NYPD Shooting Incidents Project Delivery

Learner

2023-01-24

Introdution:

We are asked to apply the Data Science Process we have learned from lectures of Dr. J Wall, we have New York City Shooting Incidents Data Set from year 2006 up to year 2021. I am going to add my comments describing each code under the 4 Steps as follows:

Step 1: Start an Rmd Document.

Start an Rmd document that describes and imports the shooting project dataset in a reproducible manner

library(tidyverse)

```
\section{table}
— Attaching packages

    tidyverse

1.3.2 -
√ ggplot2 3.4.0
                     ✓ purrr
                                1.0.1
✓ tibble 3.1.8

√ dplyr 1.0.10

√ tidyr 1.3.0

                     ✓ stringr 1.5.0

√ readr 2.1.3

                     \checkmark forcats 0.5.2
 Conflicts
tidyverse conflicts() —
X dplyr::filter() masks stats::filter()
X dplyr::lag() masks stats::lag()
library(tinytex)
library(lubridate)
\section{table}
Attaching package: 'lubridate'
The following objects are masked from 'package:base':
    date, intersect, setdiff, union
```

url <- "https://data.cityofnewyork.us/api/views/833yfsy8/rows.csv?accessType=DOWNLOAD"

nypd <- read csv(url)</pre>

nypd

message.

\section{table} # A tibble: 25,596 × 19 INCIDENT ... 1 OCCUR... 2 OCCUR... 3 BORO PRECI... 4 JURIS... 5 LOCAT... 6 STATI... 7 PERP ... 8 PERP_... PERP_... VIC_A... VIC_SEX <dbl> <chr> <time> <chr> <dbl> <dbl> <chr> <lgl> <chr>> <chr> <chr> <chr> <chr>> 79 236168668 11/11/... 15:04 0 NA FALSE BR00... NA 1 NA 18-24 NA 2 231008085 07/16/... 22:05 72 0 NA 45-64 BR00... FALSE Μ ASIAN ... 25-44 3 230717903 07/11/... 01:09 BR00... 79 0 NA FALSE <18 BLACK 25-44 Μ Μ 4 237712309 12/11/... 13:42 BR00... 81 0 NA FALSE NA NA NA 25-44 5 224465521 02/16/... 20:00 OUEE... 113 0 NA FALSE NA NA 25-44 NA TRUE 6 228252164 05/15/... 04:13 QUEE... 113 0 NA NA 25-44 NA NA Μ 0 COMMER... TRUE 7 226950018 04/14/... 21:08 BRONX 42 NA NA 18-24 NA Μ 237710987 12/10/... 19:30 8 BRONX 52 0 NA FALSE NA NA 25-44 Μ 9 224701998 02/22/... 00:18 34 0 NA FALSE NA MANH... NA 25-44 NA Μ 75 TRUE 10 225295736 03/07/... 06:15 BR00... 0 NA 25 - 44 BLACK ... 25-44 # ... with 25,586 more rows, 6 more variables: VIC RACE <chr>, X COORD CD

```
<dbl>, Y_COORD_CD <dbl>,
# Latitude <dbl>, Longitude <dbl>, Lon_Lat <chr>, and abbreviated variable
names 'INCIDENT_KEY,
# 'OCCUR_DATE, 'OCCUR_TIME, 'PRECINCT, 'JURISDICTION_CODE, 'LOCATION_DESC,
'STATISTICAL_MURDER_FLAG,
# 'PERP_AGE_GROUP, 'PERP_SEX, 'PERP_RACE, 'VIC_AGE_GROUP
# i Use `print(n = ...)` to see more rows, and `colnames()` to see all
variable names
```

Step 2: Tidy and Transform Your Data

Add to your Rmd document a summary of the data and clean up your dataset by changing appropriate variables to factor and date types and getting rid of any columns not needed. Show the summary of your data to be sure there is no missing data. If there is missing data, describe how you plan to handle it.

summary(nypd)

INCIDENT_KEY PRECINCT	OCCUR_DATE	OCCUR_TIME	BORO
Min. : 9953245 Min. : 1.00	Length:25596	Length:25596	Length:25596
1st Qu.: 61593633 1st Qu.: 44.00	Class :character	Class1:hms	Class :character
Median : 86437258 Median : 69.00	Mode :character	Class2:difftime	Mode :character
Mean :112382648 Mean : 65.87 3rd Qu.:166660833 3rd Qu.: 81.00 Max. :238490103		Mode :numeric	
Max. :123.00			
JURISDICTION_CODE PERP_SEX	LOCATION_DESC	STATISTICAL_MURDER_	_FLAG PERP_AGE_GROUP
Min. :0.0000 Length:25596	Length:25596	Mode :logical	Length:25596
1st Qu.:0.0000 :character Class		FALSE: 20668	Class
Median :0.0000 :character Mode Mean :0.3316 3rd Qu.:0.0000 Max. :2.0000 NA's :2	Mode :character :character	TRUE :4928	Mode
PERP_RACE X_COORD_CD	VIC_AGE_GROUP	VIC_SEX	VIC_RACE
Length: 25596	Length:25596	Length:25596	Length:25596

