE VIC_AGE_GROUP
## Length:23585 Length:23585 Length:23585 Length:23585
## 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:23585 Length:23585 Min. : 914928 Min. :125757
## Class :character Class :character 1st Qu.: 999925 1st Qu.:182539
## Mode :character Mode :character Median :1007654 Median :193470
## Mean :1009379 Mean :207300
## 3rd Qu.:1016782 3rd Qu.:239163
## Max. :1066815 Max. :271128
##
## Latitude Longitude Lon_Lat
## Min. :40.51 Min. : -74.25 Length:23585
## 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
##
```

## Comments

The data are pretty clean, there are no inconvenient “NAs” or missing useful values for our analyses.

Let’s select variables for our analyses. We will focus on the victims rather than the perpetrators, the “bias” here is to highlight the likelihood to be a potential victim of shootings in New York.

```
data_collected=data_collected %>% select(INCIDENT_KEY, OCCUR_DATE, OCCUR_TIME,BORO,VIC_AGE_GROUP, VIC_SEX)
data_collected
```

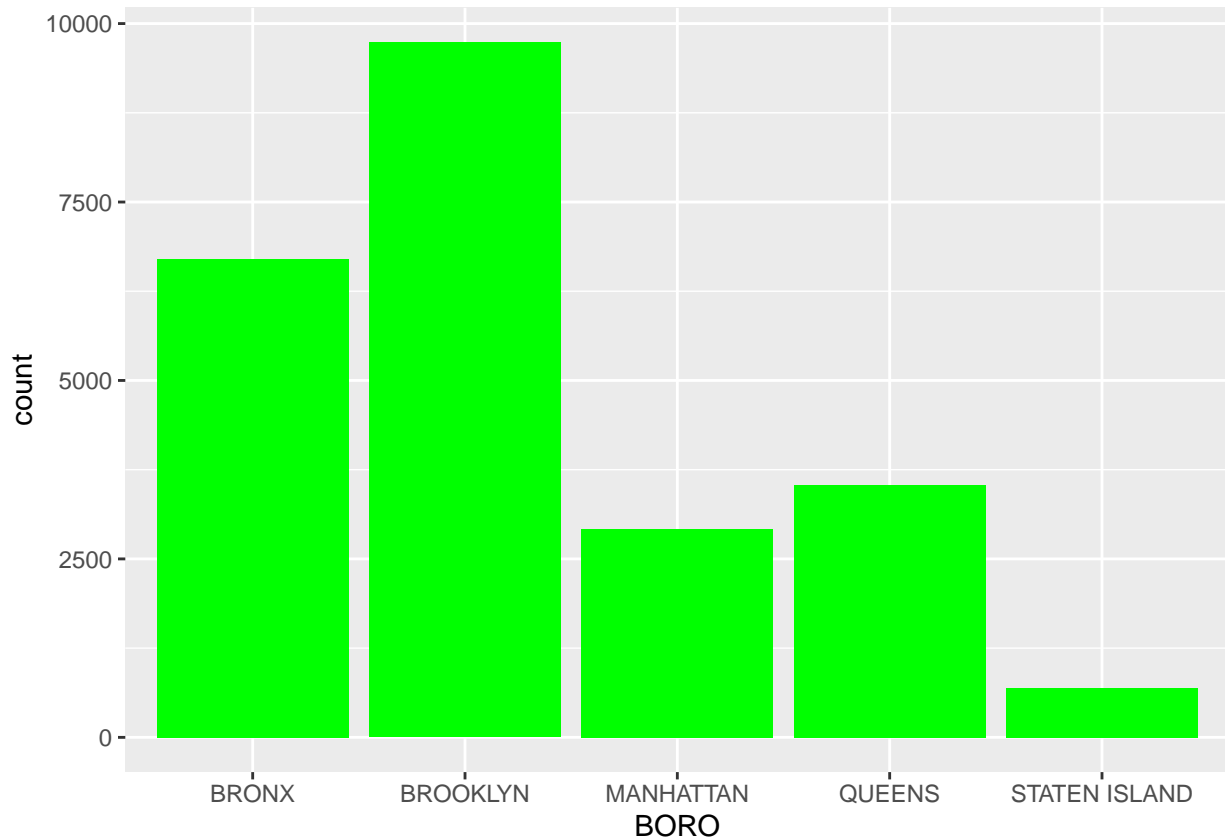
```
## # A tibble: 23,585 x 7
```

```
##      INCIDENT_KEY OCCUR_DATE OCCUR_TIME BORO      VIC_AGE_GROUP VIC_SEX VIC_RACE
##      <dbl> <chr>      <time>    <chr>    <chr>      <chr>  <chr>
## 1      24050482 08/27/2006 05:35    BRONX      25-44      F      BLACK HISP~
## 2      77673979 03/11/2011 12:03    QUEENS     65+        M      WHITE
## 3      203350417 10/06/2019 01:09    BROOKLYN   18-24      F      BLACK
## 4      80584527 09/04/2011 03:35    BRONX      <18        M      BLACK
## 5      90843766 05/27/2013 21:16    QUEENS     18-24      M      BLACK
## 6      92393427 09/01/2013 04:17    BROOKLYN   <18        M      BLACK
## 7      73057167 06/05/2010 21:16    BROOKLYN   <18        M      BLACK
## 8      211362213 03/20/2020 21:27    BROOKLYN   25-44      M      BLACK
## 9      137564752 07/04/2014 00:25    QUEENS     18-24      M      BLACK
## 10     147024011 10/18/2015 01:33    QUEENS     18-24      M      BLACK
## # ... with 23,575 more rows
```

## A) Analysis of general trends

1) Number of victims per borough

