

#### Problem Statement

- Chicago has one of the highest crime rates in United States.
- ♦ The city's crime rate is higher than the US average .
- Our project aims to explore crime patterns in Chicago and build model to predict

   1)whether an arrest is possible
   2)the type of crime

#### Data Source

The dataset that we used to solve this problem was:

Chicago crime data of 2019

https://data.cityofchicago.org/Public-Safety/Crimes-2019/w98m-zvie

Chicago crime data of 2018

https://data.cityofchicago.org/Public-Safety/Crimes-2018/3i3m-jwuy

♦ In both our crime datasets from 2018 and 2019 we had approximately 500,000 observed crimes and in these data sets we have the following variables: ID, Case Number, Date, Block, IUCR, Primary Type, Description, Location Description, Arrest, Domestic, Beat, District, Ward, Community Area, FBI Code, X Coordinate, Y Coordinate, Year, Updated on, Latitude, Longitude, Location.

#### Data Sources

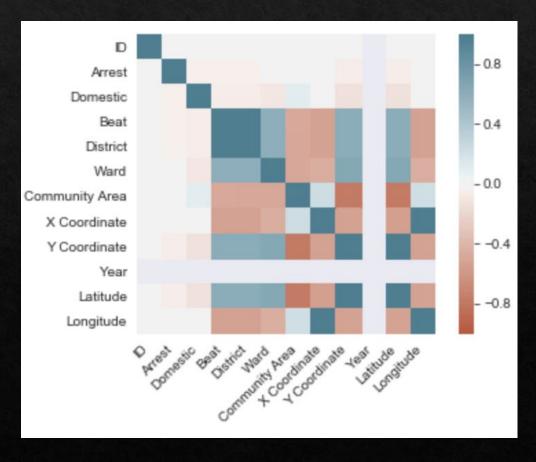
- Chicago police stations data location- <u>https://www.chicago.gov/city/en/depts/cpd/dataset/police\_stations.</u> <u>html</u>
- The Chicago police station data location set has 23 police stations with the following variables:
  - District, District Name, Address, X coordinate, Y coordinate, Latitude, Longitude, Location.
- We used the distance between the police station and crime location as one of the attribute for our prediction

## Data preprocessing

- Columns with missing values: Latitude, Longitude, X Coordinate, Y Coordinate and Location Descriptions.
- We dropped the rows containing missing values
- Other irrelevant attributes like case number, FBI code were dropped.
- Mapped the string labels to numerical values
   Arrests (0-False and 1- True)
   Domestic (0-False and 1- True)
- Dummy variables were created for categorical variables
- Type of crime were grouped in 5 categories

## **Exploratory Data Analysis**

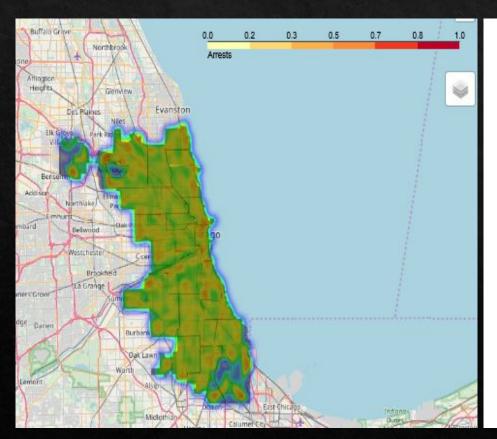
Correlation Matrix: We used the below matrix to find linearly correlated attributes

