Chicago Crime: Data Analysis and Visualisations using R

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November 3, 2018

1 Introduction

This report is the first assessment for the **MATH5741M Statistical Theory and Methods** module. Its objective is to summarise statistically a dataset sample of crimes in the city of Chicago and answer the following research questions:

- How has crime evolved over time in the city of Chicago?

- What time of day do most types of crime occur?

- In which locations are specific types of crime more likely to happen?

- Which districts of the city are potentially more dangerous per type of crime?

2 Data and methods

The dataset analysed is a sample of the original data of crimes extracted from the Chicago Police Department which content the crimes that occurred in the city of Chicago from 2001 to present.

For the analysis, first, we prepare the data creating, transforming and simplifying variables, as well as cleaning the dataset keeping the variables we are interested in. Secondly, to answer our research questions we perform the an analysis based on line graphs and heat-maps. Finally, we summarise the findings.

The report has been done with `Rmarkdown` but unfortunately does not include all the `R` code cells written for its performance[1]. However, it is available for consultation in this link https://github.com/eugenividal/Chicago-Crime-Data-Analysis.

3 Results

3.1 Data preparation

First, we load the libraries we will need for the project and get the data into the `R` environment.

```r
# load libraries
library(ggplot2)
library(ggmap)
library(lubridate)
library(scales)
library(zoo)
library(dplyr)
library(knitr)
```

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[1] In this report it is not included the code used to group categories in variables `Primary.Type` and `Location.Description`, neither the code used to generate the visualisations

```
# Read csv in R
dd=read.csv("http://www1.maths.leeds.ac.uk/~charles/math5741/crime.csv",header=T)
```

Second, we create the new variables (`Count`, `Month_year`, `Hour`) based on the existing ones, and give them the right format for later exploration.

```
# Create a variable count with value 1
dd$Count <- 1
# Convert Date from factor to date
dd$Date <- mdy_hms(dd$Date)
# Extract hour from Date
dd$Hour <- substring(dd$Date, 12,13)
# Drop time from Date
dd$Date <- as.Date(dd$Date, format="%m/%d/%Y")
# Drop days from Date
dd$Month_year <- as.Date(as.yearmon(dd$Date, "%Y-%m"))
```

Third, we group the categories of the variables `Primary.Type` and `Location.Description` to simplify the analysis and call them `Type_grouped` and `Location_grouped` respectively.

Next, we drop all variables we do not need to answer our research questions.

```
# Drop all variables we are not interested in
dd <- dd[, -c(1:11, 13:18)]
```

Then, we clean the dataset of missing values.

```
# Remove NAs
dd <- dd[complete.cases(dd),]
```

Finally, we show the the dataset ready for exploration.

```
# Show first 5 records
head(dd)
```

```
##   District Count Hour Month_year Type_grouped Location_grouped
## 1       19     1   00 2013-07-01        Batery            Street
## 2       19     1   01 2013-07-01        Others            Street
## 3        2     1   21 2013-07-01       Assault         Apartment
## 4        9     1   02 2013-07-01     Narcotics            Street
## 5        3     1   17 2013-07-01         Theft            Street
## 6        9     1   01 2013-07-01        Batery         Apartment
```

3.2 Data exploration

3.2.1 How has crime evolved over time in the city of Chicago?

To answer the first question we plot in Figure 1 the number of crimes per month from 2001 to the present. The graph shows that crime in the city of Chicago has been decreasing consistently over the whole period. The wave-shape of the graph also shows that there is a clear periodic pattern per months of the year.