```
ggplot(data_collected,aes(x=BORO))+geom_bar(position="stack", fill="green")
```



Let's import some population data from another government website to further compare these numbers with the population levels.

Source = [https://www1.nyc.gov/assets/planning/download/pdf/planning-level/nyc-population/projections\\_briefing\\_booklet.pdf](https://www1.nyc.gov/assets/planning/download/pdf/planning-level/nyc-population/projections_briefing_booklet.pdf)

There are no significant increase in the populations from 2005 to 2020, so we can use 2020 population for each borough as a reference.

*BRONX* = 1,420,277  
*BROOKLYN* = 2,628,211  
*MANHATTAN* = 1,729,530  
*QUEENS* = 2,396,949  
*STATEN ISLAND* = 517,597  
*TOTAL NEW YORK* = 8,692,564

```
sum(data_collected$BORO=="BRONX")/23585
```

```
## [1] 0.2841213
```

```
sum(data_collected$BORO=="BROOKLYN")/23585
```

```
## [1] 0.4127199
```

```
sum(data_collected$BORO=="MANHATTAN")/23585
```

```
## [1] 0.1238923
```

```
sum(data_collected$BORO=="QUEENS")/23585
```

```
## [1] 0.1497562
```

```
sum(data_collected$BORO=="STATEN ISLAND")/23585
```

```
## [1] 0.02951028
```

Comparing with population ratios

```
1420277/8692564
```

```
## [1] 0.1633899
```

```
2628211/8692564
```

```
## [1] 0.3023516
```

```
1729530/8692564
```

```
## [1] 0.1989666
```

```
2396949/8692564
```

```
## [1] 0.2757471
```

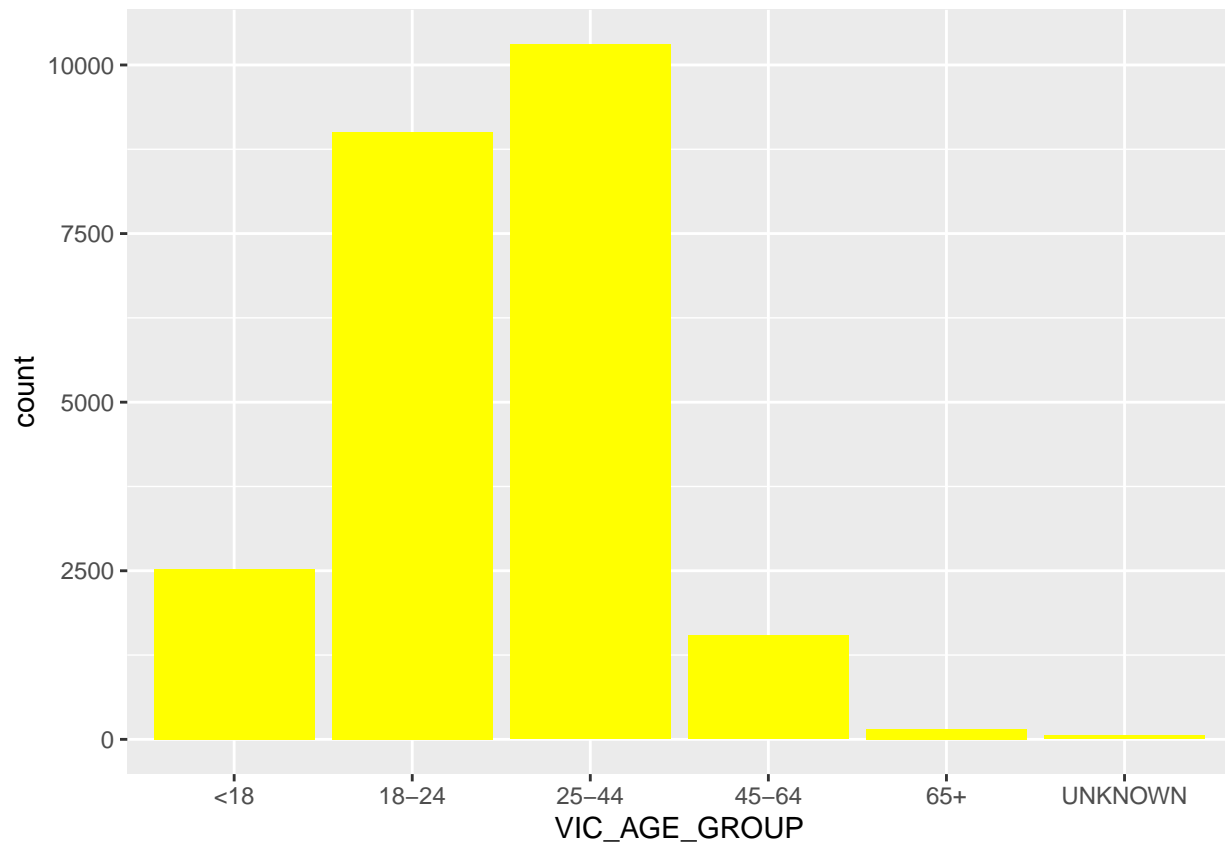
```
517597/8692564
```

```
## [1] 0.0595448
```

### Comments

So the relative crime rates for Brooklyn and also Bronx are clearly higher with respect to the population levels, 41% vs 30% and 28% vs 16%. 2)Number of victims per age category

```
ggplot(data_collected,aes(x=VIC_AGE_GROUP))+geom_bar(position="stack", fill="yellow")
```



### Comments

No surprise there, violent crime victims mostly belong to the young and relative young population in every city.

Let's isolate time, month and year variables.

