

Chicago Crimes Final

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```
rm(list = ls())

library(stringr)
library(EnvStats)
library(ggpubr)
library(ggplot2)
library(reshape2)
# library(sm)

# df <- read.csv('C:/Users/krtfe/Downloads/Crimes_-_2023-Updated.csv')
df <- read.csv("C:/Users/krtfe/Downloads/Crimes_-_2023 (ret. 082023).csv")

# simplify data, remove columns that aren't useful for current project
df <- df[,c(3, 6, 7:10, 12:14)]

# removed columns so the data could be imported to github
# write.csv(df, 'C:/Users/krtfe/Downloads/Crimes_-_2023-8-20.csv')

# remove rows with NA values, removed 685 rows 132001 -> 131999
df <- na.omit(df)

# remove duplicate rows
df <- dplyr::distinct(df)

# adding useful columns, dates, times, time of day
dates <- str_split(df$Date, pattern = ' ', simplify = TRUE)[,1]
times <- str_split(df$Date, pattern = ' ', simplify = TRUE)[,2]
time_of_day <- str_split(df$Date, pattern = ' ', simplify = TRUE)[,3]

# add the useful columns and transform data types
df['Date'] <- as.Date(dates, format = '%m/%d/%Y')
df['Time'] <- times
df['Time of Day'] <- time_of_day

# set dataframe such that it only includes months from january to july
df <- df[df$Date < lubridate::ymd("2023-08-01"),]

df %>% head
```

##	Date	Primary.Type	Description	Location.Description	Arrest
## 1	2023-06-28	HOMICIDE FIRST DEGREE MURDER		ALLEY	true

```
## 2 2023-06-29      HOMICIDE FIRST DEGREE MURDER      STREET false
## 3 2023-03-30 CRIMINAL DAMAGE      TO PROPERTY      GAS STATION false
## 4 2023-03-07      THEFT      FROM BUILDING      RESIDENCE false
## 5 2023-06-29      HOMICIDE FIRST DEGREE MURDER      STREET false
## 6 2023-06-29      HOMICIDE FIRST DEGREE MURDER      STREET false
## Domestic District Ward Community.Area      Time Time of Day
## 1      false      17      33      16 11:04:00      PM
## 2      false      7      6      68 07:40:00      PM
## 3      false      1      4      32 02:16:00      PM
## 4      false      3      20      42 10:57:00      AM
## 5      false      8      14      57 07:00:00      AM
## 6      false      7      16      67 04:39:00      PM
```

Separate data frame for counts by dates and specific variables

```
# second data frame, number of crimes

# start with the unique dates and their counts
numCrimes <- table(df$Date)
dfCounts <- data.frame(numCrimes)
colnames(dfCounts) <- c('Date', 'Number of Crimes')

# make row names the dates, for convenience
row.names(dfCounts) <- dfCounts$Date

# add a count by each date for domestic crimes
for (i in dfCounts$Date) {
  dfCounts[i, 'Domestic'] <- sum((df$Date == i & df$Domestic == 'true'))
}

# add a count by each date for crimes with arrests
for (i in dfCounts$Date) {
  dfCounts[i, 'Arrest'] <- sum((df$Date == i & df$Arrest == 'true'))
}

# table of main types of crimes
tabType <- table(df$Primary.Type)

# top types of primary types of crimes
topTypes = sort(tabType, decreasing = TRUE)[1:5]
ttLabels = labels(topTypes)[[1]]

# columns for the counts for the top types of primary types of crimes
for (j in ttLabels) {
  for (i in dfCounts$Date) {
    dfCounts[i, j] <- sum((df$Date == i & df$Primary.Type == j))
  }
}

# renaming column names for consistency
```

```
colnames(dfCounts) <- str_to_title(colnames(dfCounts))
ttLabels = str_to_title(ttLabels)
```

```
# printing the counts dataset
dfCounts %>% head
```

```
##           Date Number Of Crimes Domestic Arrest Theft Battery
## 2023-01-01 2023-01-01           969      237    115    124     206
## 2023-01-02 2023-01-02           648      134     77    110     103
## 2023-01-03 2023-01-03           730       97     67    143      91
## 2023-01-04 2023-01-04           680      107     84    148      81
## 2023-01-05 2023-01-05           654      110     83    141      92
## 2023-01-06 2023-01-06           722      113     88    136      87
##           Criminal Damage Motor Vehicle Theft Assault
## 2023-01-01           159           87      91
## 2023-01-02           99           87      45
## 2023-01-03          130           98      52
## 2023-01-04           66          111      53
## 2023-01-05           75           89      37
## 2023-01-06           90           88      51
```

EDA by arrests, domestic, crime, ward, community area

```
dateCounts <- table(df$Date)
meanD <- sum(dateCounts)/length(dateCounts)
variance <- sum((meanD - dateCounts)^2)/length(dateCounts)

# output variance and mean information about the crime counts
cat(paste('Number of Crimes by Day:',
          '\n\tMean = ', round(meanD, 5),
          '\n\tVariance = ', round(var(dateCounts), 5)))
```

```
## Number of Crimes by Day:
## Mean = 694.24528
## Variance = 3436.94429
```

```
# outputting variance and mean information
for (i in c(colnames(dfCounts)[3:length(colnames(dfCounts))])) {
  meanCounts <- round(mean(unlist(dfCounts[i])), 5)
  varCounts <- round(var(unlist(dfCounts[i])), 5)

  if (i %in% c('Domestic', 'Criminal Damage', 'Battery')) {
    cat(paste('\n\nNumber of ', i, ' Crimes by Day:',
              '\n\tMean = ', meanCounts,
              '\n\tVariance = ', varCounts,
              sep = ''))
  }
  else {
    cat(paste('\n\nNumber of ', i, 's by Day:',
```

```

        '\n\tMean = ', meanCounts,
        '\n\tVariance = ', varCounts,
        sep = ''))
    }
}

```

```

##
##
## Number of Domestic Crimes by Day:
## Mean = 120.97642
## Variance = 454.16532
##
## Number of Arrests by Day:
## Mean = 81.58019
## Variance = 121.25894
##
## Number of Thefts by Day:
## Mean = 148.21226
## Variance = 399.88364
##
## Number of Battery Crimes by Day:
## Mean = 118.77358
## Variance = 484.67835
##
## Number of Criminal Damage Crimes by Day:
## Mean = 80.80189
## Variance = 279.06957
##
## Number of Motor Vehicle Thefts by Day:
## Mean = 80.56604
## Variance = 189.95296
##
## Number of Assaults by Day:
## Mean = 60.40566
## Variance = 104.17589

