

**Capstone Project - The Battle of  
Neighborhoods  
Report**

## **1. Introduction**

### **1.1 Background**

The people will relocate to different places as and when they change jobs or their business. The average American relocate around 11 times in their lifetime. When people need to relocate, they search for lot of information about locations in the city and different types of venues in nearby to the location.

As relocation is a tidy, expensive and most difficult task, people will not take a chance by relocate first and check for safety and venues nearby. People will look for the information whether the place is good in terms of safety and venues nearby.

### **1.2 Problem**

The crime statistics dataset of Denver on Kaggle has crimes in each Neighborhood of Denver from 2016 to 2019. We will consider the year 2019 which is latest for this analysis.

This project aims to identify the safest cluster in Denver based on total crimes in the neighborhoods, nearby venues including 10 most common venues in each neighborhood. And finally Cluster the venues & neighborhoods using k-mean clustering.

### **1.3 Interest**

This will help the people who are interested in relocating to Denver in identifying the safest Neighborhood and explore nearest venues for each of the Neighborhoods.

## **2. Data Acquisition and Cleaning**

### **2.1 Data Acquisition**

The first data source of the project uses Denver crime data from kaggle website that shows the crime data in each Neighborhood in London.

The dataset contains the following columns:

- OFFENSE\_TYPE\_ID: which gives which type of offense it is.
- NEIGHBORHOOD\_ID: This will give the Neighborhood in which offense reported
- IS\_CRIME : This showing the offense is a crime
- Is\_TRAFFIC: which shows if the offense is traffic offense
- REPORTED\_DATE: This include crime reported date & time

The second data source of data Latitude & Longitude is extracted using geocoder for each Neighborhood.

The third data source is the list of nearby venues for Neighborhoods of the Denver and Venues information using Foursquare.

Leveraged Foursquare to extract the following data.

- Venue Name
- Venue Latitude
- Venue Longitude
- Venue category

## 2.2 Data Cleaning

From the Denver crime data, the crimes during the most recent year 2019 are only selected.

	OFFENSE_CODE	OFFENSE_TYPE_ID	OFFENSE_CATEGORY_ID	NEIGHBORHOOD_ID	IS_CRIME	year
1568	1208	robbery-residence	robbery	east-colfax	1	2019
23326	5499	traf-other	all-other-crimes	jefferson-park	1	2019
23327	2404	theft-of-motor-vehicle	auto-theft	west-highland	1	2019
23328	2303	theft-shoplift	larceny	city-park	1	2019
23329	2999	criminal-mischief-mtr-veh	public-disorder	montbello	1	2019

The major categories of crime are pivoted to get the total crimes per Neighborhood as per category.

	Neighborhood	aggravated-assault	all-other-crimes	Earson	auto-theft	burglary	drug-alcohol	larceny	murder	other-crimes-against-persons	public-disorder	robbery	sexual-assault	theft-from-motor-vehicle	white-collar-crime	Total
0	athmar-park	33	207	1	63	28	98	58	1	48	81	16	9	86	21	750
1	auraria	15	271	1	19	24	28	97	0	52	100	11	8	35	11	672
2	baker	28	227	6	72	49	120	262	0	55	133	21	7	123	20	1123
3	barnum	35	212	2	40	27	47	31	1	46	91	17	10	51	11	621
4	barnum-west	13	100	3	41	14	42	11	1	16	60	2	4	38	4	349

The second data of Neighborhood Latitude and Longitude data extracted using Geocoder.

neighborhood	aggravated-assault	all-other-crimes	Earson	auto-theft	burglary	drug-alcohol	larceny	murder	other-crimes-against-persons	public-disorder	robbery	sexual-assault	theft-from-motor-vehicle	white-collar-crime	Total	latitude
wellshire	1	4	0	6	12	3	5	0	4	13	0	0	18	2	68	39.658484
indian-creek	1	6	0	9	4	1	11	1	5	13	3	0	26	4	84	39.684761
country-club	3	22	0	12	34	2	17	0	3	11	3	0	33	1	141	39.720100
skyland	6	69	1	13	11	8	16	2	11	22	1	3	26	5	194	39.755713
rosedale	4	30	0	17	19	21	26	0	7	23	1	4	49	5	206	39.739236

The third data of nearby venues of Neighborhood using Foursquare API.

