

Chicago's crime data in 2021

Agenda

Introduction



Description



Results



Conclusion

Introduction

In practically every nation, crime is rising quickly. Analyzing many aspects of crime and identifying criminal trends are highly important. Security agencies throughout the world are working hard to lower these crimes, but the volume of crime data is growing quickly, making it challenging to handle such a large amount of data and to maintain records of crimes that occur over a vast geographic area and over a variety of time periods. Therefore, having a criminal information system that can analyze a lot of data quickly is essential. Also, one of the most efficient techniques to explore, analyze, find patterns, and forecast future crimes in a vast quantity of data is data analysis by using clustering, classification, and association mining.

Description

Dataset for year 2021

.csv format

48.1 MB → size

Collected from : Kaggle

Libraries used

- Data Wrangling → Panadas
- NumPy (for math operations)
- Datetime (python datetime module)
- Data Visualization
→
Matplotlib & seaborn

2.1 Data Gathering

```
# put dataset --> dataframe
df = pd.read_csv('Chicago_Crimes_2021.csv')
df.head(5)
```

	ID	Case Number	Date	Block	IUCR	Primary Type	Description	Location Description	Arrest	Domestic	...	Ward	Community Area	FBI Code	X Coordinate	Y Coordinate	Year
0	12445124	JE328077	08/05/2021 10:00:00 PM	033XX S LEAVITT ST	0610	BURGLARY	FORCIBLE ENTRY	RESIDENCE - GARAGE	0	0	...	25.0	59.0	05	1162228.0	1882544.0	2021
1	12541099	JE445417	11/14/2021 10:58:00 PM	003XX N LAVERGNE AVE	143A	WEAPONS VIOLATION	UNLAWFUL POSSESSION - HANDGUN	STREET	1	0	...	28.0	25.0	15	1142984.0	1901704.0	2021
2	12344992	JE204888	04/16/2021 12:00:00 AM	011XX W 50TH ST	0820	THEFT	\$500 AND UNDER	APARTMENT	0	0	...	20.0	61.0	06	1169689.0	1871646.0	2021
3	12345838	JE205622	03/03/2021 06:20:00 PM	012XX N WELLS ST	0820	THEFT	\$500 AND UNDER	STREET	0	0	...	2.0	8.0	06	1174488.0	1908479.0	2021
4	12444222	JE326942	08/05/2021 11:00:00 AM	044XX N BROADWAY	0486	BATTERY	DOMESTIC BATTERY SIMPLE	APARTMENT	0	1	...	46.0	3.0	08B	1168547.0	1929701.0	2021

5 rows × 22 columns



Summary

21.csv



#Summary

df.info()



```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 207914 entries, 0 to 207913
Data columns (total 22 columns):
#   Column                Non-Null Count  Dtype
---  -
0   ID                    207914 non-null int64
1   Case Number          207914 non-null object
2   Date                 207914 non-null object
3   Block               207914 non-null object
4   IUCR                207914 non-null object
5   Primary Type        207914 non-null object
6   Description         207914 non-null object
7   Location Description 207071 non-null object
8   Arrest              207914 non-null int64
9   Domestic            207914 non-null int64
10  Beat                207914 non-null int64
11  District            207914 non-null float64
12  Ward                207903 non-null float64
13  Community Area      207914 non-null float64
14  FBI Code            207914 non-null object
15  X Coordinate        202902 non-null float64
16  Y Coordinate        202902 non-null float64
17  Year                207914 non-null int64
18  Updated On         207914 non-null object
19  Latitude            202902 non-null float64
20  Longitude           202902 non-null float64
21  Location            202902 non-null object
dtypes: float64(7), int64(5), object(10)
memory usage: 34.9+ MB
```