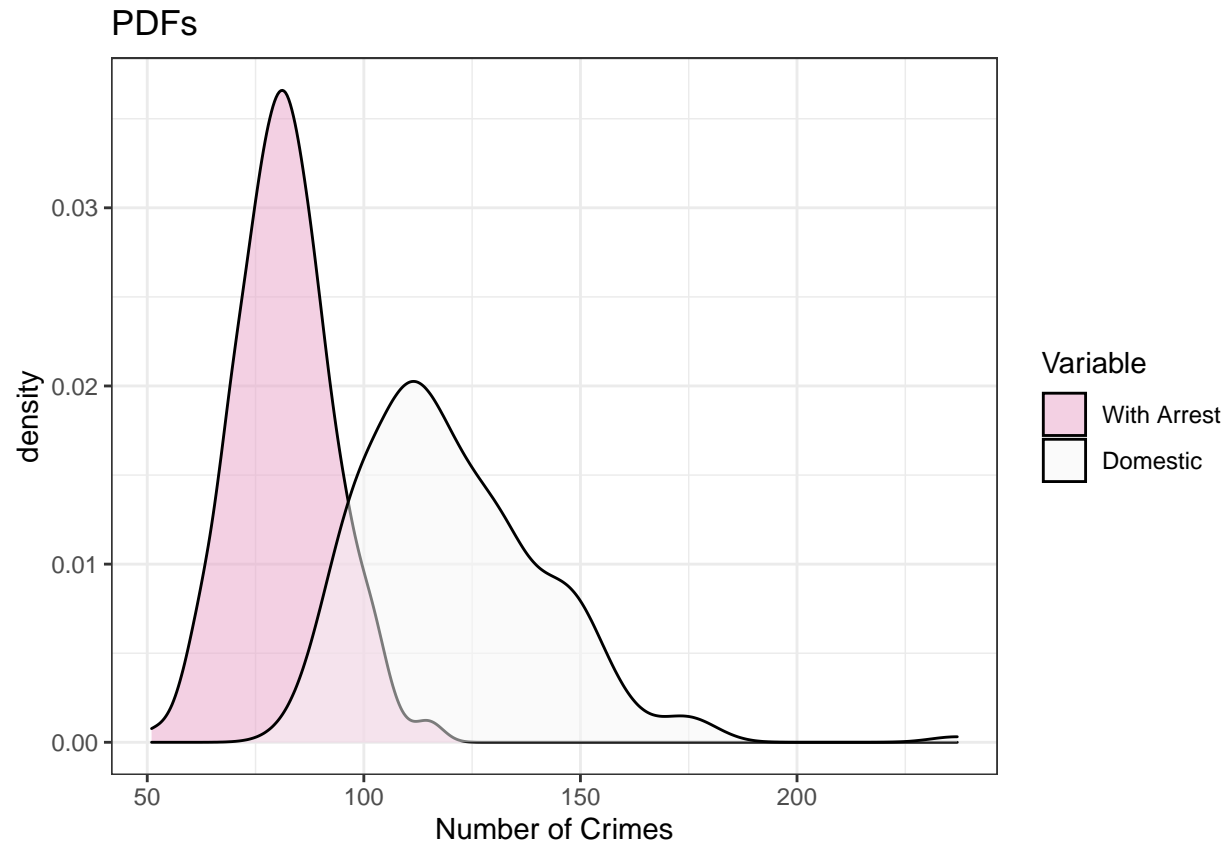
```

```

meltedDens <- melt(dfCounts[c('Arrest',
                             'Domestic')])

# pdf for the arrest and domestic count columns
ggplot(meltedDens, aes(x = value, fill = variable)) +
  geom_density(alpha = 0.5, adjust = 1) +
  # xlim(c(0, 50)) +
  xlab('Number of Crimes') +
  scale_fill_brewer('Variable', palette = 'PiYG',
                    labels = c('With Arrest',
                              'Domestic')) +
  theme_bw() +
  ggtitle('PDFs')

