## Title: DataViz Assignment4: Exploratory Data Visualizations

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This report including five charts which illustrated the relationship between Chicago public school ACT test performance and their drop out rate, as well as the distribution of suspect crime which involve in gang activities and 2016 summer Chicago crime analysis.

```
library(tidyverse)
library(markdown)
library(ggplot2)
library(readr)
library(haven)
library(dplyr)
library(tidyr)
library(stringr)
library(wesanderson)
library(RColorBrewer)
library(ggthemes)
library(plotrix)
library(ggmap)
library(cowplot)
```

### Chart 1: Public School Performance 2012

Data loading and cleaning.

```
cps <- read csv("cps1.csv")</pre>
cps <- cps %>%
  filter(`Overall Rating` != "Not Enough Data")
cps_plot <- ggplot(data = cps, aes(x=`Average Score ACT 2012`, y=`One-Year Dr</pre>
opout Rate 2012 - Percent`, color=`Overall Rating`))+geom point(alpha=0.6, si
ze = 4)+
  scale_y_continuous(breaks=seq(0, 0.3, 0.02))+theme_minimal()+
  scale_color_manual(values=wes_palette(n=3, name="GrandBudapest"))+
  theme(legend.title = element text(size = 9),
        legend.position = c(0.9, 0.8),
        legend.text = element text(size = 9))+
  labs(title = "Public School Performance 2012",
       subtitle = " The level 1 public high schools had higher average ACT
score and lower drop out rate compare with other two levels,
while none of the Level 3 school had average ACT score higher than 20 even so
me of them had drop out rate lower than 0.02.
       caption = "Source:Chicago Public Schools Progress Report", x = "Averag
```

```
e ACT Score", y = "One Year Drop out Rate")+
theme(plot.title = element_text(hjust = 0.5))
```

#### Public School Performance 2012

The level 1 public high schools had higher average ACT score and lower drop out rate compare with oth while none of the Level 3 school had average ACT score higher than 20 even some of them had drop out



Source: Chicago Public Schools Progress Report The full

range of the ACT score for this data set is 0-0.25. As mentioned above, the data showing that public high schools in higher level had higher average ACT score and lower drop out rate, while the schools in lower level had average lower ACT scores **no matter** there drop out rate is.

# **Chart 2: 2016 Chicago Summer Investigatory Stop by Age**

Data loading and cleaning.

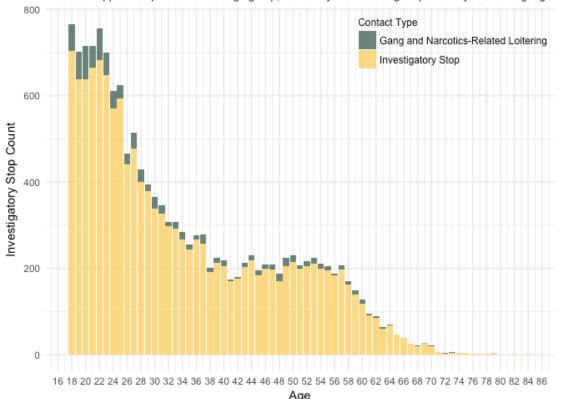
```
isr <- read_csv("isr.csv")

isr_lim <- isr %>% subset(DATE >="7/1/16" & DATE <= "8/31/16")
isr_lim <- isr_lim %>%
    filter(AGE <= 90)
isr_lim <- isr_lim %>%
    add_count(AGE)
isr_lim <- isr_lim %>%
    filter(!is.na(`Contact Type`))
```

```
isr_plot <- ggplot(data = isr_lim, aes(x=AGE, fill=`Contact Type`))+geom_bar(
alpha=0.8)+
    scale_fill_manual(values = wes_palette("Chevalier"))+
    scale_x_continuous(breaks=seq(16, 90, 2)) + theme_minimal()+
    theme(legend.title = element_text(size = 9),
        legend.position = c(0.8, 0.9),
        legend.text = element_text(size = 9))+
labs(title = "2016 Chicago Summer Investigatory Stop by Age",
        subtitle = "The data showing that, from July, 2016 to August, 2016, youn
g people who aged from 18 to 24 are more likely
        to be stopped compare with other age group, while they also have higher
probability involve in gang activities.",
    caption = "Source:Chicago Police Investigatory Stop Reports
    ", x = "Age", y = "Investigatory Stop Count") +
        theme(plot.title = element_text(hjust = 0.5))</pre>
```