```
glimpse(data_collected)
```

```
## Rows: 23,585
## Columns: 7
## $ INCIDENT_KEY <dbl> 24050482, 77673979, 203350417, 80584527, 90843766, 92393~
## $ OCCUR_DATE   <chr> "08/27/2006", "03/11/2011", "10/06/2019", "09/04/2011", ~
## $ OCCUR_TIME   <time> 05:35:00, 12:03:00, 01:09:00, 03:35:00, 21:16:00, 04:17~
## $ BORO         <chr> "BRONX", "QUEENS", "BROOKLYN", "BRONX", "QUEENS", "BROOK~
```

```
## $ VIC_AGE_GROUP <chr> "25-44", "65+", "18-24", "<18", "18-24", "<18", "<18", "~
## $ VIC_SEX <chr> "F", "M", "F", "M", "M", "M", "M", "M", "M", "M", "F", "~
## $ VIC_RACE <chr> "BLACK HISPANIC", "WHITE", "BLACK", "BLACK", "BLACK", "B~
```

```
data_collected=data_collected %>% separate(OCCUR_TIME,c("Crime_hour","Crime_min")) %>% separate(OCCUR_D
```

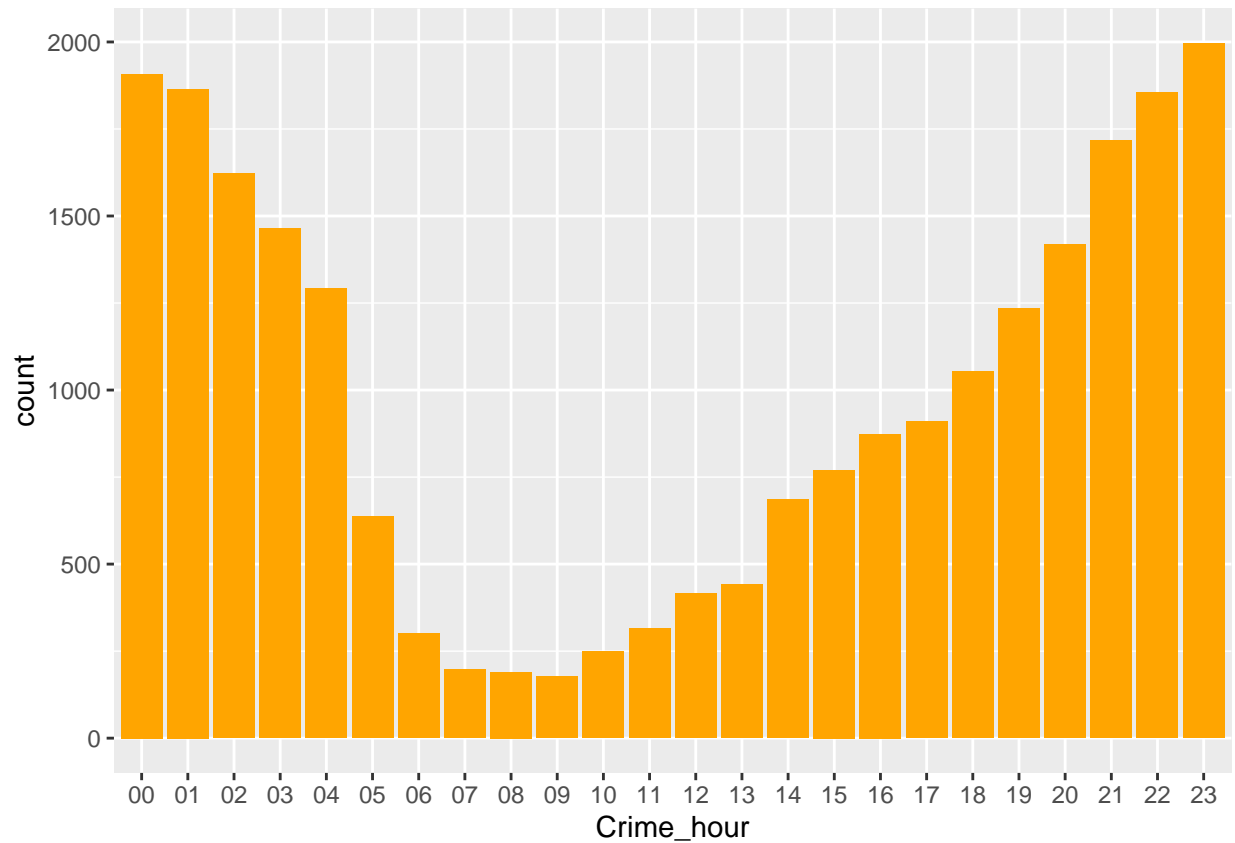
```
## Warning: Expected 2 pieces. Additional pieces discarded in 23585 rows [1, 2, 3,
## 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, ...].
```