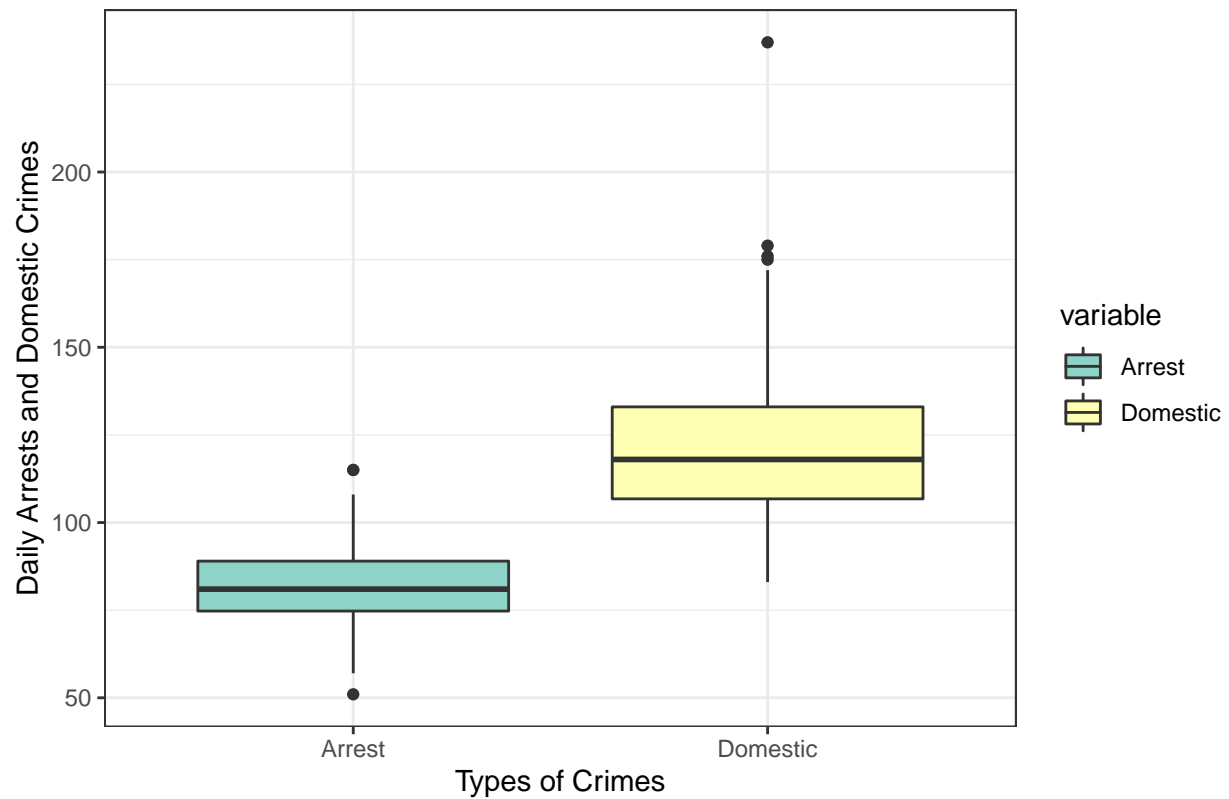
```



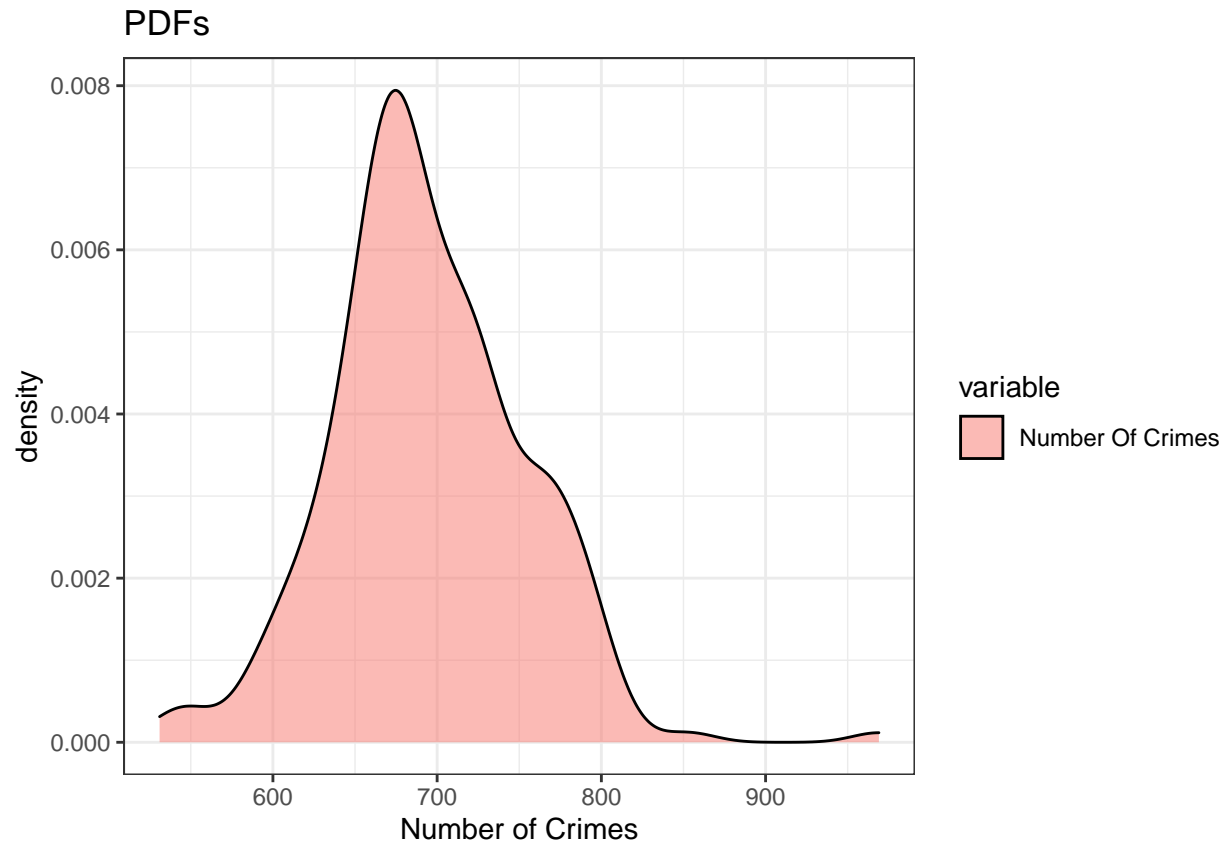
```
# boxplot for the arrest and domestic count columns
aplot <- ggplot(meltedDens,
  aes(x = variable, y = value, fill = variable),
) + geom_boxplot()