In Figure 2 we plot the same graph per type of crime. Except the deceptive practice crime, which at present keeps in similar values as in 2001, the rest of type have been falling to a greater or lesser extent.
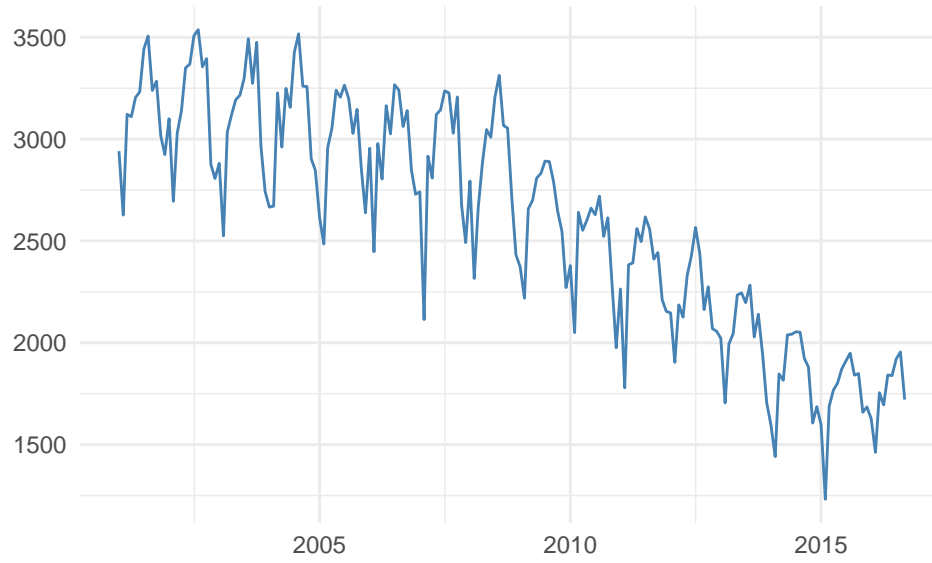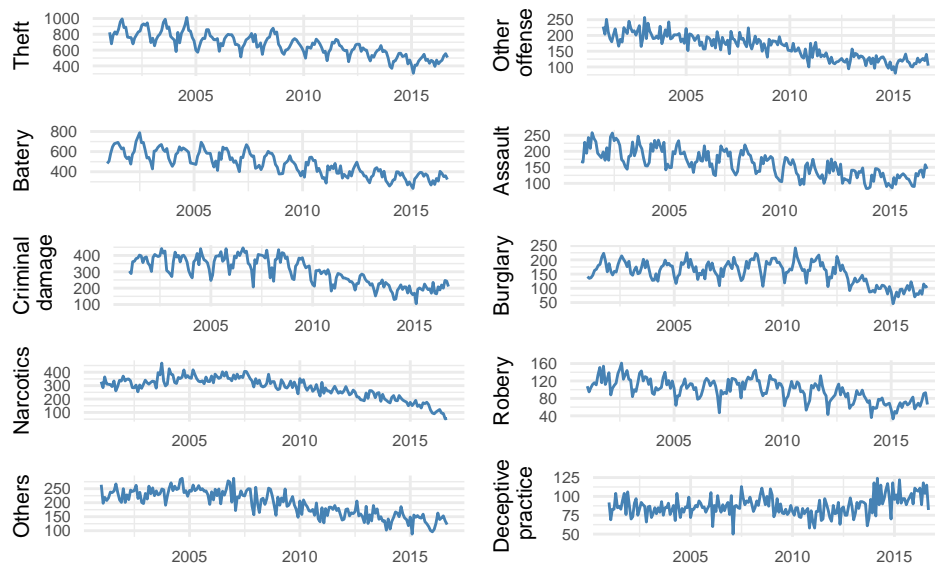
Figure 1: Crimes evolution