	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category	Cluster Label	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue
0	wellshire	39.658484	-104.941353	Wellshire Golf Course	39.658276	-104.941329	Golf Course	2	Golf Course	Athletics & Sports	Playground	Park	Yc Stu
1	wellshire	39.658484	-104.941353	Mamie D. Eisenhower Park	39.658946	-104.937348	Park	2	Golf Course	Athletics & Sports	Playground	Park	Yc Stu
2	wellshire	39.658484	-104.941353	Wellshire Inn	39.658247	-104.941277	Golf Course	2	Golf Course	Athletics & Sports	Playground	Park	Yc Stu
3	wellshire	39.658484	-104.941353	Eisenhower Rec Center	39.659563	-104.937773	Athletics & Sports	2	Golf Course	Athletics & Sports	Playground	Park	Yc Stu
4	wellshire	39.658484	-104.941353	Eisenhower Park	39.659608	-104.936572	Playground	2	Golf Course	Athletics & Sports	Playground	Park	Yc Stu

### 3. Methodology

#### 3.1 Exploratory Data Analysis

##### 3.1.1 Statistical summary of crimes

The describe function in python is used to get statistics of the Denver crime data, this returns the mean, standard deviation, minimum, 1st quartile (25%), 2nd quartile(50%), and 3rd quartile (75%) for each of the major categories of crime (See fig 3.1.1)

	aggravated-assault	all-other-crimes	Earson	auto-theft	burglary	drug-alcohol	larceny	murder	other-crimes-against-persons	public-disorder	robbery	sexual-assault	the
count	78.000000	78.000000	78.000000	78.000000	78.000000	78.000000	78.000000	78.000000	78.000000	78.000000	78.000000	78.000000	78
mean	32.294872	187.717949	1.25641	68.910256	51.589744	64.576923	125.692308	0.897436	59.705128	108.333333	15.294872	10.974359	104
std	36.949284	205.284212	1.59085	52.556756	33.051096	83.519267	165.732715	1.123419	69.044332	96.657068	15.843479	10.813621	71
min	1.000000	4.000000	0.00000	6.000000	4.000000	1.000000	5.000000	0.000000	2.000000	11.000000	0.000000	0.000000	18
25%	12.250000	59.000000	0.00000	38.250000	32.250000	16.250000	49.250000	0.000000	22.000000	52.000000	6.000000	5.000000	65
50%	21.000000	102.500000	1.00000	54.500000	44.000000	37.000000	70.500000	1.000000	40.000000	79.500000	10.500000	7.000000	87
75%	37.250000	228.500000	2.00000	82.000000	60.750000	79.250000	131.500000	1.000000	57.500000	120.000000	19.750000	12.500000	124
max	243.000000	1013.000000	7.00000	259.000000	202.000000	475.000000	959.000000	5.000000	419.000000	626.000000	87.000000	65.000000	483

Fig 3.1.1 Statistical description of the Denver crimes

The count for each of the major categories of crime returns the value 78 which is the number of Denver neighborhoods. 'larceny' is the highest reported crime during the year 2019 followed by 'public-disorder' and 'theft-from-motor-vehicle' offenses. The lowest recorded crimes are 'murder', 'Earson' & 'sexual-assault' offenses

### 3.1.2 Neighborhoods with the highest crime rates

Comparing five neighborhoods with the highest crime rate during the year 2019 it is evident that 'five-points' has the highest crimes recorded followed by stapleton, cbd, capitalhill & montbello. (see fig 3.1.2)