```
data_collected
```

```
## # A tibble: 23,585 x 10
##   INCIDENT_KEY Crime_month Crime_day Crime_year Crime_hour Crime_min BORO
##   <dbl> <chr> <chr> <chr> <chr> <chr> <chr>
## 1 24050482 08 27 2006 05 35 BRONX
## 2 77673979 03 11 2011 12 03 QUEENS
## 3 203350417 10 06 2019 01 09 BROOKLYN
## 4 80584527 09 04 2011 03 35 BRONX
## 5 90843766 05 27 2013 21 16 QUEENS
## 6 92393427 09 01 2013 04 17 BROOKLYN
## 7 73057167 06 05 2010 21 16 BROOKLYN
## 8 211362213 03 20 2020 21 27 BROOKLYN
## 9 137564752 07 04 2014 00 25 QUEENS
## 10 147024011 10 18 2015 01 33 QUEENS
## # ... with 23,575 more rows, and 3 more variables: VIC_AGE_GROUP <chr>,
## # VIC_SEX <chr>, VIC_RACE <chr>
```

3)Number of victims per hour

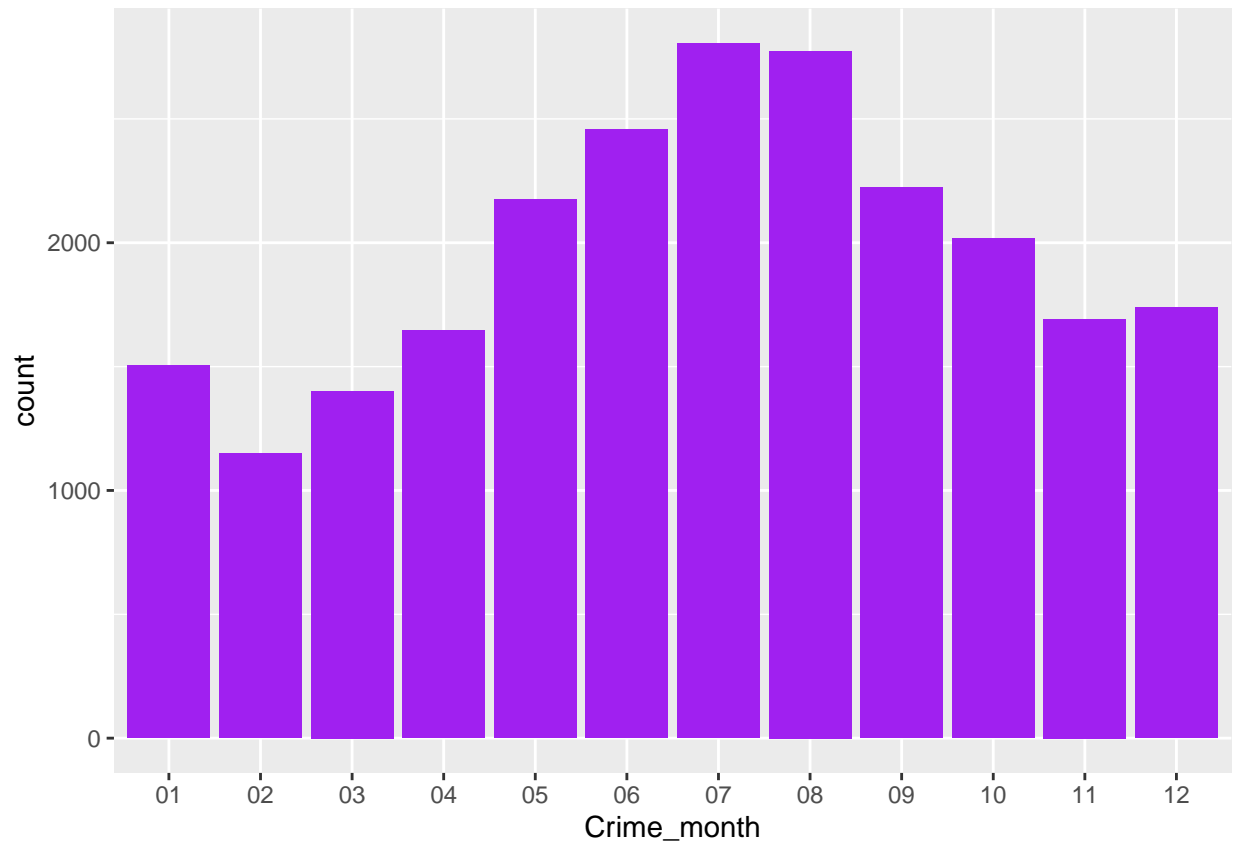
```
ggplot(data_collected,aes(x=Crime_hour))+geom_bar(position="stack", fill="orange")
```



### Comments

The slopes are definitely clear, the shootings increase the later it gets up to 23:00, the “crime ideal time”, then start to decrease as sunrise gets closer. 4)Number of victims per month

