

# IBM-DataScience-Machine\_Learning-Capstone-Project

February 27, 2021

## 1 IBM-Capstone-Project

### 1.1 Description of the Problem and Discussion of the Background

#### 1.1.1 Introduction Section:

The police of Denver, Colorado patrolling in the whole city. However, various crimes occurred in any part of the city. Due to the limited number of patrolling cars of the police department, most of the area is not covered by the police cars. Therefore, police need to know those areas, where most crimes have occurred. So that, they can cover most of the area by sending police car. The less crime area can be omitted.

To understand these, the police needs answers to several questions. The answers to those questions must be supported by data and analytics. These are their questions:

- 1) How can we find those areas, where the most crimes have occurred?
- 2) How can we find the most offensive crimes and the area?
- 3) How can we find the most traffic area?

#### 1.1.2 Target Audience

The analysis would help any the police department of the city of Denver to predict the place of the crimes and types of the crimes. Moreover, the police can increase/decrease the number of patrolling cars in an area, where it's needed.

#### 1.1.3 Data Description and Data Sources

Data used in this project is collected from several sources. A brief description of the sources of data are given below:

This dataset includes criminal offenses in the City and County of Denver for the previous five calendar years plus the current year to date. The data is based on the National Incident Based Reporting System (NIBRS) which includes all victims of person crimes and all crimes within an incident. The data is dynamic, which allows for additions, deletions and/or modifications at any time, resulting in more accurate information in the database. Due to continuous data entry, the number of records in subsequent extractions are subject to change. Crime data is updated Monday through Friday. The following image shows an some data from the crimes data:

	INCIDENT_ID	OFFENSE_ID	OFFENSE_CODE	OFFENSE_CODE_EXTENSION	OFFENSE_TYPE_ID	OFFENSE_CATEGORY_ID	FIRST_OCCURRENCE_DATE
0	20176005213	20176005213239901	2399		1	theft-bicycle	6/8/2017 1:15:00 PM
1	2016461725	2016461725549900	5499		0	traf-other	7/21/2016 6:40:00 PM
2	2017409119	2017409119549900	5499		0	traf-other	6/22/2017 5:20:00 PM
3	2016829592	2016829592110200	1102		0	sex-aslt-rape	12/30/2016 11:00:00 PM
4	2017455505	2017455505544100	5441		0	traffic-accident	7/10/2017 6:00:00 PM

For more info please check the following link:

<https://www.denvergov.org/opendata/dataset/city-and-county-of-denver-crime>

Foursquare API to explore venue types surrounding each neighborhood of the city of Denver. The query was made for the number of venues in each category within a 1000m radius around each neighborhood (“Documentation — Foursquare Developer”, 2020).

```
[53]: # importing all neccessary dependencies
import pandas as pd
pd.set_option('display.max_columns', None)
pd.set_option('display.max_rows', None)
import numpy as np
import json
import requests
import folium
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
import seaborn as sns
import re
from sklearn.cluster import KMeans
from pandas.io.json import json_normalize #for transforming JSON file into a
    ↪ pandas dataframe
import matplotlib.cm as cm
import matplotlib.colors as colors
```

```
[54]: #loading crimes data of Denver City.
df=pd.read_csv('crime.csv')
df.head()
```

```
[54]: INCIDENT_ID      OFFENSE_ID  OFFENSE_CODE  OFFENSE_CODE_EXTENSION  \
0  20176005213  20176005213239901          2399              1
1  2016461725  2016461725549900          5499              0
2  2017409119  2017409119549900          5499              0
3  2016829592  2016829592110200          1102              0
4  2017455505  2017455505544100          5441              0

OFFENSE_TYPE_ID  OFFENSE_CATEGORY_ID  FIRST_OCCURRENCE_DATE  \
0  theft-bicycle          larceny      6/8/2017 1:15:00 PM
1    traf-other    all-other-crimes      7/21/2016 6:40:00 PM
```

2	traf-other	all-other-crimes	6/22/2017 5:20:00 PM
3	sex-aslt-rape	sexual-assault	12/30/2016 11:00:00 PM
4	traffic-accident	traffic-accident	7/10/2017 6:00:00 PM

	LAST_OCCURRENCE_DATE	REPORTED_DATE	INCIDENT_ADDRESS \
0	6/8/2017 5:15:00 PM	6/12/2017 8:44:00 AM	1705 17TH ST
1	NaN	7/21/2016 7:09:00 PM	N COLUMBINE ST / E 48TH AVE
2	NaN	6/22/2017 5:20:00 PM	E COLFAX AVE / N DOWNING ST
3	12/30/2016 11:45:00 PM	12/31/2016 4:59:00 AM	NaN
4	NaN	7/10/2017 6:45:00 PM	E EVANS AVE / S GRAPE ST