aplot +
  scale_fill_brewer(palette="Set3") +
  ylab('Daily Arrests and Domestic Crimes') +
  ggtitle('Quantile Plot, Number of Daily Domestic Crimes and Arrests') +
  scale_x_discrete(name = 'Types of Crimes',
    limits = c('Arrest', 'Domestic') ) +
  theme_bw()
```

Quantile Plot, Number of Daily Domestic Crimes and Arrests

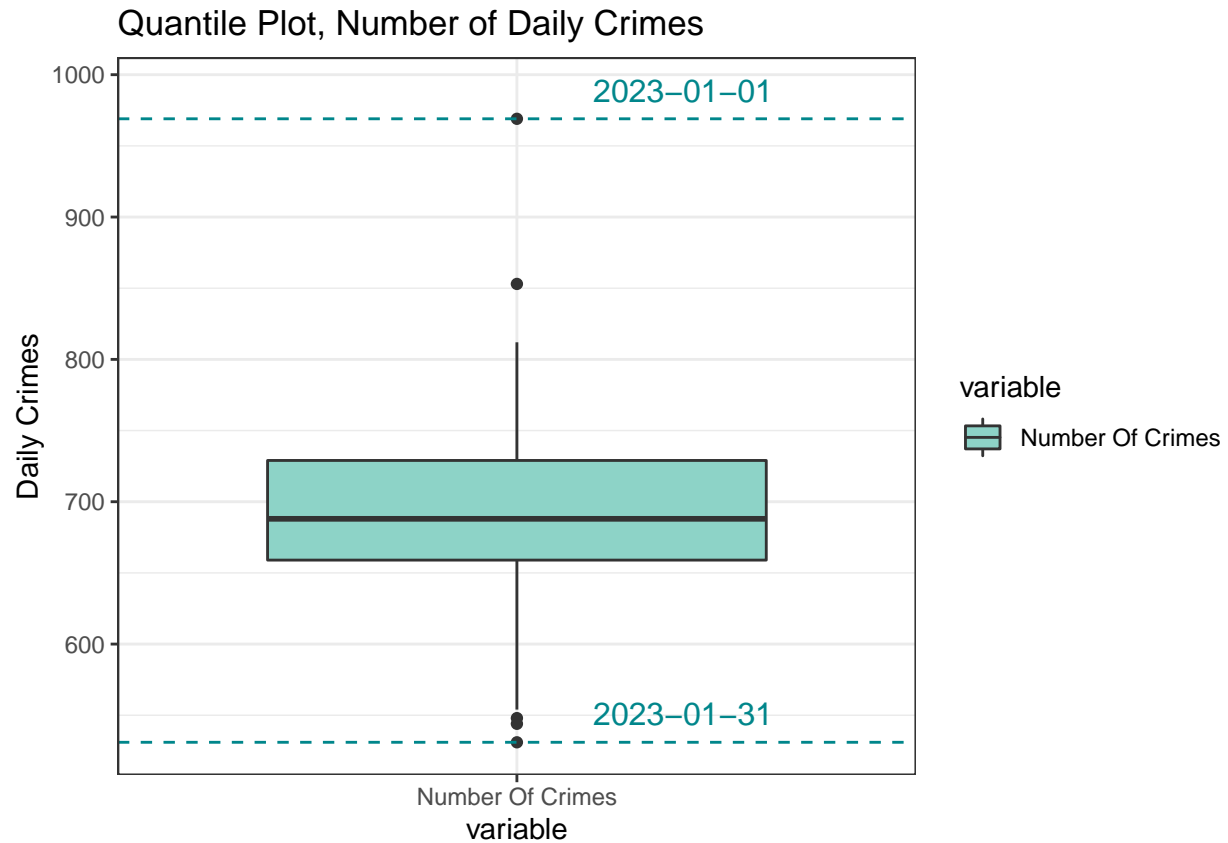


```
# pdf for the number of crimes counts
meltedDens <- melt(dfCounts[c('Number Of Crimes')])
ggplot(meltedDens, aes(x = value, fill = variable)) +
  geom_density(alpha = 0.5, adjust = 1) +
  xlab('Number of Crimes') +
  theme_bw() +
  ggtitle('PDFs')
```



```
# boxplot for the number of crimes counts
aplot <- ggplot(meltedDens,
  aes(x = variable, y = value, fill = variable),
) + geom_boxplot()

aplot +
  scale_fill_brewer(palette="Set3") +
  ylab('Daily Crimes') +
  ggtitle('Quantile Plot, Number of Daily Crimes') +
  geom_hline(yintercept = max(dfCounts$`Number Of Crimes`), linetype = 'dashed', color = 'turquoise4') +
  annotate(geom = 'text',
    label = dfCounts[dfCounts$`Number Of Crimes` == max(dfCounts$`Number Of Crimes`), ]$Date, size = 12,
    color = 'turquoise4', x = 1.25, y = max(dfCounts$`Number Of Crimes`)+20) +
  geom_hline(yintercept = min(dfCounts$`Number Of Crimes`), linetype = 'dashed', color = 'turquoise4') +
  annotate(geom = 'text',
    label = dfCounts[dfCounts$`Number Of Crimes` == min(dfCounts$`Number Of Crimes`), ]$Date, size = 12,
    color = 'turquoise4', x = 1.25, y = min(dfCounts$`Number Of Crimes`)+20) +
  # scale_x_discrete(name = 'Types of Crimes',
  #   limits = c('Domestic', 'Arrest') ) +
  theme_bw()
```

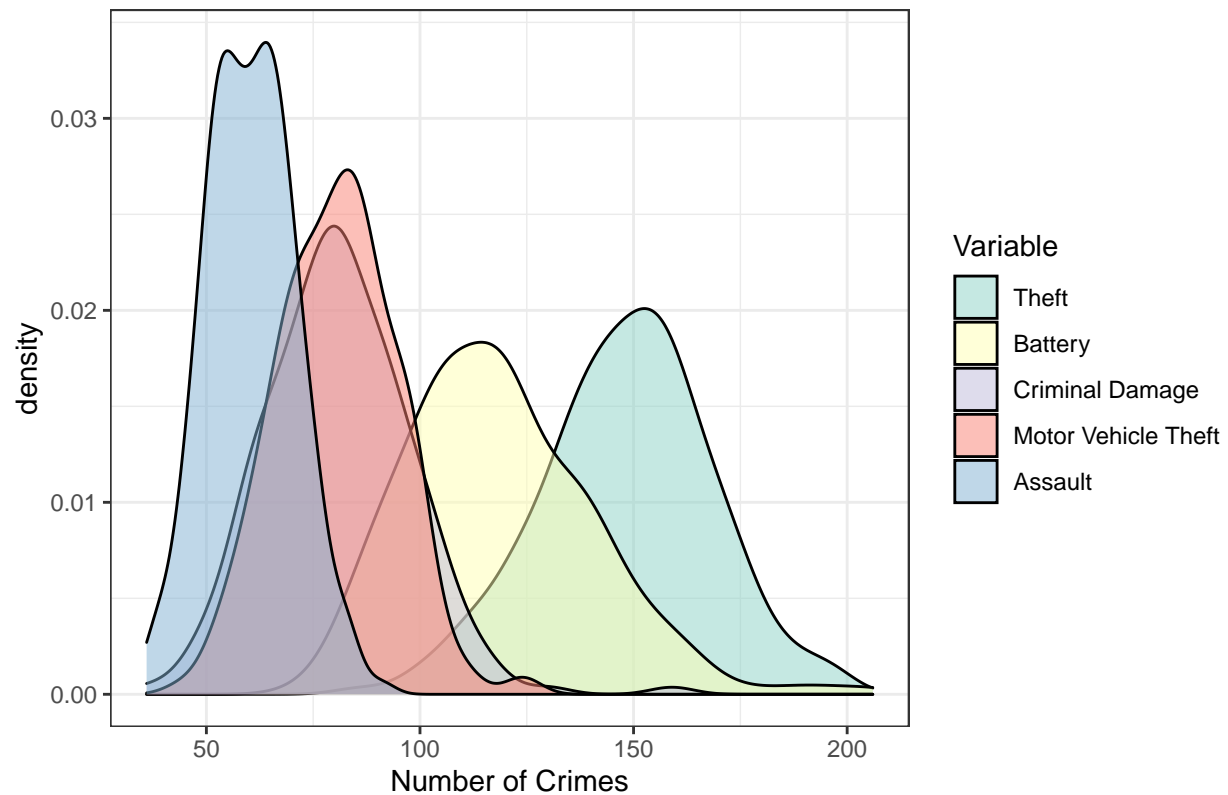


Visualizations for the most common types of crimes in the dataset