```
ggplot(data_collected,aes(x=Crime_month))+geom_bar(position="stack", fill="purple")
```



### Comments

Same as for the day analysis, there is also a clear trend throughout the year.

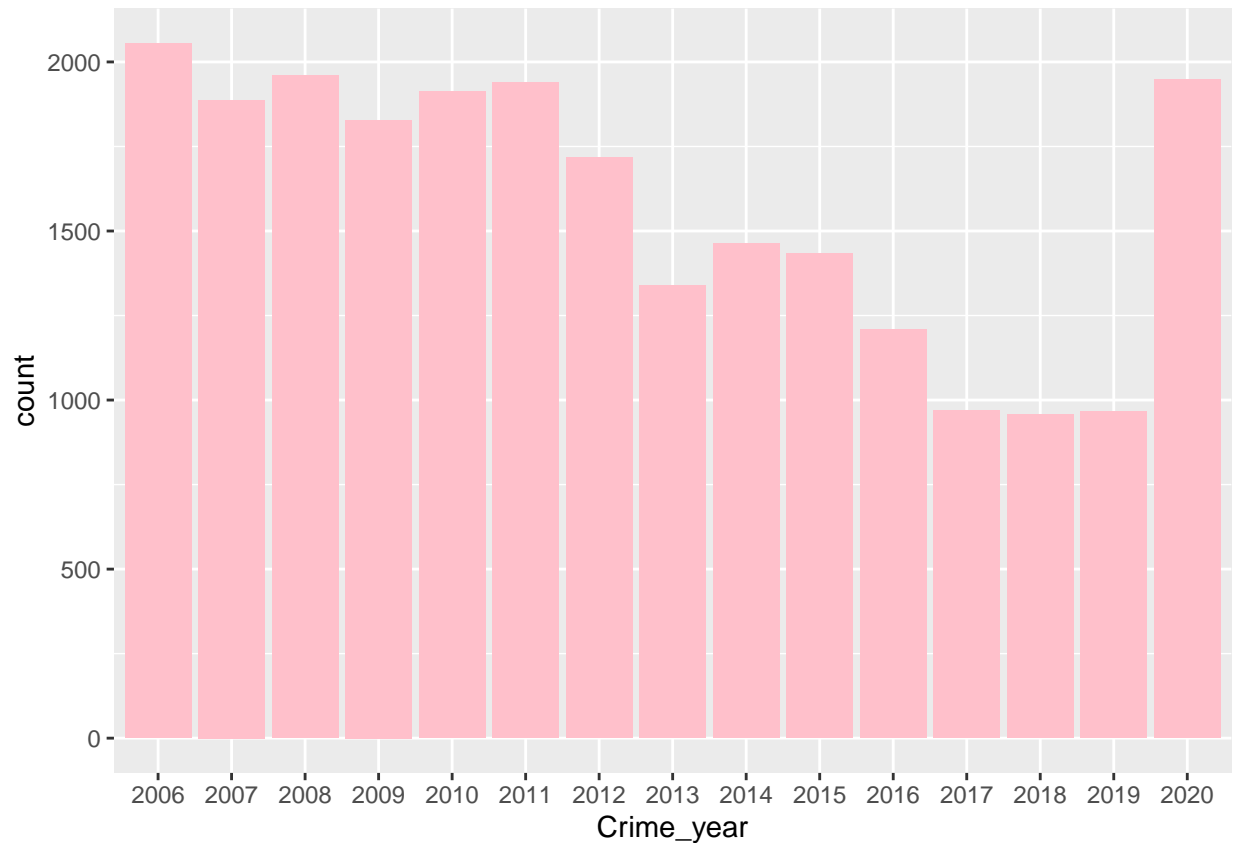
It's surprising to see an increase as the weather gets nicer as if crime rate was boosted by the sunny days!!

This is a “funny” paradox given that the shooting peak is at night time when there is no sun anymore.

5)Number of victims per year