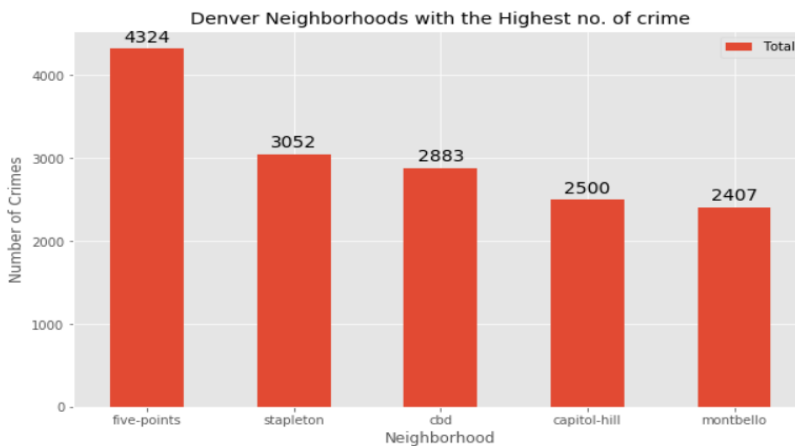


Fig 3.1.2 Neighborhoods with the highest crime rates

### 3.1.3 Neighborhoods with the lowest crime rates

Comparing five neighborhoods with the lowest crime rate during the year 2019, 'Wellshire' has the lowest recorded crimes followed by indian-creek, country-club, skyland and rosedale... (see fig 3.1.3)

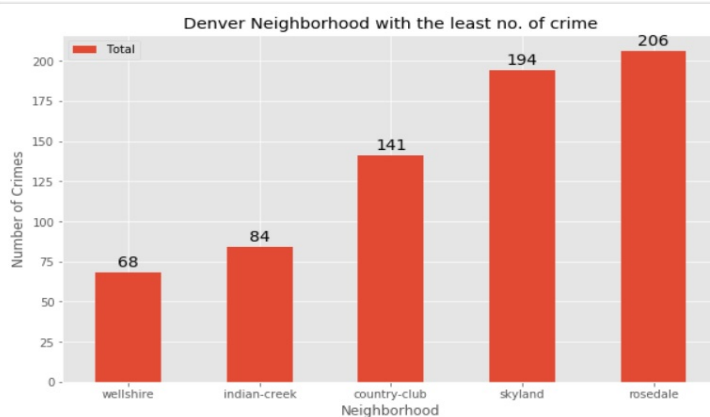


Fig 3.1.3 Neighborhoods with lowest crime rates

Both 'Wellshire' and indian-creeks' have the significantly lower crime rate compare to other neighborhoods.