84.74 GB available

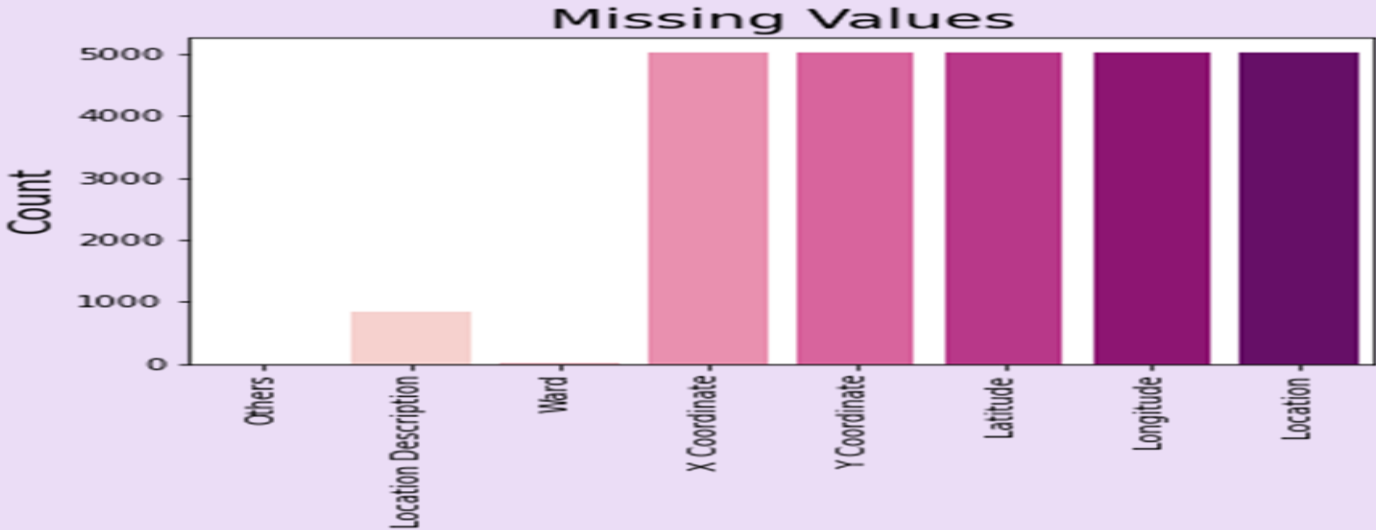
Missing Values

1.csv

✓ [9] The number of missing values --> per feature is :

	Others	Location	Description	Ward	X Coordinate	Y Coordinate	Latitude	Longitude	Location
Missing Values	0		843	11	5012	5012	5012	5012	5012

✓ # Plotting the missing values in the dataset



Missing Values Cont.