```
ggplot(data_collected,aes(x=Crime_year))+geom_bar(position="stack", fill="pink")
```



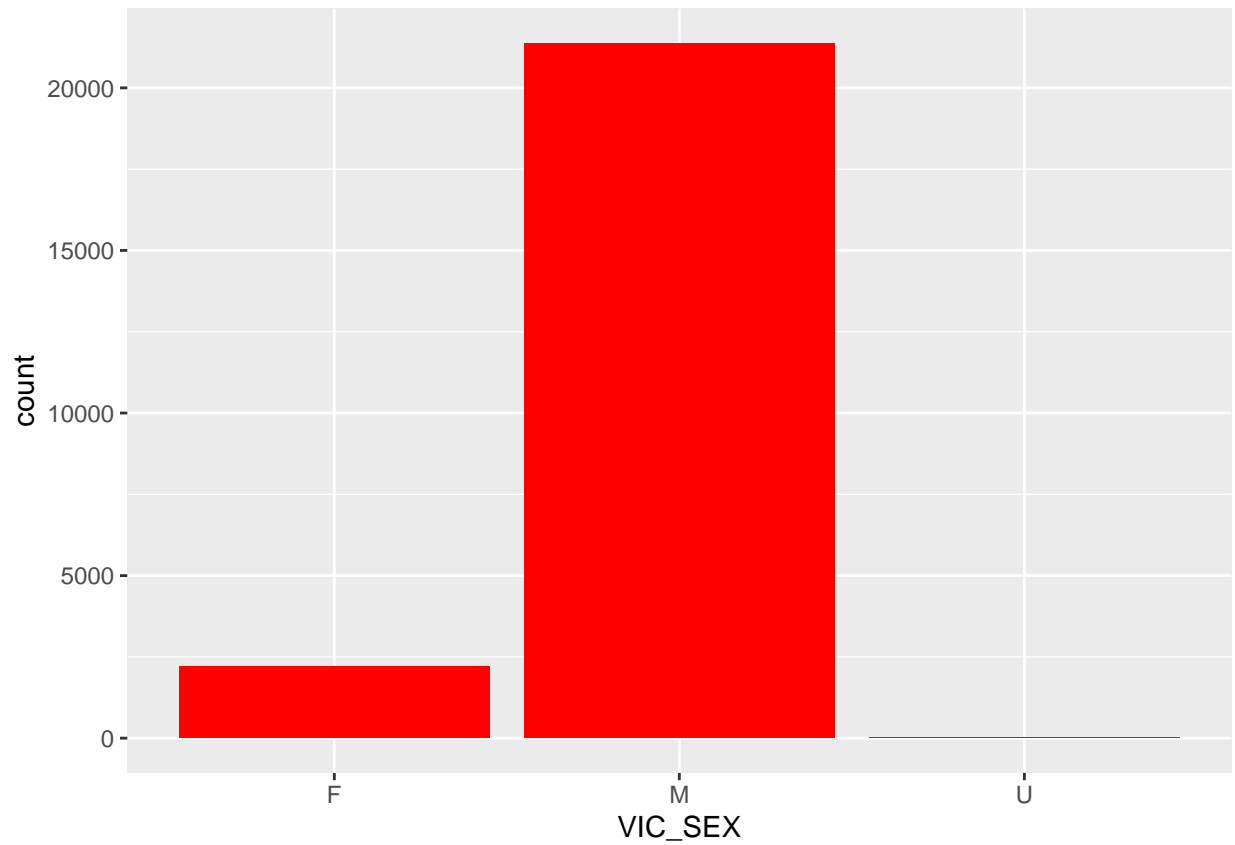


### Comments

It might be interesting to understand the rationale behind the shocking rise in shootings in 2020, the first covid year. We would expect the steady trend from 2017 to 2019 to keep going or even decrease, but we are facing twice as much shootings as during each of all three previous years. Has covid crisis definitely ruined so many years of improvement in violent crime numbers in New York?

6)Number of victims per gender

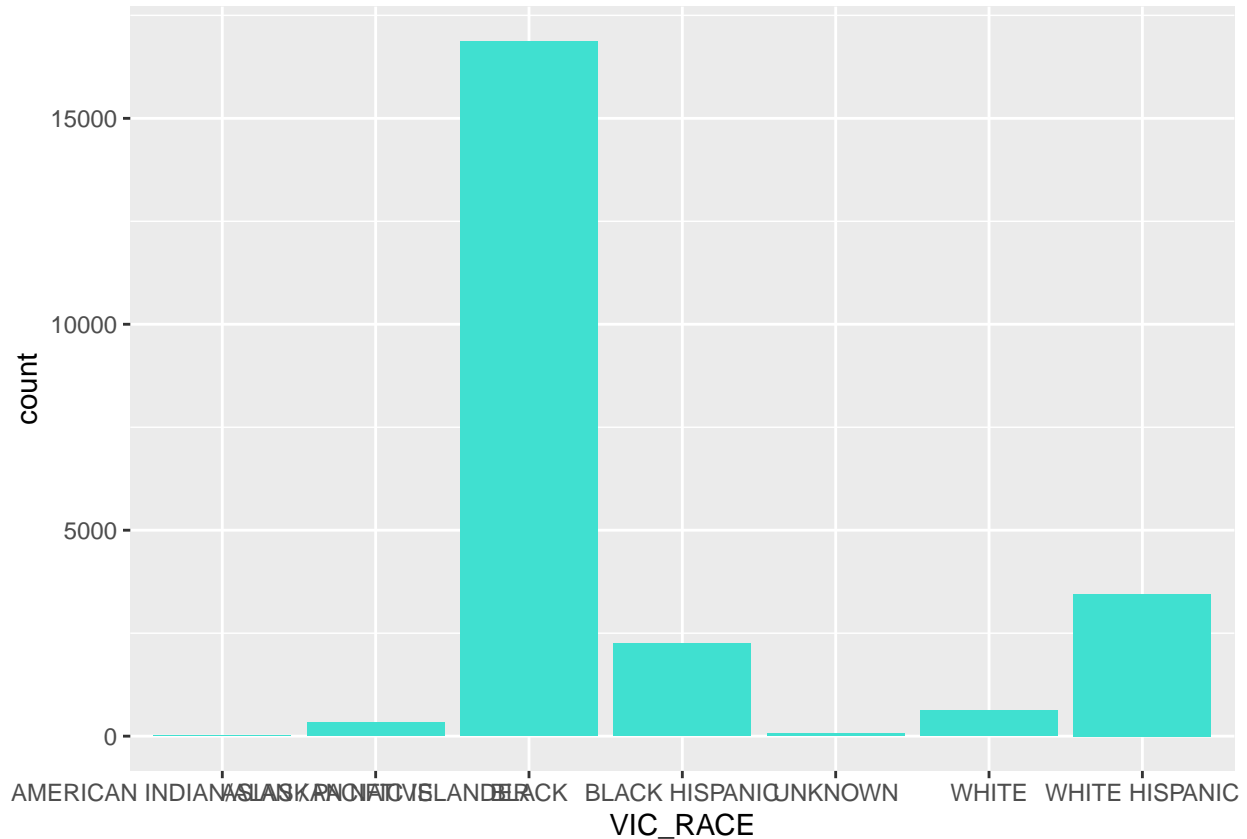
```
ggplot(data_collected,aes(x=VIC_SEX))+geom_bar(position="stack", fill="red")
```



#### Comments

No surprise there, violent crime rate has always been higher within male population. 7) Number of victims per race