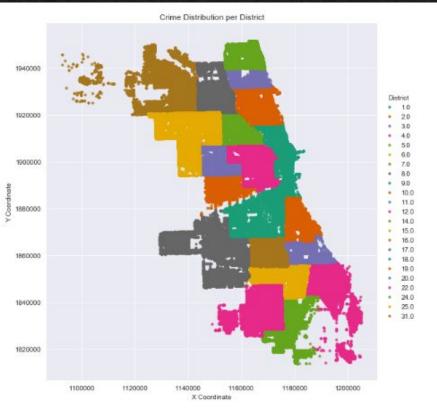


## Distribution Map

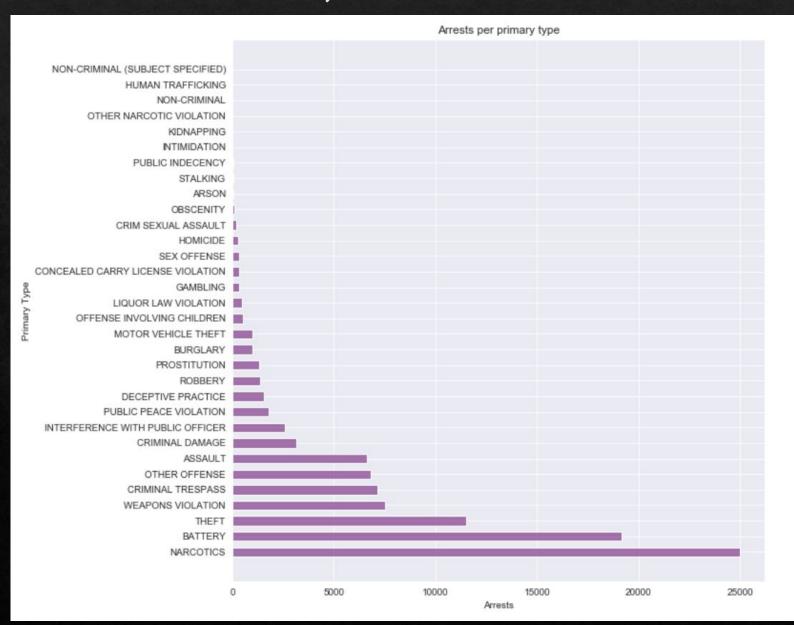
Heat Map of arrests across districts of Chicago

