# Prediction of crime affected districts of Chicago

#### Introduction

- Violent crimes, thefts, and robberies occur in majority of places. On daily basis, local businesses, houses etc encounter such crimes.
- Our job as data scientist is to analyze the pattern of crimes. study the distribution of types of crime that occur in various part of a particular place. It is important to know where the crimes are happening, usually areas with police stations farther in distance can result into increase crime count

### Data Understanding

Here in this project, I will utilize the crime dataset & police stations provided by <a href="Chicago Data Portal">Chicago Data Portal</a>.

#### Crime data set

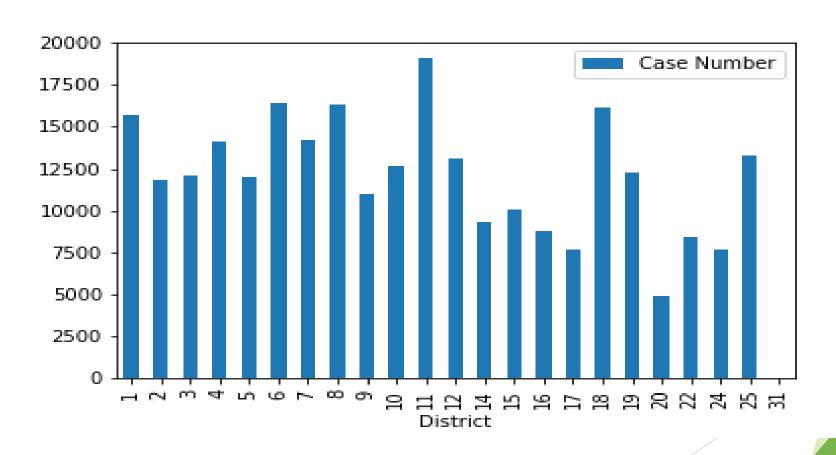
|    | Case Number | Date                   | Primary Type      | Location Description                           | District | Community Area | Latitude  | Longitude  |
|----|-------------|------------------------|-------------------|--|----------|----------------|-----------|------------|
| 1  | JC104662    | 12/31/2018 11:59:00 PM | CRIMINAL DAMAGE   | STREET   | 22       | 74.0           | 41.689079 | -87.696064 |
| 2  | JC100043    | 12/31/2018 11:57:00 PM | CRIMINAL DAMAGE   | APARTMENT                                      | 6        | 71.0           | 41.740521 | -87.647391 |
| 3  | JC100006    | 12/31/2018 11:56:00 PM | BATTERY           | OTHER  | 12       | 31.0           | 41.857068 | -87.657625 |
| 4  | JC100031    | 12/31/2018 11:55:00 PM | BATTERY           | APARTMENT                                      | 6        | 71.0           | 41.751914 | -87.647717 |
| 5  | JC100026    | 12/31/2018 11:49:00 PM | BATTERY           | STREET   | 15       | 25.0           | 41.875684 | -87.760479 |
| 6  | JC100011    | 12/31/2018 11:48:00 PM | BATTERY           | APARTMENT                                      | 6        | 71.0           | 41.750154 | -87.661009 |
| 7  | JC100089    | 12/31/2018 11:47:00 PM | BATTERY           | VEHICLE - OTHER RIDE SHARE SERVICE (E.G., UBER | 19       | 5.0            | 41.939625 | -87.673996 |
| 8  | JC101094    | 12/31/2018 11:45:00 PM | THEFT             | BAR OR TAVERN                                  | 19       | 6.0            | 41.940519 | -87.654124 |
| 9  | JC101652    | 12/31/2018 11:45:00 PM | CRIMINAL DAMAGE   | APARTMENT                                      | 14       | 23.0           | 41.905562 | -87.707589 |
| 10 | JB574407    | 12/31/2018 11:44:00 PM | CRIMINAL TRESPASS | MOVIE HOUSE/THEATER                            | 19       | 3.0            | 41.968463 | -87.659670 |

The dataset consist of **22 fields** and over **226K rows**. The fields include attributes related to crime such as *Case number*, *block*, *IUCR codes*, *Description*, *Type of Crime*, *location*, *Lattitude*, *Longitude*, *district*.

#### Police data set

|   | DISTRICT | DISTRICT NAME  | STATE | ZIP   | LATITUDE  | LONGITUDE  |
|---|----------|----------------|-------|-------|-----------|------------|
| 0 | 1        | Central        | IL    | 60616 | 41.858373 | -87.627356 |
| 1 | 2        | Wentworth      | IL    | 60609 | 41.801811 | -87.630560 |
| 2 | 3        | Grand Crossing | IL    | 60637 | 41.766431 | -87.605748 |
| 3 | 4        | South Chicago  | IL    | 60617 | 41.707933 | -87.568349 |
| 4 | 5        | Calumet        | IL    | 60628 | 41.692723 | -87.604506 |

relation between each district and the number of recorded cases in 2018 for each of them.