#### 2016 Chicago Summer Investigatory Stop by Age

The data showing that, from July, 2016 to August, 2016, young people who aged from 18 to 24 are more li to be stopped compare with other age group, while they also have higher probability involve in gang ac



Source: Chicago Police Investigatory Stop Reports

In this

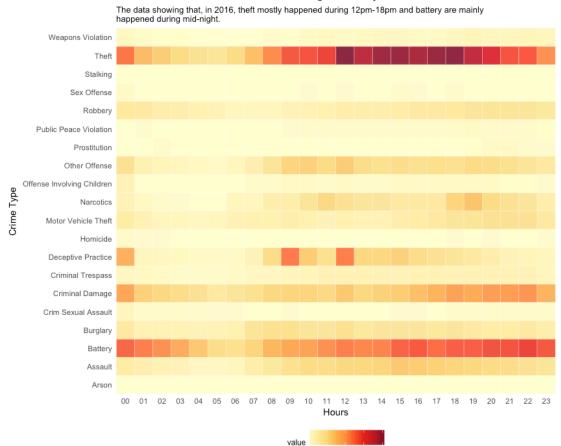
chart, only the investigatory stop data been studied while the legend shows weather the suspects involve in gang activities. It's pretty obvious that young people aged from 18-24 had higher possibility to be stopped while they also had higher likelihood to involving in **gang activities**.

# **Chart 3: 2016 Chicago Crime by Hours**

Data loading and cleaning.

```
cri lim <- read csv("cri lim trim.csv")</pre>
cri_lim <- arrange(cri_lim,desc(value))</pre>
cri_plot<- ggplot(cri_lim, aes(x=Hour, y=`Primary Type`, fill= value)) + geom</pre>
_tile(colour = "white",alpha=0.9) +
  theme_minimal() +
  scale_fill_gradientn(colours = brewer.pal(9, 'YlOrRd'))+
  scale_x_discrete(expand = c(0, 0)) +
  scale y discrete(expand = c(0, 0)) +
  coord_equal() +
  theme(legend.position = "bottom", legend.direction = "horizontal",
              legend.box = "horizontal",
        legend.title = element text(size = 9),
        legend.text = element text(size = 9))+
theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank())
  labs(title = "2016 Chicago Crime by Hours",
       subtitle = "The data showing that, in 2016, theft mostly happened duri
ng 12pm-18pm and battery are mainly
happened during mid-night.",
       caption = "Source:Chicago Data Portal
         ", x = \text{"Hours"}, y = \text{"Crime Type"})+
       theme(plot.title = element_text(hjust = 0.5))
```

#### 2016 Chicago Crime by Hours



Source: Chicago Data Portal

This

chart shows the most frequently appeared crime incidents during 24 hours in the summer of 2016. The dark red color indicating the highest volume while the light yellow shows the lowest volume. From the chart, we can tell that theft had significantly higher incidence rate compare with other crime types and it's more likely happening **during afternoon**. Battery had a similar tendency while deceptive practice more likely to appear at morning.

1000 2000 3000 4000

# **Chart 4: Comparison between Investigation Stop and Crime Offense by District**

```
cri <- read_csv("cri.csv")

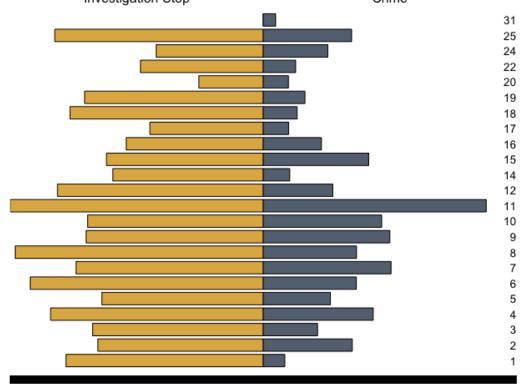
cri <- cri %>% subset(Date >="1/1/16" & Date <= "12/31/16")
isr <- isr %>% subset(DATE >="1/1/16" & DATE <= "12/31/16")

cri_n <-cri %>%
    group_by(District) %>%
    summarise(IS = n())

isr_n <- isr %>%
    group_by(DISTRICT) %>%
```

```
summarise(Cr = n())
names(cri n)[names(cri n)=="District"] <- "DISTRICT"</pre>
cri_n <- cri_n %>%
  filter(!is.na(DISTRICT))
isr_n <- isr_n %>%
  filter(!is.na(DISTRICT))
join <- left join(cri n, isr n, by = "DISTRICT")</pre>
Investigation_Stop <- join$IS</pre>
Crime <- join$Cr</pre>
District <-join$DISTRICT</pre>
par(cex=0.85,bg="transparent")
par(mar=pyramid.plot(Investigation_Stop, Crime, top.labels=c("Investigation S")
top", "", "Crime"), labels = District, main = "Comparison between Investigation St
op and Crime Offense by District
Source: Chicago Police Investigatory Stop Reports", lxcol="#d9a744", rxcol="#5
75e70", unit="", gap=0 ))
```