```
ggplot(data_collected,aes(x=VIC_RACE))+geom_bar(position="stack", fill="turquoise")
```



### Comments

The highest potential victim is a black man, aged between 25 and 44, living in or visiting Brooklyn, at night time specifically around 23:00 in July. Can we then safely conclude that any man meeting those race and age criterion should avoid at all costs Brooklyn in summer at night time? Additional variables such as the circumstances of the shootings might be necessary to further conclusions. **B ) Further analysis for Brooklyn**

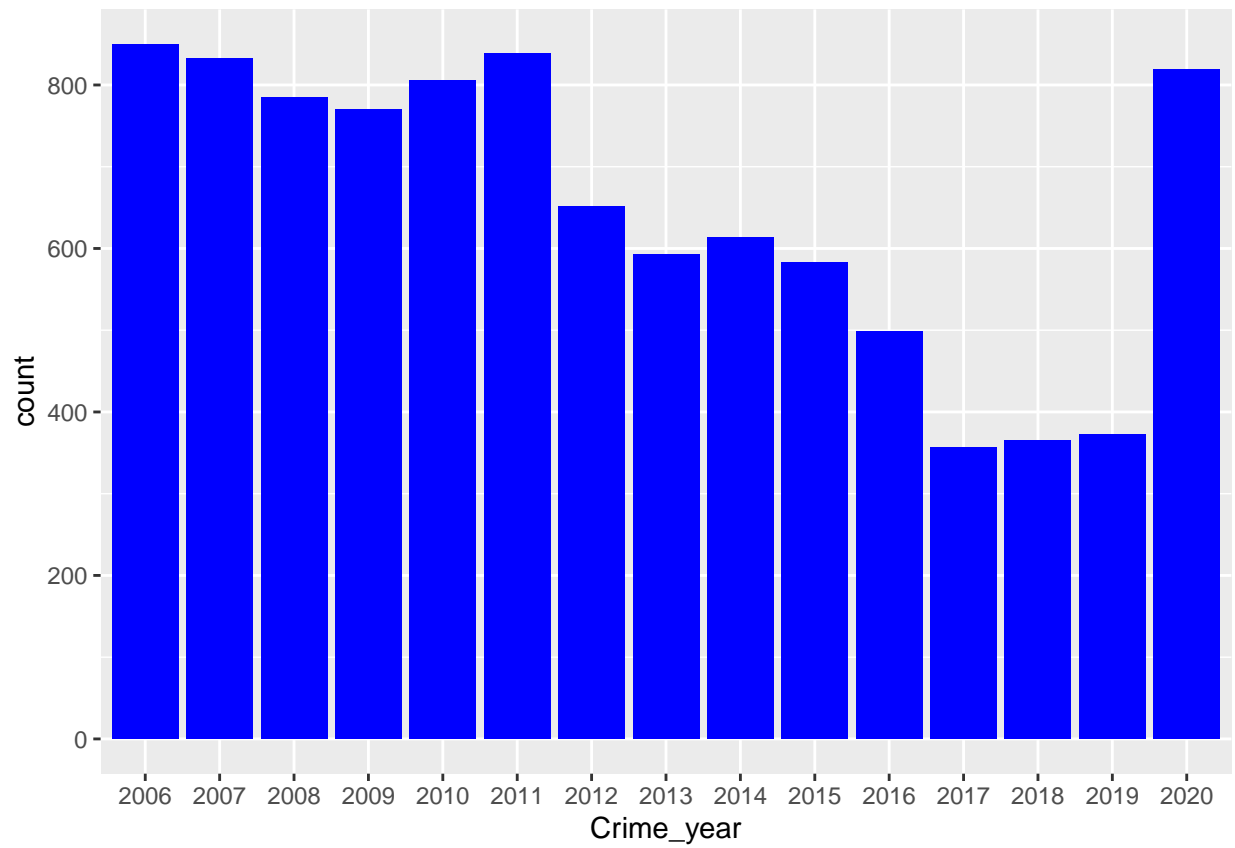
Since Brooklyn won the award of the absolute shooting cases numbers and also was outstanding in the relative shooting cases numbers with regards to population levels, let's do a deeper analysis for this borough.