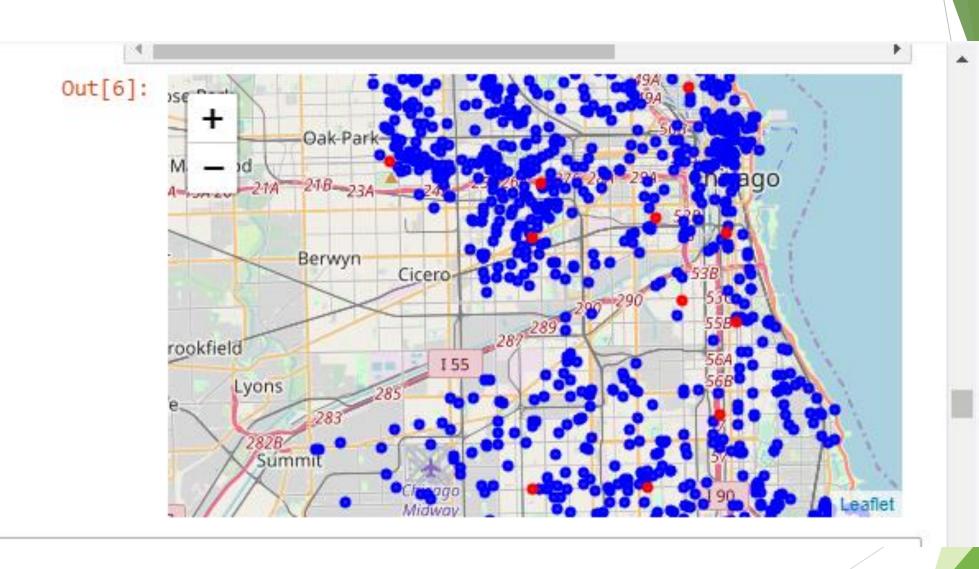
#### Analysing Top 10 crimes in the district

|   | District | 1st Most<br>Common<br>Crime | 2nd Most<br>Common<br>Crime | 3rd Most<br>Common<br>Crime | 4th Most<br>Common<br>Crime | 5th Most<br>Common<br>Crime | 6th Most<br>Common<br>Crime | 7th Most<br>Common<br>Crime | 8th Most<br>Common<br>Crime | 9th Most<br>Common<br>Crime | 10th Most<br>Common<br>Crime |   |
|---|----------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|------------------------------|---|
| 0 | 1        | THEFT                       | DECEPTIVE<br>PRACTICE       | BATTERY                     | CRIMINAL<br>DAMAGE          | ASSAULT                     | OTHER<br>OFFENSE            | ROBBERY                     | CRIMINAL<br>TRESPASS        | MOTOR<br>VEHICLE<br>THEFT   | BURGLARY                     | ٨ |
| 1 | 2        | THEFT                       | BATTERY                     | CRIMINAL<br>DAMAGE          | ASSAULT                     | OTHER<br>OFFENSE            | DECEPTIVE<br>PRACTICE       | ROBBERY                     | MOTOR<br>VEHICLE<br>THEFT   | BURGLARY                    | CRIMINAL<br>TRESPASS         | ٨ |
| 2 | 3        | BATTERY                     | THEFT                       | CRIMINAL<br>DAMAGE          | ASSAULT                     | OTHER<br>OFFENSE            | BURGLARY                    | DECEPTIVE<br>PRACTICE       | ROBBERY                     | NARCOTICS                   | MOTOR<br>VEHICLE<br>THEFT    |   |
| 3 | 4        | BATTERY                     | THEFT                       | CRIMINAL<br>DAMAGE          | ASSAULT                     | OTHER<br>OFFENSE            | BURGLARY                    | DECEPTIVE<br>PRACTICE       | MOTOR<br>VEHICLE<br>THEFT   | NARCOTICS                   | ROBBERY                      |   |
| 4 | 5        | BATTERY                     | THEFT                       | CRIMINAL<br>DAMAGE          | OTHER<br>OFFENSE            | ASSAULT                     | BURGLARY                    | DECEPTIVE<br>PRACTICE       | NARCOTICS                   | WEAPONS<br>VIOLATION        | ROBBERY                      |   |
| 5 | 6        | BATTERY                     | THEFT                       | CRIMINAL<br>DAMAGE          | ASSAULT                     | OTHER<br>OFFENSE            | NARCOTICS                   | DECEPTIVE<br>PRACTICE       | BURGLARY                    | ROBBERY                     | WEAPONS<br>VIOLATION         | _ |
| 6 | 7        | BATTERY                     | THEFT                       | CRIMINAL<br>DAMAGE          | ASSAULT                     | OTHER<br>OFFENSE            | NARCOTICS                   | WEAPONS<br>VIOLATION        | BURGLARY                    | ROBBERY                     | DECEPTIVE<br>PRACTICE        |   |