```
meltedDens <- melt(dfCounts[ttLabels])

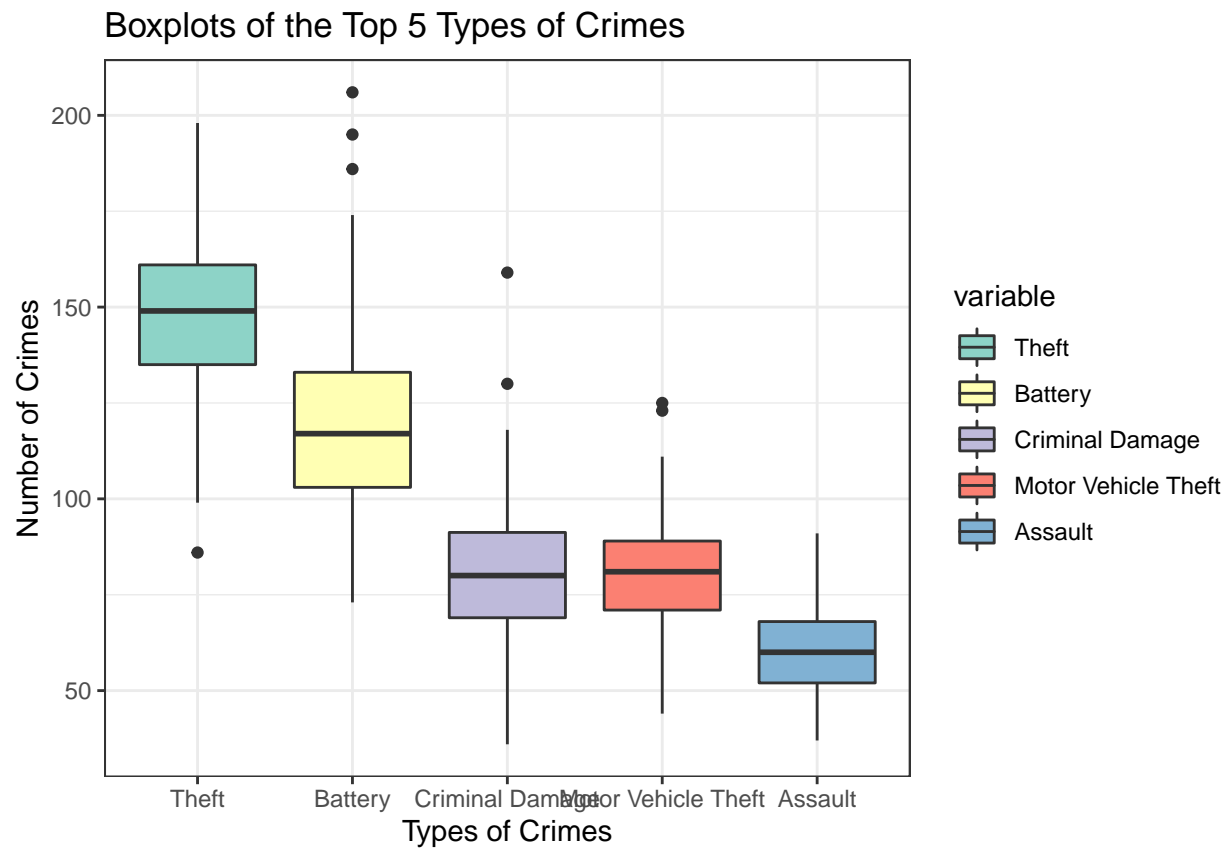
# pdf for the most common types of crimes
ggplot(meltedDens, aes(x = value, fill = variable)) +
  geom_density(alpha = 0.5, adjust = 1) +
  # xlim(c(0, 50)) +
  xlab('Number of Crimes') +
  scale_fill_brewer('Variable', palette = 'Set3',
                    labels = c(ttLabels)) +
  theme_bw() +
  ggtitle('PDFs of the Top 5 Types of Crimes')
```


PDFs of the Top 5 Types of Crimes



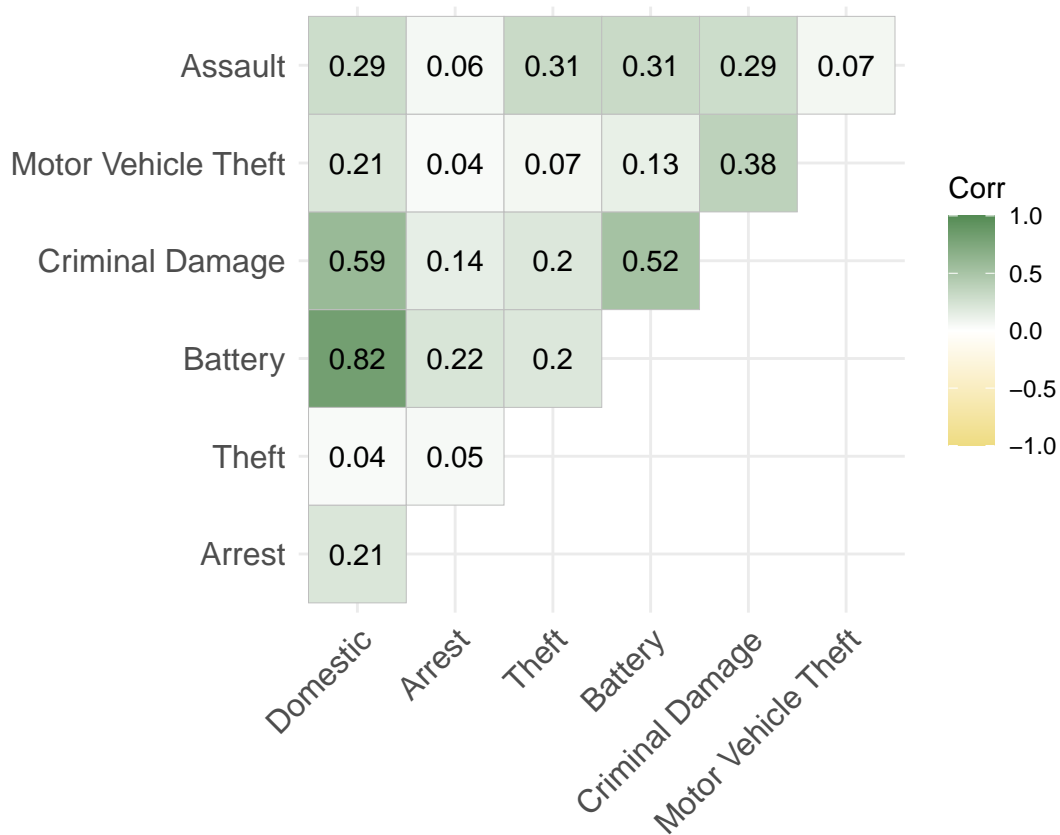
```
# boxplot for the most common types of crimes
aplot <- ggplot(meltedDens,
  aes(x = variable, y = value, fill = variable),
  ) + geom_boxplot()