	GEO_X	GEO_Y	GEO_LON	GEO_LAT	DISTRICT_ID	PRECINCT_ID \
0	3140790.0	1699792.0	-104.999264	39.753669	6	612
1	3152605.0	1710822.0	-104.957009	39.783762	2	212
2	3148176.0	1694866.0	-104.973097	39.740032	6	623
3	NaN	NaN	NaN	NaN	6	611
4	3161788.0	1672521.0	-104.925197	39.678463	3	323

	NEIGHBORHOOD_ID	IS_CRIME	IS_TRAFFIC
0	union-station	1	0
1	elyria-swanssea	1	0
2	capitol-hill	1	0
3	five-points	1	0
4	goldsmith	0	1

```
[55]: #Removing some unnecessary columns
df=df.drop(['INCIDENT_ID', 'OFFENSE_ID', 'OFFENSE_CODE',
↳ 'OFFENSE_CODE_EXTENSION', 'PRECINCT_ID'], axis=1)
df.head()
```

	OFFENSE_TYPE_ID	OFFENSE_CATEGORY_ID	FIRST_OCCURRENCE_DATE \
0	theft-bicycle	larceny	6/8/2017 1:15:00 PM
1	traf-other	all-other-crimes	7/21/2016 6:40:00 PM
2	traf-other	all-other-crimes	6/22/2017 5:20:00 PM
3	sex-aslt-rape	sexual-assault	12/30/2016 11:00:00 PM
4	traffic-accident	traffic-accident	7/10/2017 6:00:00 PM

	LAST_OCCURRENCE_DATE	REPORTED_DATE	INCIDENT_ADDRESS \
0	6/8/2017 5:15:00 PM	6/12/2017 8:44:00 AM	1705 17TH ST
1	NaN	7/21/2016 7:09:00 PM	N COLUMBINE ST / E 48TH AVE
2	NaN	6/22/2017 5:20:00 PM	E COLFAX AVE / N DOWNING ST
3	12/30/2016 11:45:00 PM	12/31/2016 4:59:00 AM	NaN
4	NaN	7/10/2017 6:45:00 PM	E EVANS AVE / S GRAPE ST

	GEO_X	GEO_Y	GEO_LON	GEO_LAT	DISTRICT_ID	NEIGHBORHOOD_ID \
0	3140790.0	1699792.0	-104.999264	39.753669	6	union-station
1	3152605.0	1710822.0	-104.957009	39.783762	2	elyria-swanssea

2	3148176.0	1694866.0	-104.973097	39.740032	6	capitol-hill
3	NaN	NaN	NaN	NaN	6	five-points
4	3161788.0	1672521.0	-104.925197	39.678463	3	goldsmith

	IS_CRIME	IS_TRAFFIC
0	1	0
1	1	0
2	1	0
3	1	0
4	0	1

```
[56]: #dropping NaN values from the dataframe
df.dropna(inplace=True)
df.head()
```

```
[56]:
```

	OFFENSE_TYPE_ID	OFFENSE_CATEGORY_ID	FIRST_OCCURRENCE_DATE	\
0	theft-bicycle	larceny	6/8/2017 1:15:00 PM	
7	burglary-residence-no-force	burglary	6/29/2017 8:00:00 AM	
9	criminal-trespassing	all-other-crimes	6/7/2016 8:00:00 PM	
17	theft-of-motor-vehicle	auto-theft	5/10/2016 3:00:00 PM	
19	burglary-residence-by-force	burglary	1/31/2016 7:50:00 PM	

	LAST_OCCURRENCE_DATE	REPORTED_DATE	\
0	6/8/2017 5:15:00 PM	6/12/2017 8:44:00 AM	
7	7/5/2017 6:30:00 PM	7/5/2017 9:54:00 PM	
9	6/8/2016 10:20:00 AM	6/8/2016 3:52:00 PM	
17	5/10/2016 3:15:00 PM	5/10/2016 3:50:00 PM	
19	2/1/2016 1:30:00 AM	2/1/2016 3:08:00 AM	

	INCIDENT_ADDRESS	GEO_X	GEO_Y	GEO_LON	\
0	1705 17TH ST	3140790.0	1699792.0	-104.999264	
7	5004 N STUART ST	3128600.0	1712081.0	-105.042398	
9	3291 N OSCEOLA ST	3129744.0	1703432.0	-105.038484	
17	E HARVARD AVE / S MILWAUKEE ST	3154183.0	1669827.0	-104.952274	
19	1488 N MADISON ST	3155797.0	1694829.0	-104.945998	

	GEO_LAT	DISTRICT_ID	NEIGHBORHOOD_ID	IS_CRIME	IS_TRAFFIC
0	39.753669	6	union-station	1	0
7	39.787581	1	regis	1	0
9	39.763822	1	west-highland	1	0
17	39.671197	3	university-park	1	0
19	39.739806	2	congress-park	1	0

```
[57]: #Shape of the dataframe
df.shape
```

```
[57]: (48557, 14)
```

```
[58]: df.columns = df.columns.str.replace(' ', '')
```

```
[60]: df=df.rename(columns={"OFFENSE_TYPE_ID": "Offence Types", "GEO_LON":_
    ↳"Longitude",
    "GEO_LAT": "Latitude", "NEIGHBORHOOD_ID": "Neighborhood"})
df.head()
```

```
[60]:
```