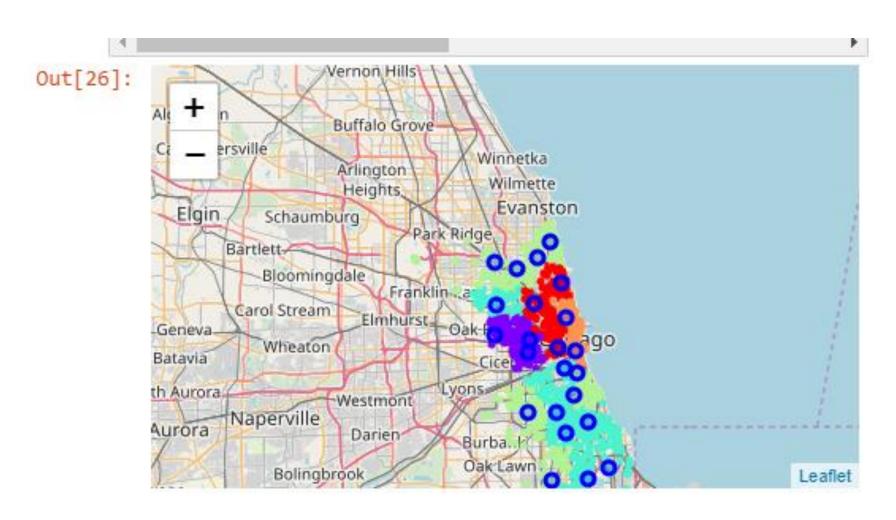
## Scatter plot showing densely affected areas by crime

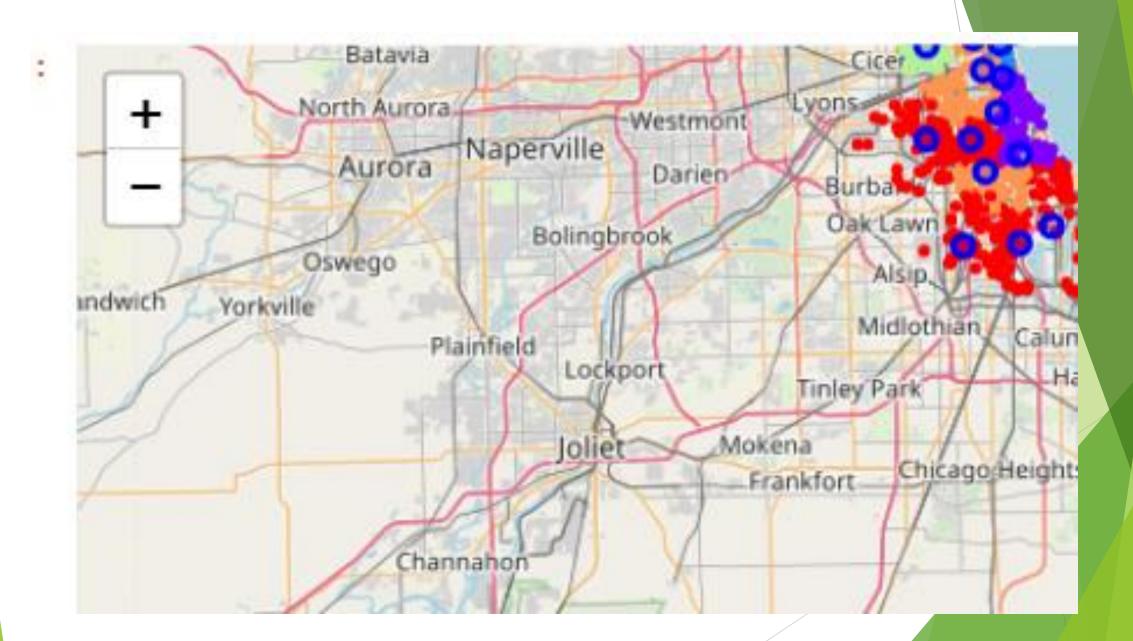






# Provide relationship between the Location Type of a crime and the districts





#### Conclusion

- In this project we analyzed various distribution of crimes in the chicago city area. The city being divided into 23 districts became the key driver of the distribution.
- As we can see there is a strong co-relation of the clusters along with its primary type as well as location of the crime that is taking place