5392 4652 3912 3172 2432 1692 999 394 0 471 1143 1883 2623 3363 4103 4843

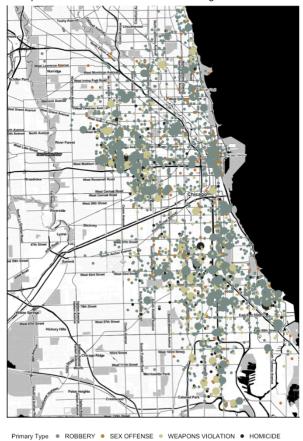
This chart compared the frequency of investigatory stop with the frequency of real crime been found among districts. District 31 is the only one that investigatory stop less than crime incidents, while district 11 had a pretty close rate in both side. The police officers in district 1 had relatively high stop checking rate, compare with others.

# **Chart 5: 2016 Summer Chicago Severe Crime Map**

```
cri offense<- cri %>% subset(Date >="7/1/16" & Date <= "8/31/16")</pre>
cri offense$Location [cri offense$Location == ''] <- NA</pre>
cri offense<- na.omit(cri offense)</pre>
cri offense$Longitude <- round(as.double(cri offense$Longitude), 3)</pre>
cri_offense$Latitude <- round(as.double(cri_offense$Latitude), 3)</pre>
cri offense <- cri offense %>% filter(`Primary Type` =="WEAPONS VIOLATION"|`
Primary Type` =="SEX OFFENSE"|`Primary Type` =="HO
MICIDE")
cri offense$`Primary Type` <- factor(cri offense$`Primary Type`, levels = c("</pre>
ROBBERY", "SEX OFFENSE", "WEAPONS VIOLATION", "HOMICIDE"))
cri offense<- cri offense %>%
  group_by(`Longitude`,`Latitude`,`Primary Type`) %>%
  summarise(TOTAL = n())
chicago \leftarrow get stamenmap(bbox = c(left = -87.885169, bottom=41.643919,
                                    right = -87.523984, top = 42.023022),
                             zoom=12,maptype="toner")
chicago <- ggmap(chicago, extent ="device")</pre>
map<- chicago+geom point( data = cri offense, aes(x = Longitude, y = Latitude</pre>
, color = Primary Type , size=TOTAL), alpha=1)+guides(size=FALSE)+ scale_color
_manual(values=wes_palette(n=4, name="Moonrise2")) +
  theme(legend.position = "bottom", legend.direction = "horizontal",
        legend.box = "horizontal",
        legend.title = element text(size = 8),
        legend.text = element text(size = 8))+
theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank())
  labs(title = "2016 Summer Chicago Severe Crime Map",
       subtitle = "The data showing that the residents live in suburb and
  south area had higher possibility involve in homicide
  compare with downtown and north area during summer.",
       caption = "Source:Chicago Data Portal")+
       theme(plot.title = element text(hjust = 0.5))
```

#### 2016 Summer Chicago Severe Crime Map

The data showing that the residents live in suburb and south area had higher possibility involve in homicide compare with downtown and north area during summer.



Source:Chicago Data Portal

This map is focus on the severe crime incidents appeared during summer 2016 at Chicago. The point size indicating the frequency of the incidents while the four crime types been denoted with four different colors. It's easy to tell that robbery and weapons violence are roughly spread around the whole city, while homicide incidents are more likely to appear in **southern** part of the city.