aplot +
  scale_fill_brewer(palette="Set3") +
  ylab('Number of Crimes') +
  ggtitle('Boxplots of the Top 5 Types of Crimes') +
  scale_x_discrete(name = 'Types of Crimes',
    limits = c(ttLabels) ) +
  theme_bw()
```



Correlation plot for the counts dataset

```
corrCounts <- dfCounts[3:9] %>% cor(method = 'pearson')
corrCounts %>% ggcorrplot::ggcorrplot(lab = TRUE, type = 'upper',
                                       colors = c('lightgoldenrod2', 'white',
                                                  'palegreen4'))
```



Cleaning for data visualization

```
# dataset for each type of primary type of crime, with their frequency counts
tabTypeDF <- data.frame(tabType)
colnames(tabTypeDF) <- c('Types', 'Frequency')

# convert types of crimes from factor to string
tabTypeDF$Types <- sapply(tabTypeDF$Types, toString)

# Move specific crime types to the "other offense" column for better visualization
tabTypeDF[tabTypeDF$Types == 'OTHER OFFENSE', 2] <- sum(sum(tabTypeDF[tabTypeDF$Frequency < 1000,]$Frequency,
                                                             tabTypeDF[tabTypeDF$Types == 'OTHER OFFENSE',]$Frequency))

# sorting the dataset
tabTypeDF <- tabTypeDF[order(tabTypeDF$Frequency),]

# factor strings so that it sorts properly
tabTypeDF$Types <- factor(tabTypeDF$Types, levels = rev(tabTypeDF$Types))
```

Pie and bar plots

```

# Pie plot for most common types of crimes commit
pieP <- ggplot(tabTypeDF[tabTypeDF$Frequency > 1000,],
  aes(x="", y=Frequency, fill=Types)) +
  geom_bar(stat="identity", width=1, color = 'white') +
  coord_polar("y", start=0) +
  theme_void() +
  ggtitle(paste('Types of Crimes')) +
  theme(plot.title = element_text(hjust = 0.5)) +
  geom_text(aes(label = paste0(round(100*Frequency/sum(Frequency)),
    "%")),
    position = position_stack(vjust = 0.5))

# Bar plot for the most common types of crimes commit
barP <- ggplot(tabTypeDF[tabTypeDF$Frequency > 1000,],
  aes(x=Types, y=Frequency, fill = Types)) +
  geom_bar(stat="identity", width=1, color = 'white') +
  theme_bw() +
  theme(axis.text.x = element_blank(),
    axis.ticks.x = element_blank()) +
  ggtitle(paste('Types of Crimes'))

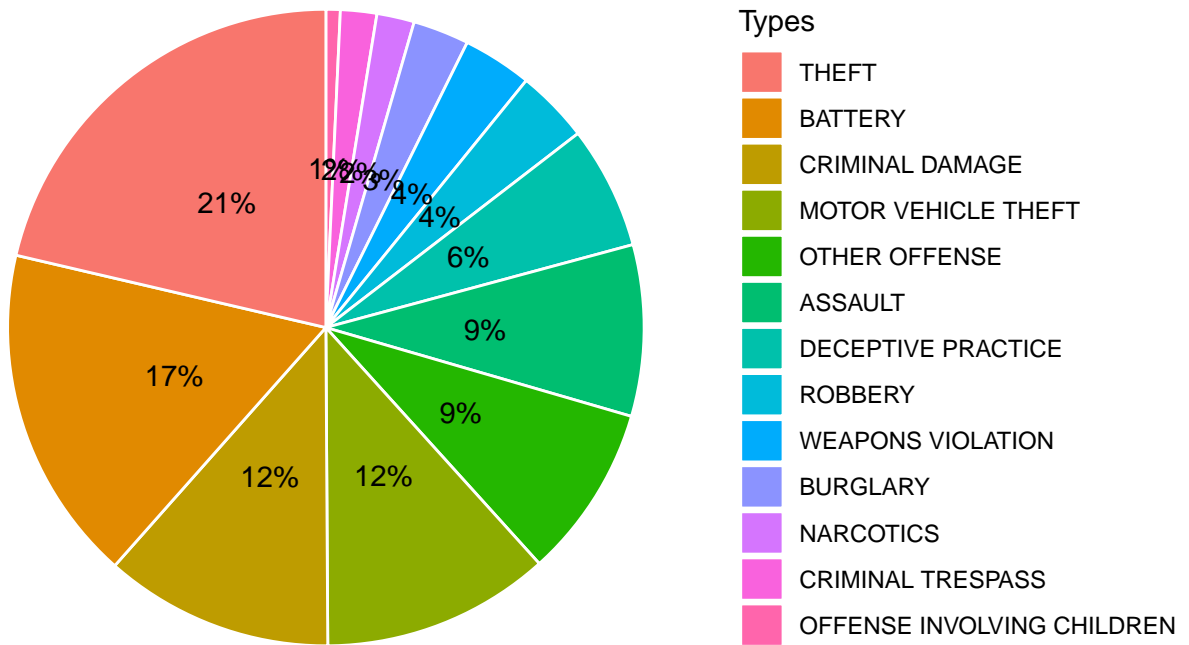
# Calculations for below bar plot
numDays <- length(unique(df$Date))
tabTypeDFAvg <- tabTypeDF
tabTypeDFAvg$Frequency <- tabTypeDFAvg$Frequency / numDays