Min. : 914928

Class :character Class :character Class :character Class :character

1st Qu.:1000011

Mode :character Mode :character Mode :character Mode :character

Median :1007715

Mean :1009455

3rd Qu.:1016838

Max. :1066815

Max. :271128

Y COORD CD Latitude Longitude Lon Lat Length: 25596 Min. :125757 Min. :40.51 Min. :-74.25 1st Qu.:40.67 1st Qu.:-73.94 Class :character 1st Qu.:182782 Median :194038 Median :40.70 Median :-73.92 Mode :character :-73.91 Mean :207894 Mean :40.74 Mean 3rd Qu.:239429 3rd Qu.:40.82 3rd Qu.:-73.88

Max. :40.91

*Removing unwanted columns

nypd = nypd %>% select(-c(INCIDENT_KEY, OCCUR_TIME, PRECINCT,
JURISDICTION_CODE, STATISTICAL_MURDER_FLAG, PERP_RACE, VIC_RACE, X_COORD_CD,
Y_COORD_CD, Latitude, Longitude, Lon_Lat))

Max. :-73.70

summary(nypd)

\section{table}			
OCCUR_DATE PERP_SEX	BORO	LOCATION_DESC	PERP_AGE_GROUP
Length: 25596 Length: 25596	Length:25596	Length:25596	Length:25596
Class :character Class :character	Class :character	Class :character	Class :character
Mode :character Mode :character	Mode :character	Mode :character	Mode :character
VIC_AGE_GROUP Length:25596	VIC_SEX Length:25596		
Class :character Mode :character	Class :character Mode :character		

Changing the "OCCUR_DATE" column data type into date for correct analysis

class(nypd\$OCCUR_DATE) \section{table} [1] "character"
nypd <- nypd %>% mutate (OCCUR_DATE = mdy(OCCUR_DATE))
class(nypd\$OCCUR_DATE)

```
\section{table}
[1] "Date"
```

changing names for some columns to better analysis

```
colnames(nypd)[1] <- "Date" colnames(nypd)[2] <- "Borough" colnames(nypd)[3] <- "place" colnames(nypd)[4] <- "Perp_Age" colnames(nypd)[5] <- "Perp_Gender" colnames(nypd)[6] <- "Vic_Age" colnames(nypd)[7] <- "Vic_Gender"
```

Step 3 Add Visualizations and Analysis

Add at least two different visualizations & some analysis to your Rmd. Does this raise additional questions that you should investigate?

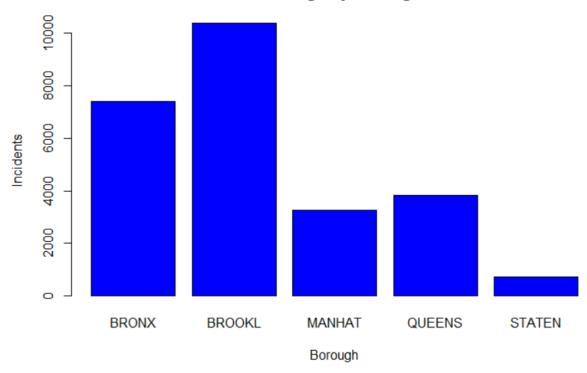
```
nypd_Borough <- nypd %>% count(Borough)
nypd_Borough
```

```
a <- nypd_Borough %>% mutate(across(everything(), str_sub, 1,6))
```

```
a <- a %>% mutate(n=as.numeric(n))
```

barplot(an, names = aBorough, xlab="Borough", ylab="Incidents", main="Shootings By Borough", col="blue")

Shootings By Borough



Analysis Conclusion: Brookl has the highest shooting incidents

```
nypd_Vic_Age <- nypd %>% count(Vic_Age)
```

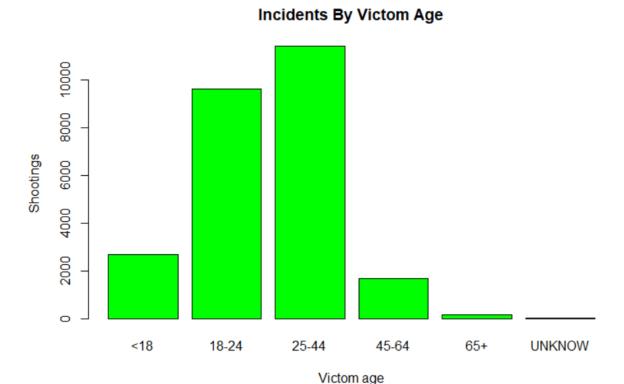
 $nypd_Vic_Age$

```
\section{table}
A tibble: 6 \times 2
 Vic_Age
  <chr>
          <int>
1 <18
           2681
2 18-24
           9604
3 25-44
          11386
4 45-64
           1698
5 65+
            167
6 UNKNOWN
             60
```

b <- nypd_Vic_Age %>% mutate(across(everything(), str_sub, 1,6))

b <- b %>% mutate(n=as.numeric(n))

 $barplot(bn, names = bVic_Age, xlab="Victom age", ylab="Shootings", main="Incidents By Victom Age", col="green")$



Analysis Conclusion: the Age Group 25-44 has the highest shooting incidents

Step 4: Add Bias Identification

Write the conclusion to your project report and include any possible sources of bias. Be sure to identify what your personal bias might be and how you have mitigated that.

This data set has much amount of work to manipulate with, and I did only very basic analysis, also I would like to mention that this project was the first project I make analysis in R, and I am sure that It will be the first step forward to me for being a Data Scientist. To avoid bias I did not like to analyze the data in terms of race.