Scatter plot of crimes across districts

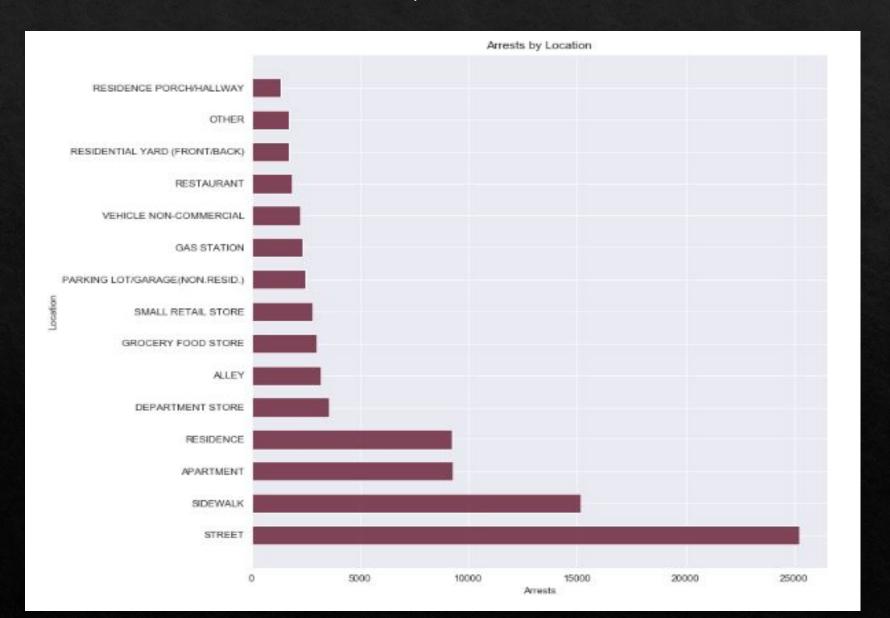




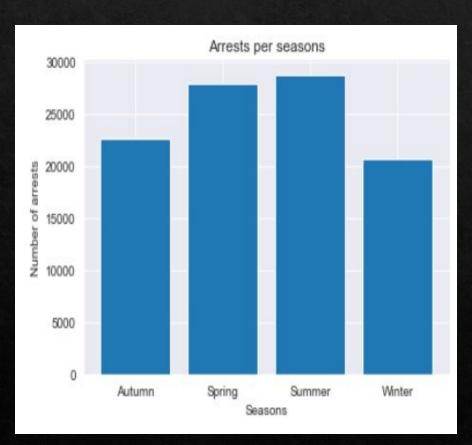
## Primary Crime Vs Arrests

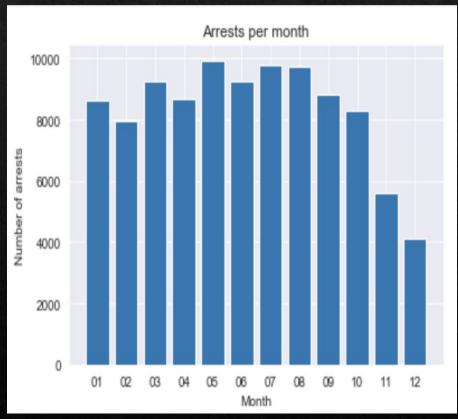


## Arrests by location

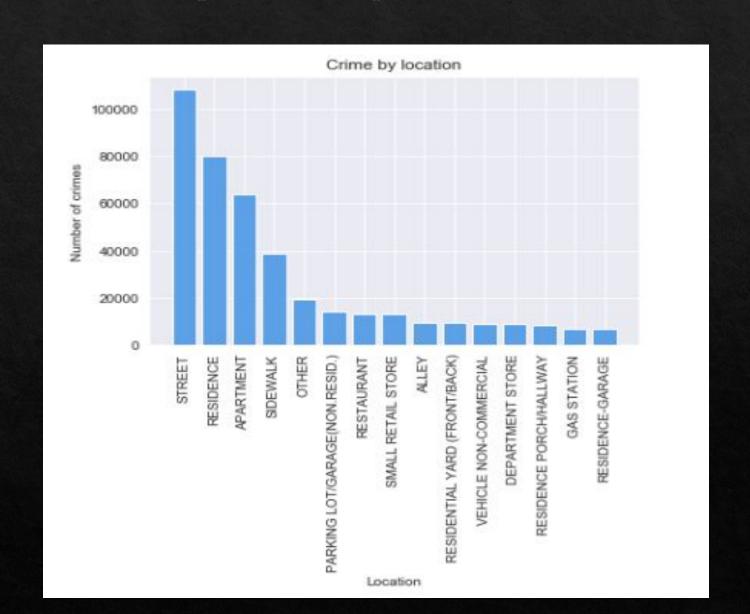


#### Arrests Vs Seasons and Months





## Top 15 Crimes per location



# Data science solution

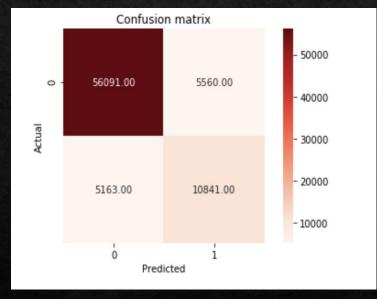
- A classification model to predict whether an arrest will be made.
- And a classification and clustering model to predict the type of crime

# How was the data prepared and techniques used to build models?

- ♦ 20% of data was used as Test set
- ♦ 20% as validation set
- ♦ 60% as train set
- Based on the plots and correlation matrix, the parameters and attributes were chosen for building the model.
- The classification techniques used were:
  - Logistic Regression, Random Forests, Naïve Bayes,
- The clustering techniques we used were:
- ♦ K means , K-means++