The simplest cleaning technique here
would be to drop all the rows with at least
one missing value --> as every feature in
the dataset is important

```
rt Runtime Tools Help All changes saved

+ Code + Text

df.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 202264 entries, 0 to 207903
Data columns (total 22 columns):
#   Column                Non-Null Count  Dtype
---  -
0   ID                    202264 non-null int64
1   Case Number          202264 non-null object
2   Date                 202264 non-null object
3   Block                202264 non-null object
4   IUCR                 202264 non-null object
5   Primary Type         202264 non-null object
6   Description           202264 non-null object
7   Location Description  202264 non-null object
8   Arrest               202264 non-null int64
9   Domestic             202264 non-null int64
10  Beat                 202264 non-null int64
11  District             202264 non-null float64
12  Ward                 202264 non-null float64
13  Community Area       202264 non-null float64
14  FBI Code             202264 non-null object
15  X Coordinate         202264 non-null float64
16  Y Coordinate         202264 non-null float64
17  Year                 202264 non-null int64
18  Updated On          202264 non-null object
19  Latitude             202264 non-null float64
20  Longitude            202264 non-null float64
21  Location             202264 non-null object
dtypes: float64(7), int64(5), object(10)
memory usage: 35.5+ MB

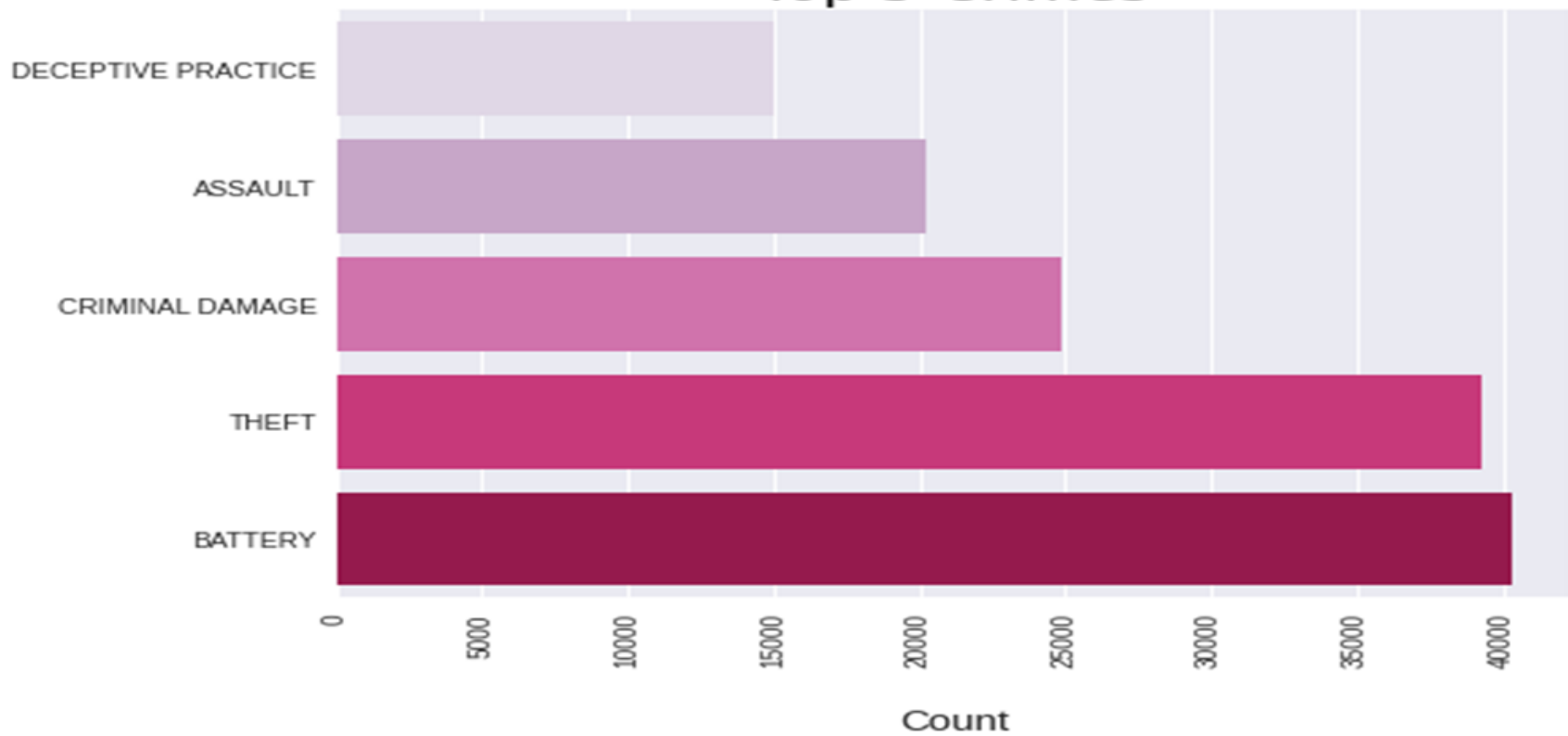
21.csv

✓ [12] # How much of the data has been deleted?
```