```
data_Brooklyn = data_collected %>% filter(data_collected$BORO=="BROOKLYN")
head(data_Brooklyn)
```

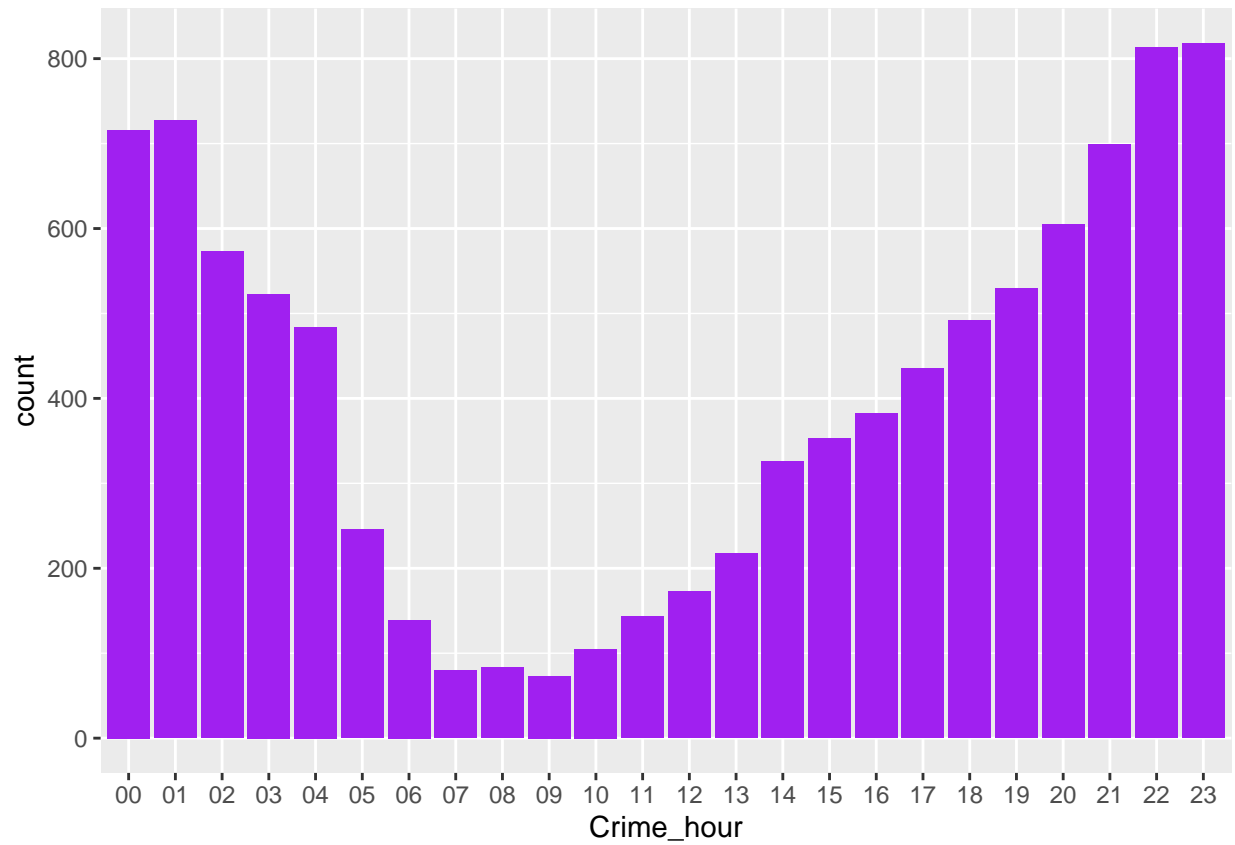
```
## # A tibble: 6 x 10
##   INCIDENT_KEY Crime_month Crime_day Crime_year Crime_hour Crime_min BORO
##   <dbl> <chr> <chr> <chr> <chr> <chr> <chr>
## 1 203350417 10 06 2019 01 09 BROOKLYN
## 2 92393427 09 01 2013 04 17 BROOKLYN
## 3 73057167 06 05 2010 21 16 BROOKLYN
## 4 211362213 03 20 2020 21 27 BROOKLYN
## 5 82333894 12 26 2011 03 00 BROOKLYN
## 6 214693508 06 27 2020 00 35 BROOKLYN
## # ... with 3 more variables: VIC_AGE_GROUP <chr>, VIC_SEX <chr>, VIC_RACE <chr>
```

- 1) Brooklyn shooting cases trend from 2006 to 2020
- 2) Brooklyn shooting cases day trend
- 3) Brooklyn shooting cases month trend

```
ggplot(data_Brooklyn,aes(x=Crime_year))+geom_bar(position="stack", fill="blue")
```



```
ggplot(data_Brooklyn,aes(x=Crime_hour))+geom_bar(position="stack", fill="purple")
```



```
ggplot(data_Brooklyn,aes(x=Crime_month))+geom_bar(position="stack", fill="pink")
```



### Conclusion

I chose to focus on the likelihood to be victim of a shooting in New York because I was interested in knowing whether it was a safe place to visit. But the variables here are not sufficient for a thorough analysis. It would be interesting to know the circumstances of the shootings, meaning whether they occurred within drug trafficking or in the middle of the city etc.

We see that the trends for Brooklyn with respect to day, month and year are the same as those for total population. We do not need to run the plots for gender and race, as they will certainly have the same trends as for total New York population. It might be interesting to see if these trends are the same on the country level.