Figure 2: Evolution per type of crime

3.2.1.1 Crime per hour

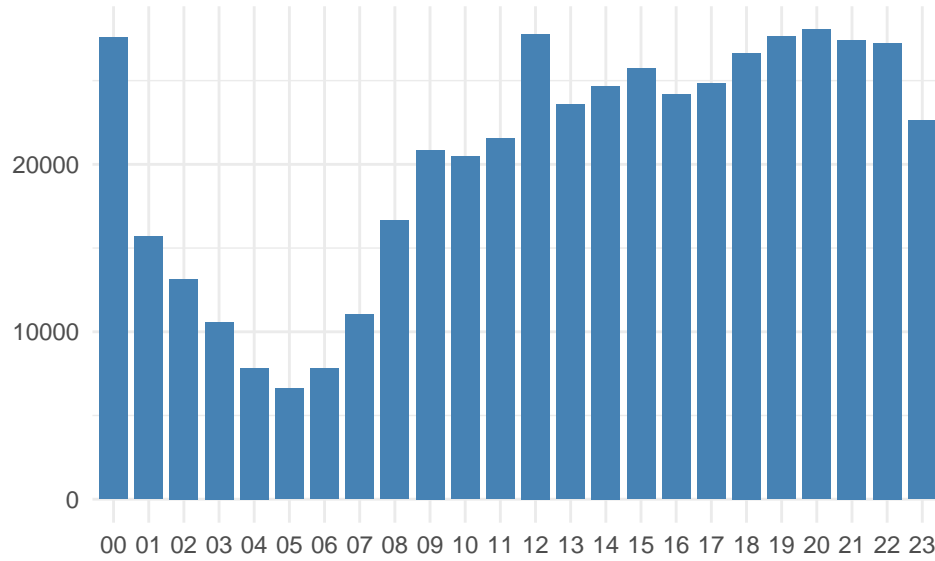The crimes are concentrated in hours

Figure 3: Crimes per hour

3.2.1.2 Type of crimes

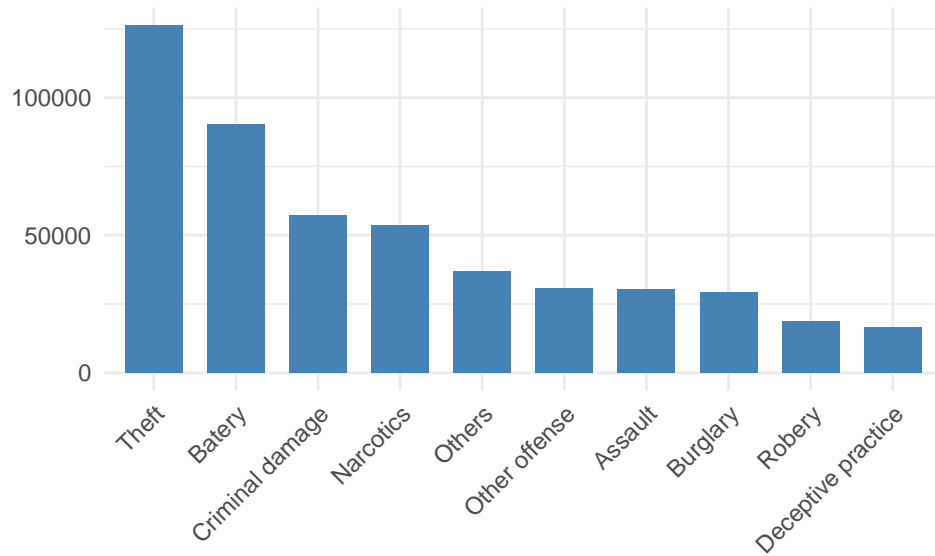Per type of crime Theft is in difference the biggest number. Change the scientifyc number.

Figure 4: Crimes per type

3.2.1.3 Location of crimes
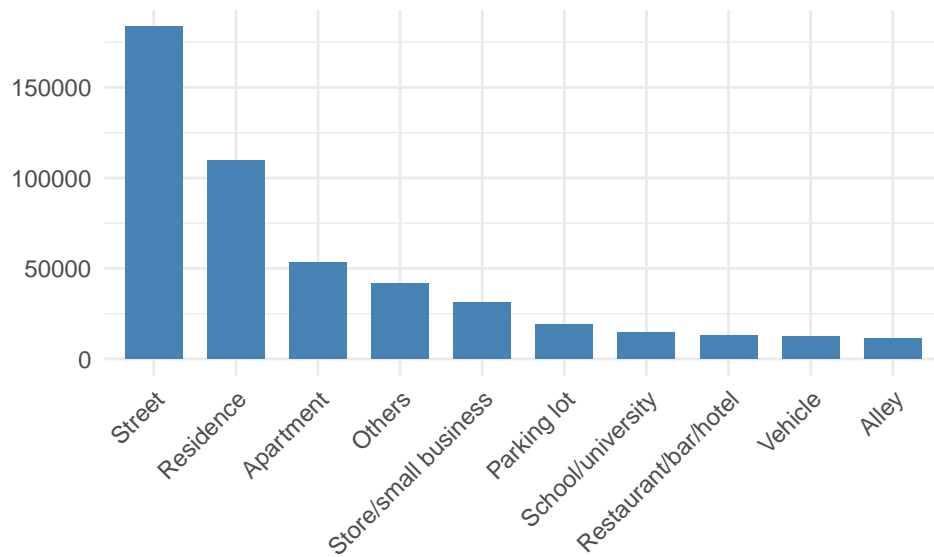
These crimes are concentrated in Streets, give percentage.