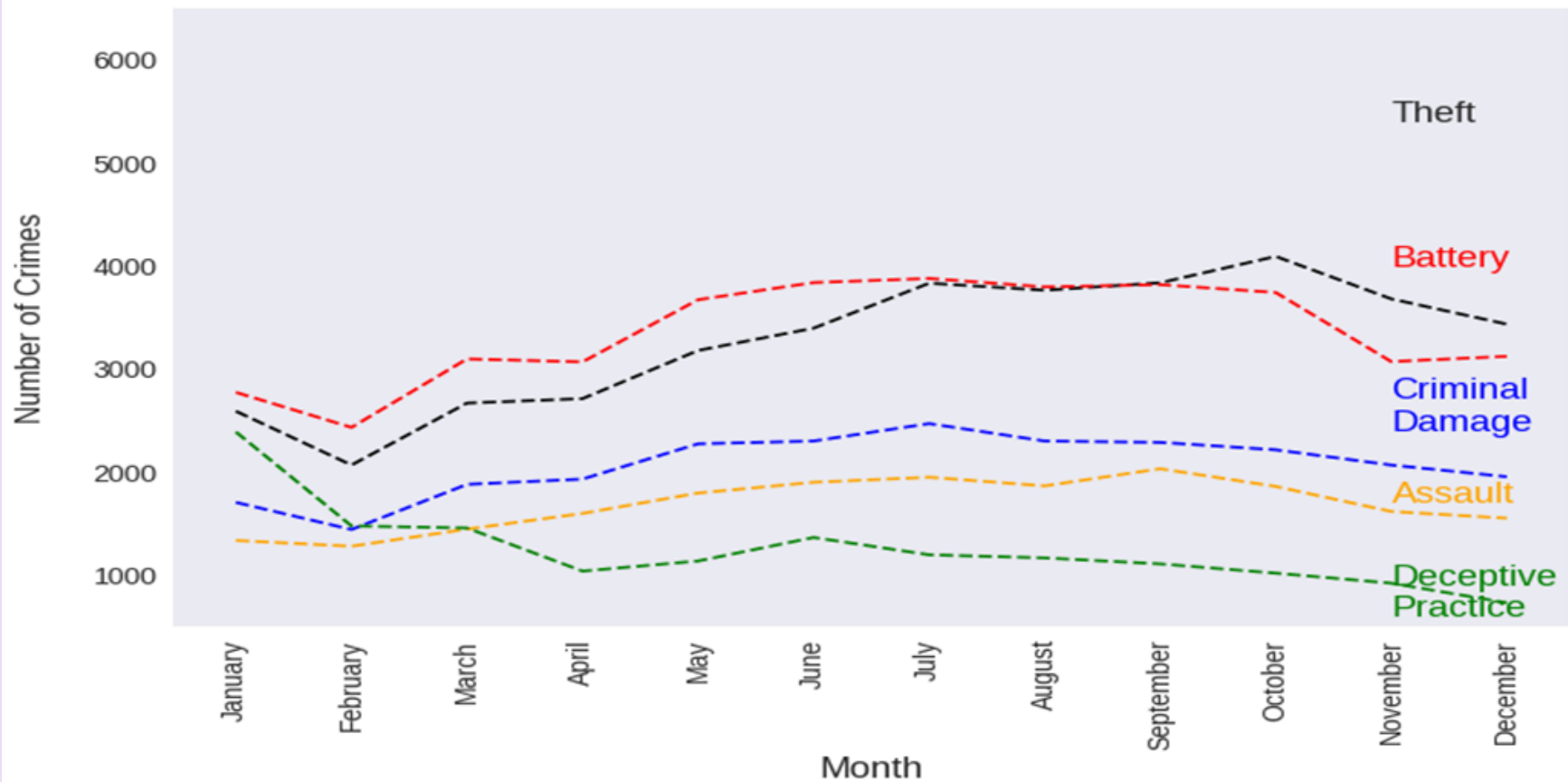

Observations

- Most occurring top 5 crimes in Chicago in 2021
- Modify the date column to a pythonic format
- Convert the date-time column to a known format
- Top crimes in Chicago in the year 2021
- Create a new column month
- The probability of arrest of Chicago
- Distribution of arrests across the months
- What are the most unsafe hours
- Is your house safe from burglary during the day
- Visualize a crime pattern for 24 hours
- Crime & time
- Crime & location
- Most common Occurrences per district
- Which region is given crime concentrated in?
- Visualizing Narcotics (to get an understanding of the most prevalent of drugs in the area)
- Visualizing with heatmaps
- Using Supervised machine learning to predict crime hotspots