3.1.4 The lowest crime neighborhood visualized on a map using folium on python (see fig 3.1.4)

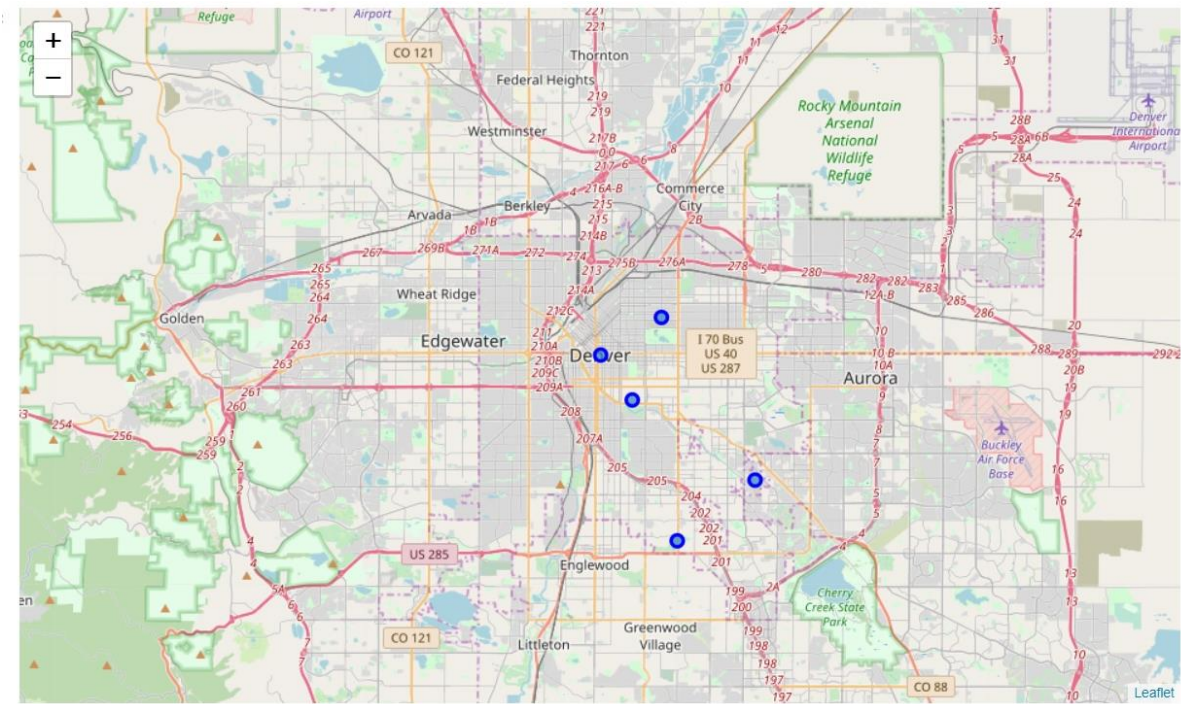


Fig 3.1.4 Lowest crime neighborhoods in Denver

3.2 Modelling

Using the final dataset containing the neighborhoods in Denver along with latitude and longitude, we can find all the venues within a 500 meter radius of each neighborhood by connecting to the Foursquare API. This returns a json file containing all the venues in each neighborhood which is converted to a pandas data frame. This data frame contains all the venues along with their coordinates and category (see fig 3.2.1)

	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category	Cluster Label	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue
0	wellshire	39.658484	-104.941353	Wellshire Golf Course	39.658276	-104.941329	Golf Course	2	Golf Course	Athletics & Sports	Playground	Park	Yc Stu
1	wellshire	39.658484	-104.941353	Mamie D. Eisenhower Park	39.658946	-104.937348	Park	2	Golf Course	Athletics & Sports	Playground	Park	Yc Stu
2	wellshire	39.658484	-104.941353	Wellshire Inn	39.658247	-104.941277	Golf Course	2	Golf Course	Athletics & Sports	Playground	Park	Yc Stu
3	wellshire	39.658484	-104.941353	Eisenhower Rec Center	39.659563	-104.937773	Athletics & Sports	2	Golf Course	Athletics & Sports	Playground	Park	Yc Stu
4	wellshire	39.658484	-104.941353	Eisenhower Park	39.659608	-104.936572	Playground	2	Golf Course	Athletics & Sports	Playground	Park	Yc Stu

Fig 3.2.1 Venue details of each Neighborhood

One hot encoding is done on venues data. (One hot encoding is a process by which categorical variables are converted into a form that could be provided in ML algorithms to do a better job in prediction). The Venues data is then grouped by Neighborhood and the mean of venues are calculated, finally the 10 common venues are calculated for each of the neighborhoods.

To help people find similar neighborhoods in the safest borough we will be clustering similar neighborhoods using K-means clustering which is a form of unsupervised machine learning algorithm that clusters data based on predefined cluster size. We will use cluster size of 4 for this project. The reason to conduct a K-means clustering is to cluster neighborhoods with similar venues together so that people can shortlist the area of their interests based on the venues/amenities around each neighborhood.