# Bar plot for the most common types of crimes commit, as mean values
barPAvg <- ggplot(tabTypeDFAvg[tabTypeDFAvg$Frequency > 1,],
  aes(x=Types, y=Frequency, fill = Types)) +
  geom_bar(stat="identity", width=1, color = 'white') +
  theme_bw() +
  theme(axis.text.x = element_blank(),
    axis.ticks.x = element_blank()) +
  ggtitle(paste('Mean Crimes Per Day'))

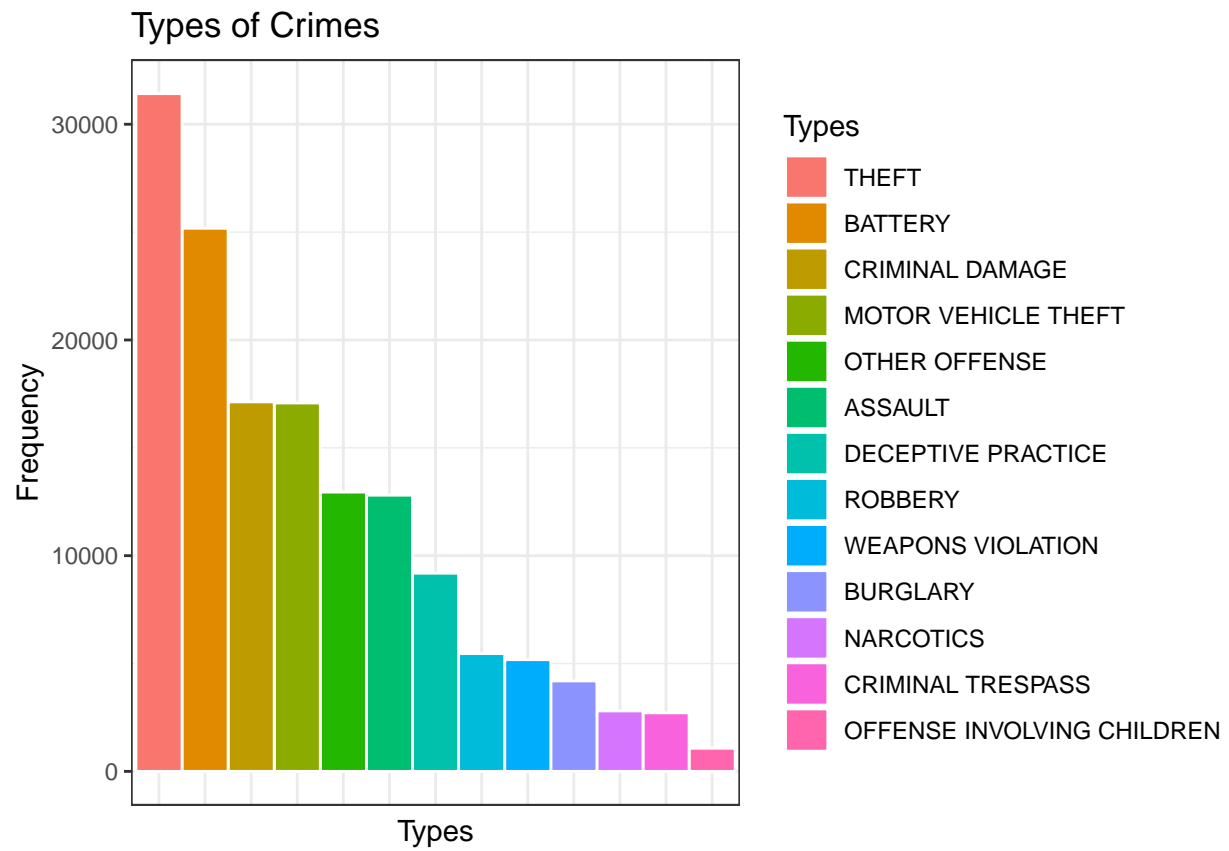
# visualization of the plots
pieP

```

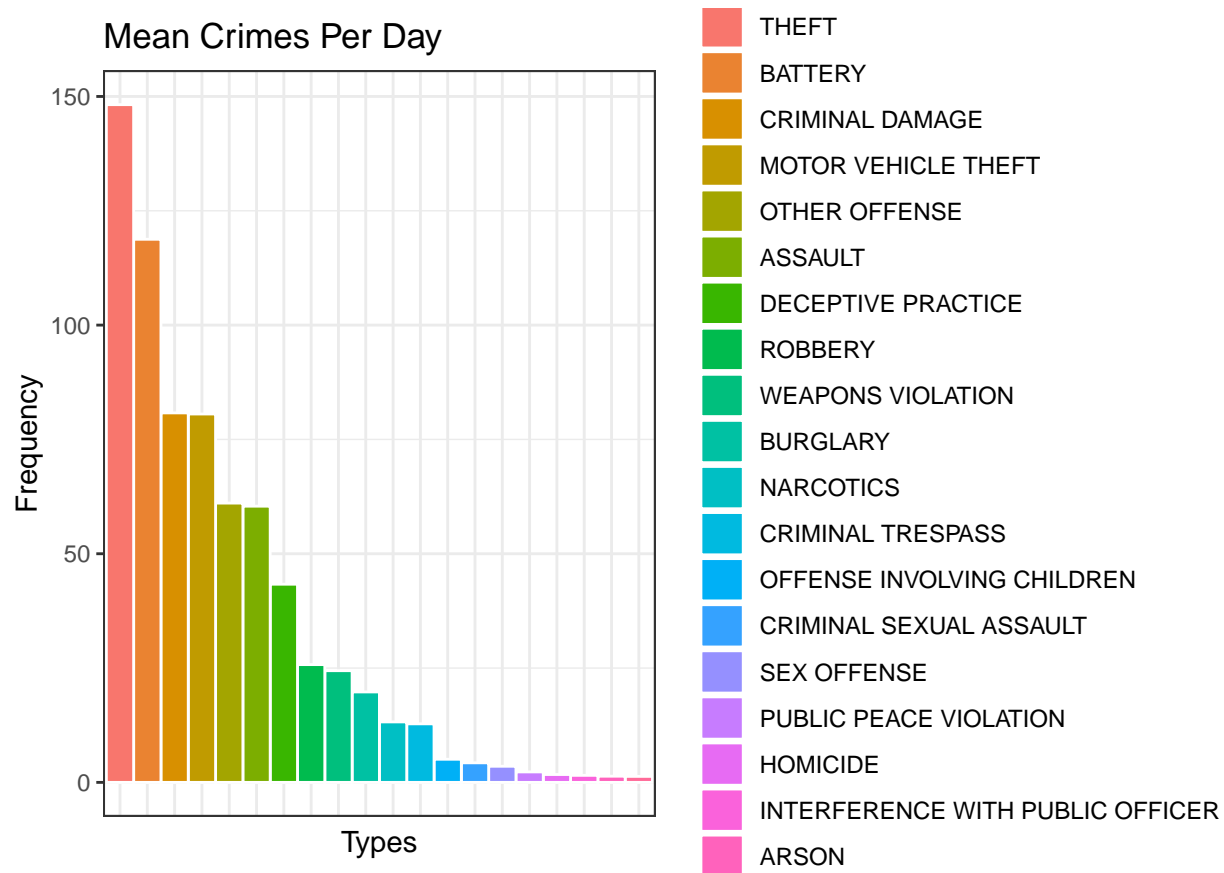
Types of Crimes



barP



barPAvg



Clock of when most crimes happened during the day

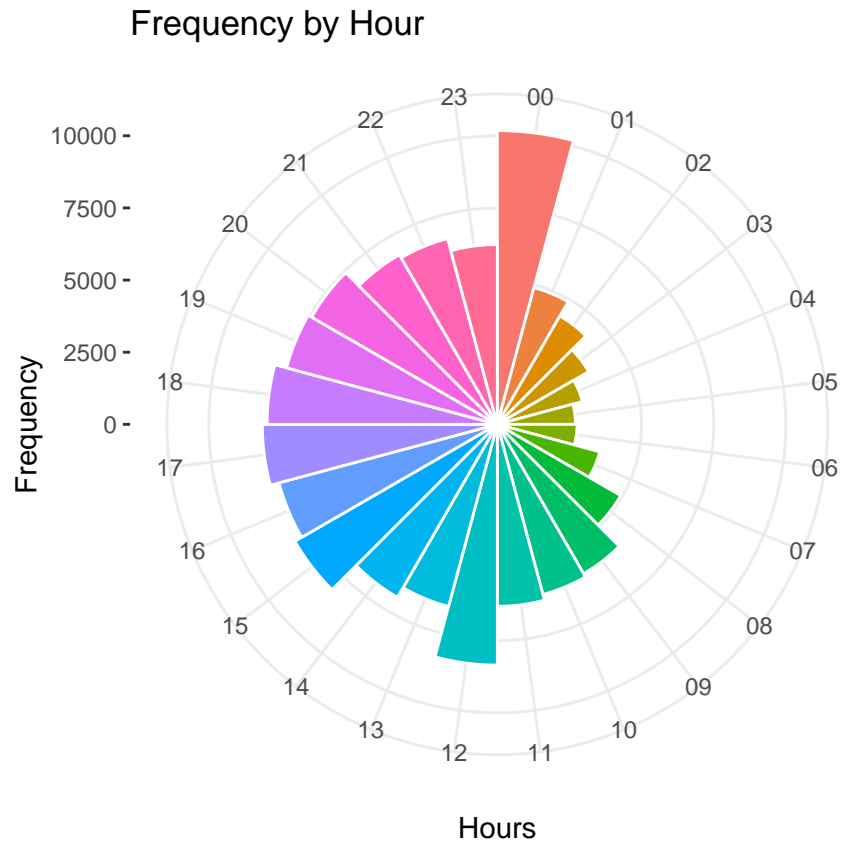
```
# Getting hour integers for when the time of day is AM
hourTimesAM <- df[df$`Time of Day` == 'AM',]$Time %>%
  substr(start = 1, stop = 2)
# turn 12 AM to 00 AM, for easier viz
hourTimes <- ifelse(hourTimesAM == '12', '00', hourTimesAM)