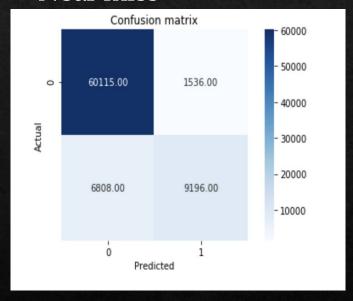
## Classification

 Logistic Regression After SMOTE



Accuracy - 86.19% F1 Score - 86.25%

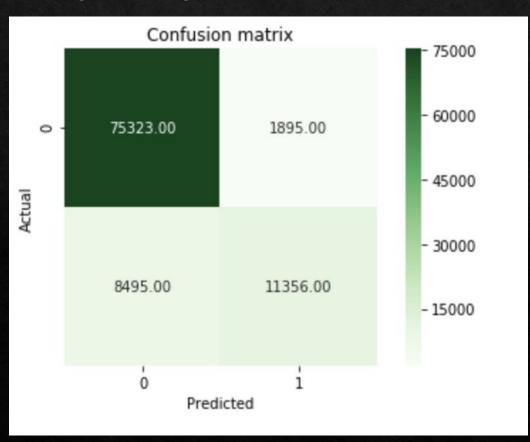
 Logistic Regression after Near miss



Accuracy - 89.25% F1 Score- 88.41%

#### Result of Classification

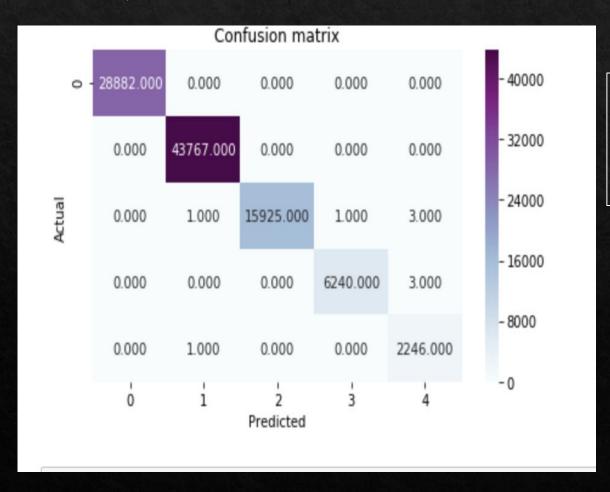
#### Logistic Regression Final Model



Accuracy- 89.29%
F1 Score- 88.44%

# Result of Classification Crime Groups

Naive Bayes

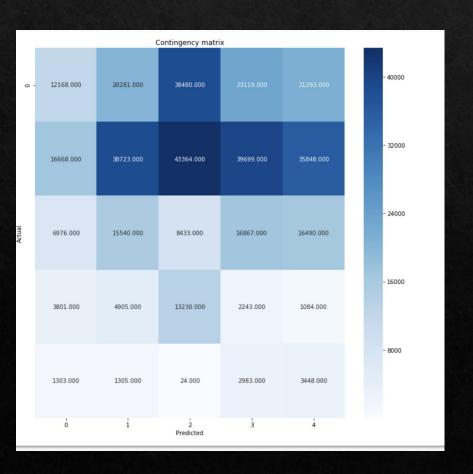


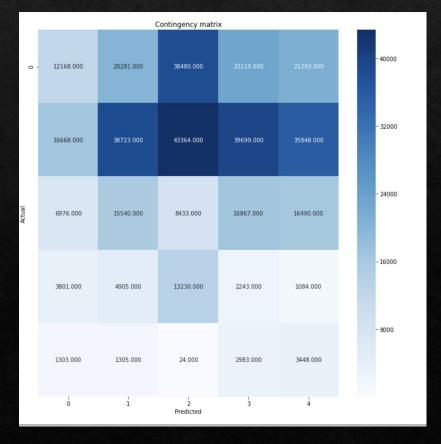
Accuracy- 99.9% Fl Score- 99.99%

# Result of Clustering Crime Groups

K-Means++ [0.0109, 0.2247]

K-means [0.0112, 0.2253]





# Conclusions

- Logistic regression did a good job predicting arrests even though the class was imbalanced.
- 2. Naives Bayes had a high accuracy in predicting crime groups.
- 3. Clustering didn't perform good because the data was not well separable and high correlation between crime types.

Thankyou!!