## 4. Results

After running the K-means clustering we can access each cluster created to see which neighborhoods were assigned to each 4 clusters. Looking into the neighborhoods in the first cluster (see fig 4.1)

	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category	Cluster Label	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue
5	country-club	39.7201	-104.9667	Denver Country Club	39.717805	-104.966252	Golf Course	0	Health & Beauty Service	Golf Course	Yoga Studio	Carpet Store	Gym
6	country-club	39.7201	-104.9667	Namaste Health Care	39.718477	-104.969934	Health & Beauty Service	0	Health & Beauty Service	Golf Course	Yoga Studio	Carpet Store	Gym

Fig 4.1 Cluster 1

The first Neighborhood cluster has two Venues which are Golf Course and Health & Beauty Service

Looking into the venues in the second cluster (See Fig 4.2)

	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category	Cluster Label	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue
11	rosedale	39.739236	-104.984862	Sassafras American Eatery	39.739949	-104.982756	Breakfast Spot	1	Yoga Studio	Noodle House	Marijuana Dispensary	Art Museum	Ballroom
12	rosedale	39.739236	-104.984862	Pho-natic	39.740081	-104.984111	Noodle House	1	Yoga Studio	Noodle House	Marijuana Dispensary	Art Museum	Ballroom
13	rosedale	39.739236	-104.984862	City, O' City	39.736724	-104.984669	Vegetarian / Vegan Restaurant	1	Yoga Studio	Noodle House	Marijuana Dispensary	Art Museum	Ballroom
14	rosedale	39.739236	-104.984862	Capitol Hill Books	39.739979	-104.983472	Bookstore	1	Yoga Studio	Noodle House	Marijuana Dispensary	Art Museum	Ballroom
15	rosedale	39.739236	-104.984862	Shish Kabob Grill	39.740246	-104.983633	Middle Eastern Restaurant	1	Yoga Studio	Noodle House	Marijuana Dispensary	Art Museum	Ballroom
16	rosedale	39.739236	-104.984862	Kindness Yoga	39.736721	-104.984407	Yoga Studio	1	Yoga Studio	Noodle House	Marijuana Dispensary	Art Museum	Ballroom
17	rosedale	39.739236	-104.984862	Good Chemistry - Denver Dispensary	39.739907	-104.982668	Marijuana Dispensary	1	Yoga Studio	Noodle House	Marijuana Dispensary	Art Museum	Ballroom
18	rosedale	39.739236	-104.984862	CorePower Yoga	39.737070	-104.983057	Yoga Studio	1	Yoga Studio	Noodle House	Marijuana Dispensary	Art Museum	Ballroom
19	rosedale	39.739236	-104.984862	City Grille	39.740165	-104.982861	Burger Joint	1	Yoga Studio	Noodle House	Marijuana Dispensary	Art Museum	Ballroom
20	rosedale	39.739236	-104.984862	Civic Center Park	39.739370	-104.988776	Park	1	Yoga Studio	Noodle House	Marijuana Dispensary	Art Museum	Ballroom

Fig 4.2 Cluster 2

The second Neighborhood cluster is the biggest cluster with consists of 40 venues nearby. the venues includes Restaurants, Yoga centers, nightclubs, Parks....etc

	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category	Cluster Label	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue
0	wellshire	39.658484	-104.941353	Wellshire Golf Course	39.658276	-104.941329	Golf Course	2	Golf Course	Athletics & Sports	Playground	Park	Yc Stu
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3	wellshire	39.658484	-104.941353	Eisenhower Rec Center	39.659563	-104.937773	Athletics & Sports	2	Golf Course	Athletics & Sports	Playground	Park	Yc Stu
4	wellshire	39.658484	-104.941353	Eisenhower Park	39.659608	-104.936572	Playground	2	Golf Course	Athletics & Sports	Playground	Park	Yc Stu

Fig 4.3 Cluster 3

The third Neighborhood cluster has five venues nearby which include Golf Course, Park, Rec Center & Play Ground.