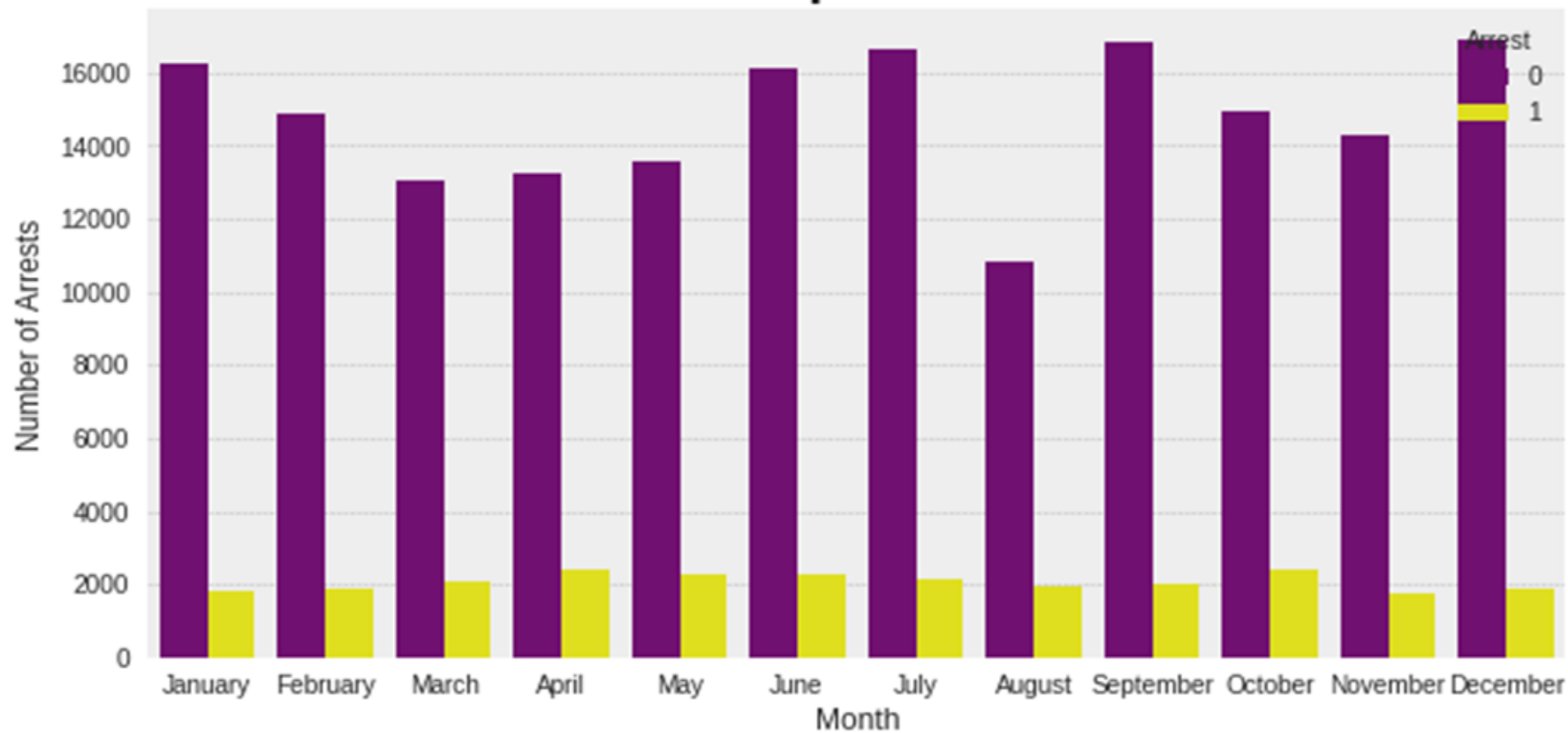
Top 5 Crimes



The frequency of Top 5 Crimes