Figure 5: Crimes per location

3.2.1.4 Crime per districts

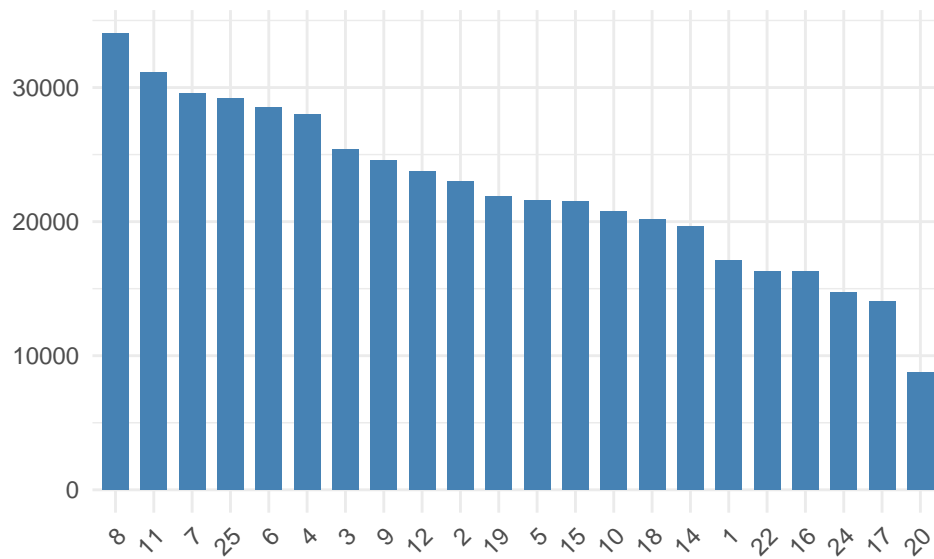Per districts the most dangerous are 8.

Figure 6: Crimes per district

3.2.2 What time of day do most types of crime occur?

Figure 3 a head map between type of crime and hour. The peak hour of crimes morning, other crimes peak during the and the final group of crimes peak at night.
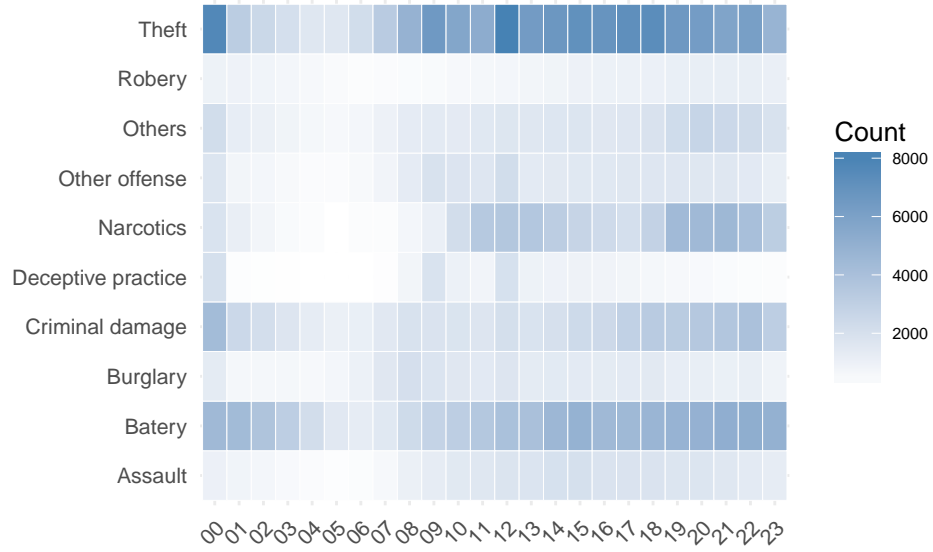
Figure 7: Type of crime vs hour