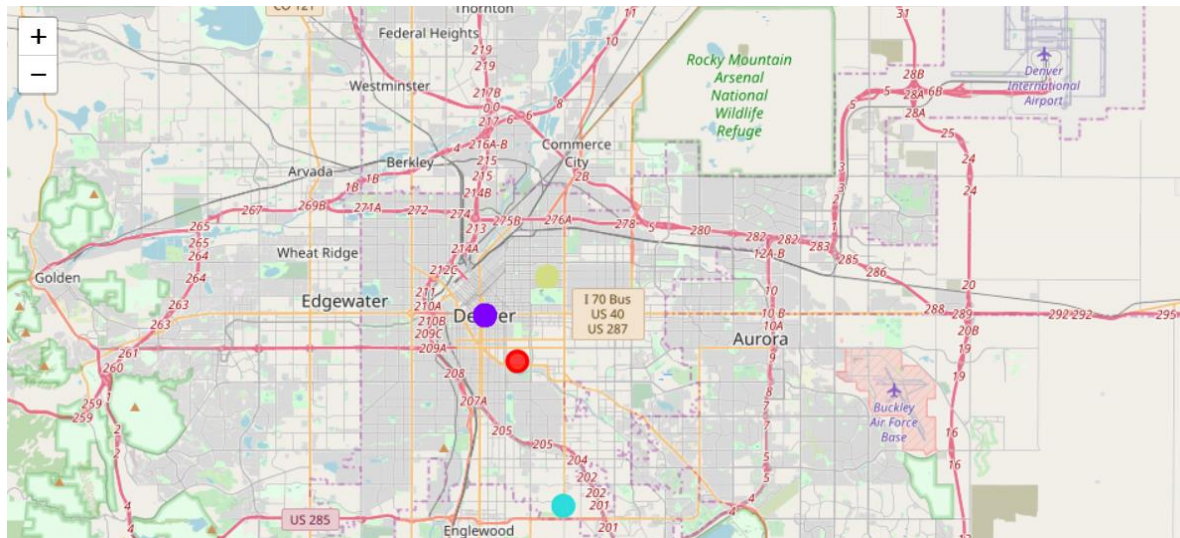
	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category	Cluster Label	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue
7	skyland	39.755713	-104.950224	Street Food Warriors	39.754849	-104.947826	Food Truck	3	Carpet Store	Shop & Service	Food Truck	Nightlife Spot	Yoga Studio
8	skyland	39.755713	-104.950224	Owl Club of Denver	39.757190	-104.946836	Nightlife Spot	3	Carpet Store	Shop & Service	Food Truck	Nightlife Spot	Yoga Studio
9	skyland	39.755713	-104.950224	East 28th Liquor	39.757153	-104.946129	Shop & Service	3	Carpet Store	Shop & Service	Food Truck	Nightlife Spot	Yoga Studio
10	skyland	39.755713	-104.950224	Modern and Custom Rugs by Doris	39.757791	-104.955253	Carpet Store	3	Carpet Store	Shop & Service	Food Truck	Nightlife Spot	Yoga Studio

Fig 4.4 Cluster 4

The fourth Neighborhood cluster has four venues nearby which include Street Food, Nightlife spot, shop & a carpet store .



Visualizing the clustered neighborhoods on a map using the folium library (see fig 4.6).



#### 4.6 Clustered neighborhoods in Denver

Each cluster is color coded for ease of presentation. Four clusters have four different color codes.

### 5. Results Discussion

The aim of this project is to help people who want to relocate to the safest neighborhood in Denver, expats can choose the neighborhoods to which they want to relocate based on the most common venues in it. For example, if a person is looking for a neighborhood with good connectivity and public transportation, we can see that Cluster 2 has the most common venues. If a person is looking for a neighborhood with stores and sports centers in a close proximity then the neighborhoods in the third cluster is suitable.

### 6. Conclusion

This project helps a person get a better understanding of the neighborhoods with respect to the most common venues in that neighborhood. It is always helpful to make use of technology to stay one step ahead i.e. finding out more about places before moving into a neighborhood. We have just taken safety as a primary concern to shortlist the Neighborhoods of Denver. The future of this project includes taking other factors such as cost of living in the areas into consideration to shortlist the neighborhoods based on safety and a predefined budget