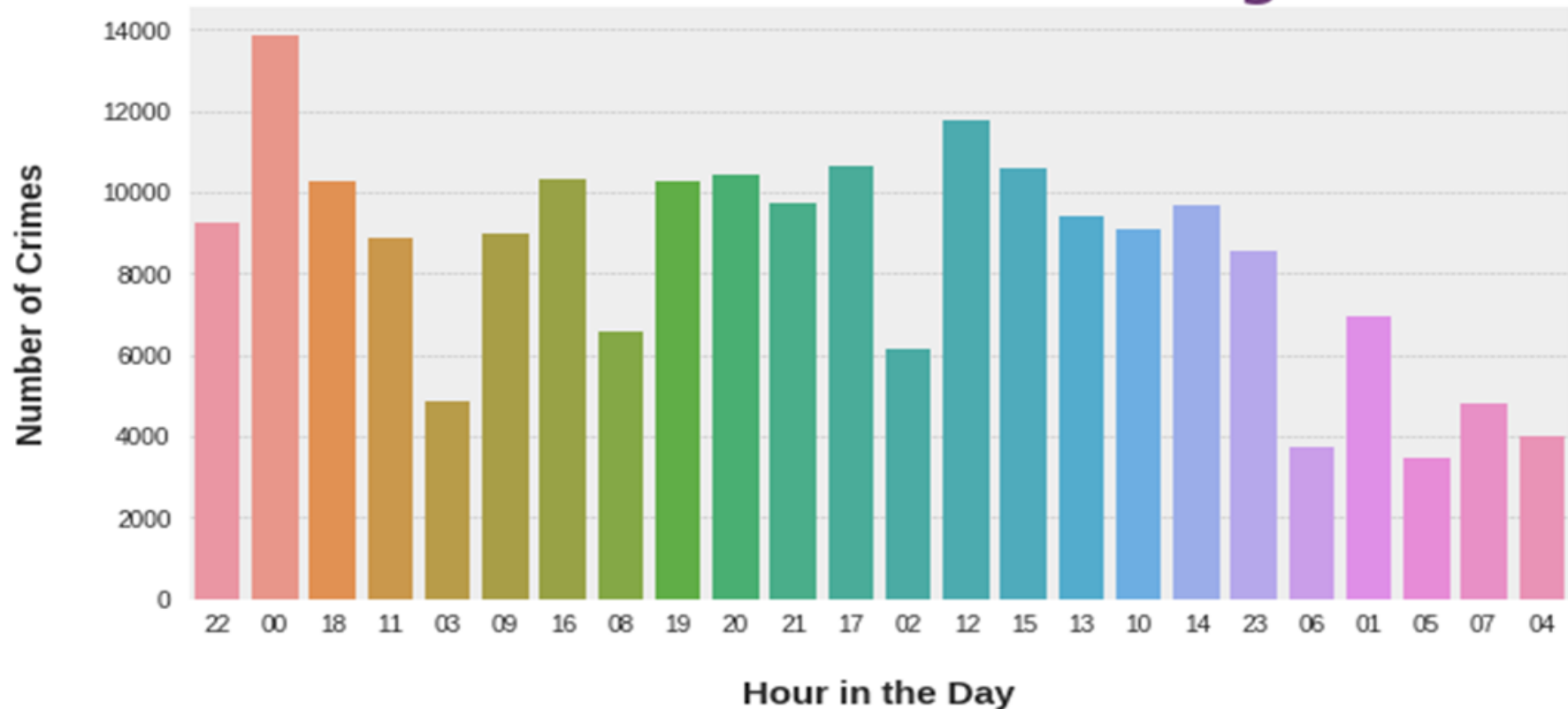
Arrests Made per Month in 2021



Crimes rise during Summer !