3.2.3 In which locations are specific types of crime more likely to happen?

Figure 4 shows that some locations are more likely per type of crime. For theft for instance it is better this or that.
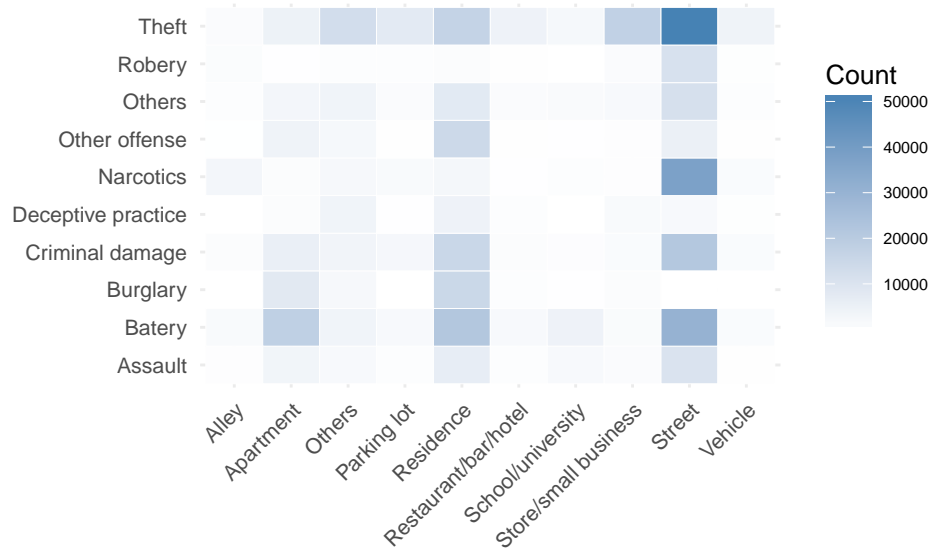

Figure 8: Type of crime vs location

3.2.4 Which districts of the city are potentially more dangerous per type of crime?

Finally, Figure 5 depicts the districts which are more potentially dangerous per tyep of crime. District 11 for instance seems clearly problematic for Narcotics.
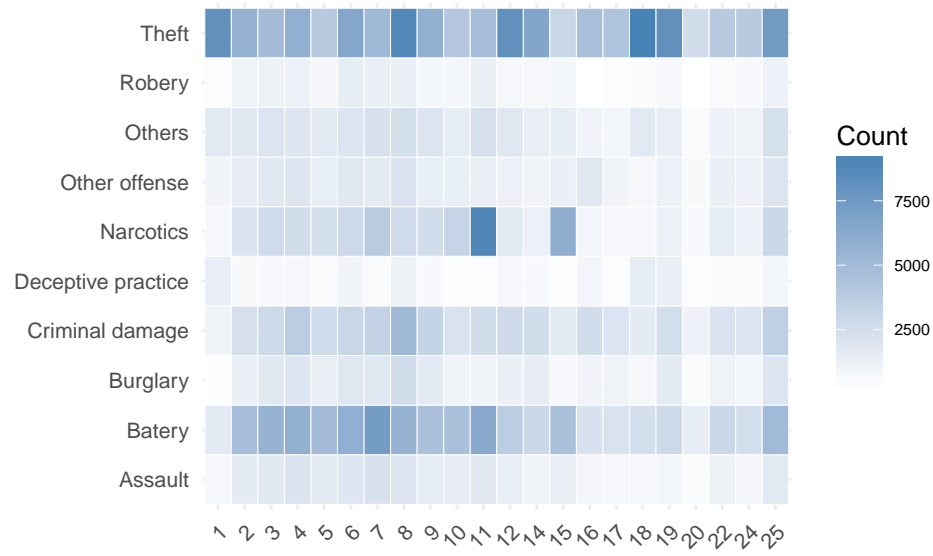
Figure 9: Type of crime vs district

4 Conclusions