# Getting hour integers for when the time of day is PM
hourTimesPM <- df[df$`Time of Day` == 'PM',]$Time %>%
  substr(start = 1, stop = 2) %>%
  c
# add 12 to all except 12 PM for PM times for easier viz
hourTimes <- c(hourTimes, ifelse(as.integer(hourTimesPM) != 12,
  as.integer(hourTimesPM) + 12,
  hourTimesPM))

# calculating frequencies for specific hour integers
hourTimesFreq <- table(hourTimes)
hourTimesDF <- data.frame(hourTimesFreq)
colnames(hourTimesDF) <- c('Hours', 'Frequency')

# visualizing a clock for when crimes most frequently occur
```

```
barP <- ggplot(hourTimesDF,
               aes(x=Hours, y=Frequency, fill = Hours)) +
  geom_bar(stat="identity", width=1, color = 'white') +
  coord_polar() +
  theme_bw() +
  theme(legend.position = 'none',
        panel.border = element_blank()) +
  ggtitle(paste('Frequency by Hour'))
barP
```



Total crimes by each month

```
# add column just to specify months in the dataframe
dfCounts['Month'] <- dplyr::case_when(grepl('-01-', dfCounts$Date) ~ '1',
                                     grepl('-02-', dfCounts$Date) ~ '2',
                                     grepl('-03-', dfCounts$Date) ~ '3',
                                     grepl('-04-', dfCounts$Date) ~ '4',
                                     grepl('-05-', dfCounts$Date) ~ '5',
                                     grepl('-06-', dfCounts$Date) ~ '6',
                                     grepl('-07-', dfCounts$Date) ~ '7')

# create dataframe specifically for months, for data viz
dfCountsMonths <- data.frame('Months' = unique(dfCounts$Month),
```



```

row.names = c(unique(dfCounts$Month))

# sum total crimes for each month
for (i in unique(dfCounts$Month)) {
  dfCountsMonths[i, 'Total Crimes'] <- sum(dfCounts[dfCounts$Month == i,]$`Number Of Crimes`)
}

# bar plot for unique months
dfCountsMonths %>% ggplot(aes(x=Months, y=`Total Crimes`, fill = Months)) +
  scale_fill_brewer(palette="Set3") +
  geom_bar(stat="identity", width=1, color = 'white') +
  theme_bw() +
  ggtitle(paste('Total Crimes by Months'))

```