	Offence Types	OFFENSE_CATEGORY_ID	FIRST_OCCURRENCE_DATE	\
0	theft-bicycle	larceny	6/8/2017 1:15:00 PM	
7	burglary-residence-no-force	burglary	6/29/2017 8:00:00 AM	
9	criminal-trespassing	all-other-crimes	6/7/2016 8:00:00 PM	
17	theft-of-motor-vehicle	auto-theft	5/10/2016 3:00:00 PM	
19	burglary-residence-by-force	burglary	1/31/2016 7:50:00 PM	

	LAST_OCCURRENCE_DATE	REPORTED_DATE	\
0	6/8/2017 5:15:00 PM	6/12/2017 8:44:00 AM	
7	7/5/2017 6:30:00 PM	7/5/2017 9:54:00 PM	
9	6/8/2016 10:20:00 AM	6/8/2016 3:52:00 PM	
17	5/10/2016 3:15:00 PM	5/10/2016 3:50:00 PM	
19	2/1/2016 1:30:00 AM	2/1/2016 3:08:00 AM	

	INCIDENT_ADDRESS	GEO_X	GEO_Y	Latitude	\
0	1705 17TH ST	3140790.0	1699792.0	-104.999264	
7	5004 N STUART ST	3128600.0	1712081.0	-105.042398	
9	3291 N OSCEOLA ST	3129744.0	1703432.0	-105.038484	
17	E HARVARD AVE / S MILWAUKEE ST	3154183.0	1669827.0	-104.952274	
19	1488 N MADISON ST	3155797.0	1694829.0	-104.945998	

	Longitude	DISTRICT_ID	Neighborhood	IS_CRIME	IS_TRAFFIC
0	39.753669	6	union-station	1	0
7	39.787581	1	regis	1	0
9	39.763822	1	west-highland	1	0
17	39.671197	3	university-park	1	0
19	39.739806	2	congress-park	1	0

```
[63]: df['Neighborhood']=df['Neighborhood'].replace({'\-' : ' '}, regex = True)
df['Neighborhood']=df['Neighborhood'].str.title()
df.head()
```

```
[63]:
```

	Offence Types	OFFENSE_CATEGORY_ID	FIRST_OCCURRENCE_DATE	\
0	theft-bicycle	larceny	6/8/2017 1:15:00 PM	
7	burglary-residence-no-force	burglary	6/29/2017 8:00:00 AM	
9	criminal-trespassing	all-other-crimes	6/7/2016 8:00:00 PM	
17	theft-of-motor-vehicle	auto-theft	5/10/2016 3:00:00 PM	
19	burglary-residence-by-force	burglary	1/31/2016 7:50:00 PM	

	LAST_OCCURRENCE_DATE	REPORTED_DATE	\
--	----------------------	---------------	---

```

0    6/8/2017 5:15:00 PM  6/12/2017 8:44:00 AM
7    7/5/2017 6:30:00 PM  7/5/2017 9:54:00 PM
9    6/8/2016 10:20:00 AM  6/8/2016 3:52:00 PM
17   5/10/2016 3:15:00 PM  5/10/2016 3:50:00 PM
19   2/1/2016 1:30:00 AM   2/1/2016 3:08:00 AM

```

```

          INCIDENT_ADDRESS      GEO_X      GEO_Y      Latitude \
0          1705 17TH ST  3140790.0  1699792.0 -104.999264
7          5004 N STUART ST  3128600.0  1712081.0 -105.042398
9          3291 N OSCEOLA ST  3129744.0  1703432.0 -105.038484
17  E HARVARD AVE / S MILWAUKEE ST  3154183.0  1669827.0 -104.952274
19          1488 N MADISON ST  3155797.0  1694829.0 -104.945998

```

```

          Longitude  DISTRICT_ID      Neighborhood  IS_CRIME  IS_TRAFFIC
0    39.753669           6      Union Station           1           0
7    39.787581           1           Regis             1           0
9    39.763822           1      West Highland           1           0
17   39.671197           3  University Park             1           0
19   39.739806           2    Congress Park             1           0

```

```

[65]: # create map for Denver City
map_denver = folium.Map(location=[39.744137, -104.950050], zoom_start=10)

# add markers to map
for lat, lng, neighborhood in zip(df['Longitude'], df['Latitude'],
    ↪df['Neighborhood']):
    label = '{}'.format(neighborhood)
    label = folium.Popup(label, parse_html=True)
    folium.CircleMarker(
        [lat, lng],
        radius=5,
        popup=label,
        color='blue',
        fill=True,
        fill_color='#3186cc',
        fill_opacity=0.7,
        parse_html=False).add_to(map_denver)

map_denver

```

```

[65]: <folium.folium.Map at 0x1b3c49e51c8>

```

```

[ ]:

```