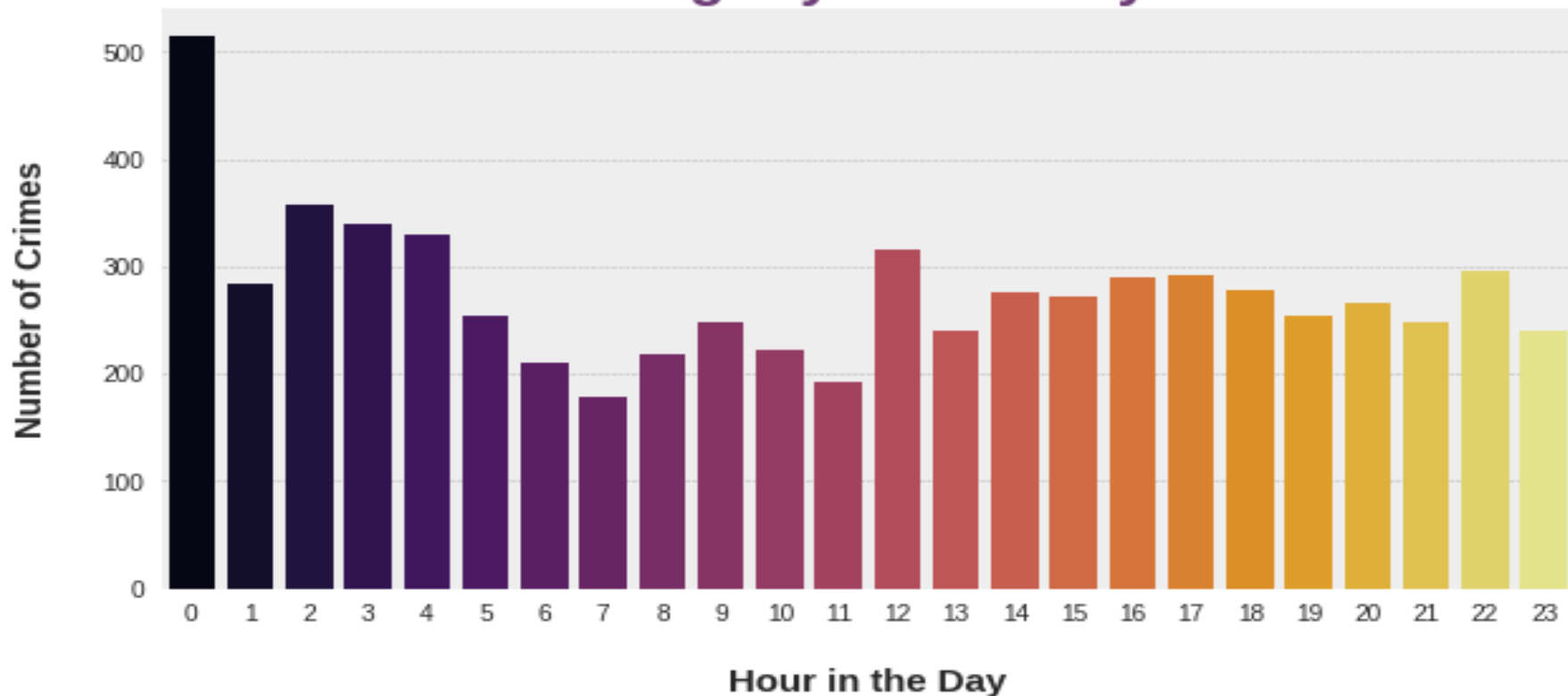


Unsafest Hours in Chicago

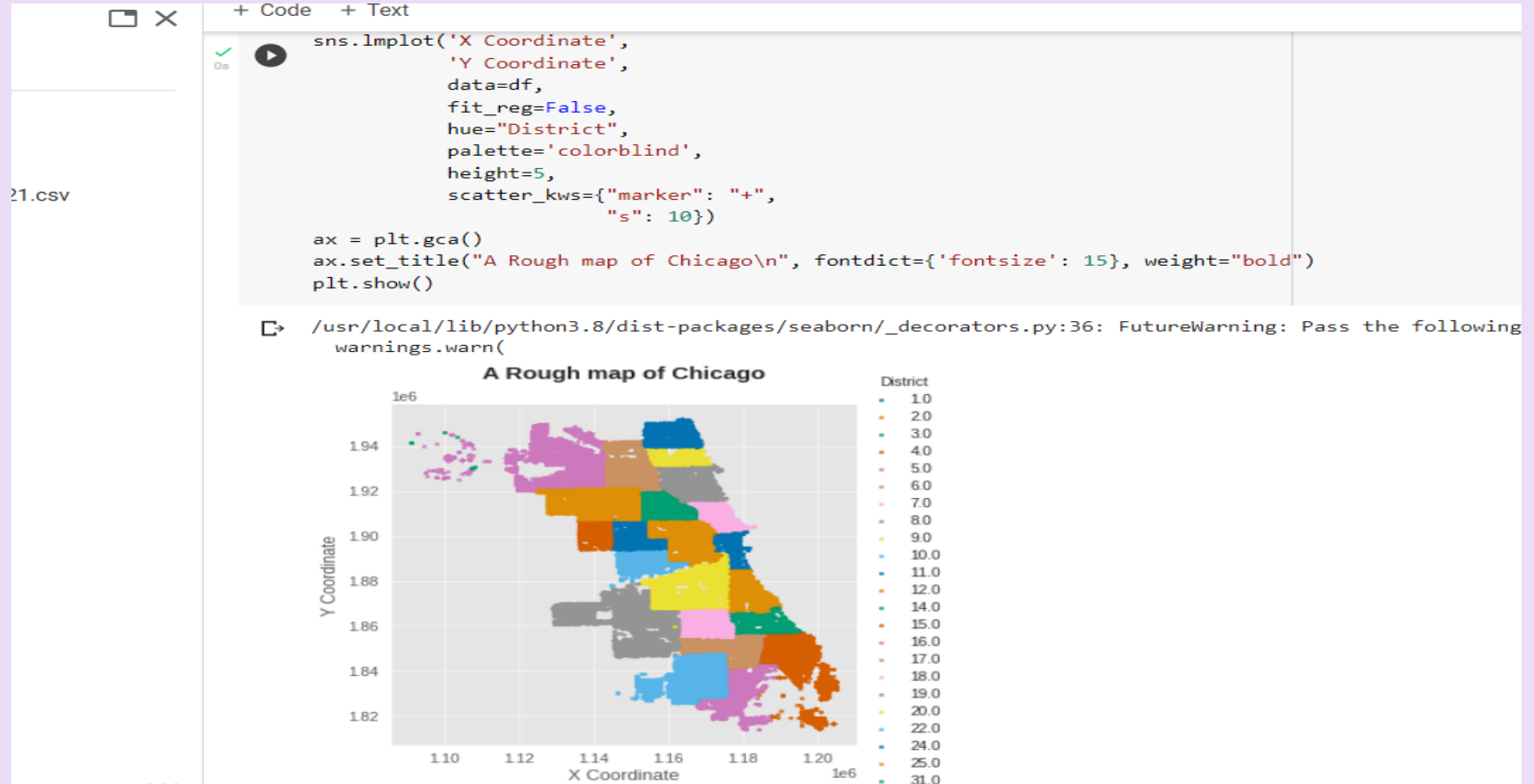


Is the house safe from a burglary during the day ?

Burglary over a day



Crime vs Locations



Heatmap



Supervised Machine Learning

```
Code Text
```

```
[60] 34465    10    5    31.0    7    1
      34464    10    5    31.0    2    1
      497     1    0    31.0    0    1
```

```
[61] cri4 = cri4[['Month_num','Day','Hour','Primary Type','District']]
      cri4.head()
      cri4.shape

      (42047, 5)
```

```
[62] print(cri4['Primary Type'].max(),cri4['Primary Type'].min())

      32 1
```

```
[63] print("Average no. of crime per month per day per district per hour :",cri4['Primary Type'].sum()/42774, ".")

      Average no. of crime per month per day per district per hour : 4.728666947210923 .
```

```
[64] # Feature Engineer and create a new feature
      def crime_rate_assign(x):
          if(x<=7):
              return 0
          elif(x>7 and x<=15):
              return 1
          else:
              return 2
      cri4['Alarm'] = cri4['Primary Type'].apply(crime_rate_assign)
      cri4 = cri4[['Month_num','Day','Hour','District','Primary Type','Alarm']]
      cri4.head()
```

B available

THANK YOU