CAPSTONE

The Battle of Neighborhoods - Presentation

1. Introduction

Description Of The Problem.

- London is a developed country, still they face more crime cases in day to day life. Still there are least crime cases in few cities and also they are safe to stay. They even attract tourist people to visit these least crime place as more venues in and around these areas.
- The crime statistics dataset of London found on Kaggle has crimes in each boroughs of London from 2008-2016. The crime rates in each Borough may have changed over time. In past Borough was least crime rated and also have more Attractive places nearby. The year 2016 being the latest we will be considering the data of that year which is actually old information as of now.
- This battle of neighborhood aims to select the safest place is Borough in London based on the total crimes. We can also explore the neighborhood of that Borough to find the 10 most common venues in each neighborhood and finally cluster the neighborhood using k-means clustering.

2. Data Acquisition and Cleaning

- London is one of the biggest IT hub and Borough has more companies and offices and tourist attractive places. whenever a person moves to the city for a job or have a tour, such a person search for venues in new city. The person might want to know how good a given restaurant is or the price range it falls under. combining the location of the venues in the city with their price and rating information would surely help visitors in a city make better informed decisions about the places they should visit.
- In Borough west, there are so many venues (especially restaurants, cafes, hotels) explored. To explore this information, this project involves data from foursquare API to fetch complete information on various venues (including name, address, rating, price and category). Further a map of venues with colour specification indicated for highlighting the postion and information about the venues.

Data Cleaning:

The data preparation for each of the three sources of data is done separately. From the London crime data, the crimes during the most recent year (2016) are only selected. The major categories of crime are pivoted to get the total

| | Borough | Burglary | Criminal Damage | Drugs | Other Notifiable Offences | Robbery | Theft and Handling | Violence Against the Person | Total |
|---|----------------------|----------|-----------------|-------|---------------------------|---------|--------------------|-----------------------------|-------|
| 0 | Barking and Dagenham | 1287 | 1949 | 919 | 378 | 534 | 5607 | 6067 | 16741 |
| 1 | Barnet | 3402 | 2183 | 906 | 499 | 464 | 9731 | 7499 | 24684 |
| 2 | Bexley | 1123 | 1673 | 646 | 294 | 209 | 4392 | 4503 | 12840 |
| 3 | Brent | 2631 | 2280 | 2096 | 536 | 919 | 9026 | 9205 | 26693 |
| 4 | Bromley | 2214 | 2202 | 728 | 417 | 369 | 7584 | 6650 | 20164 |

The second data is scraped from a wikipedia page using the **Beautiful Soup** library in python. Using this library we can extract the data in the tabular format as shown in the website.

3. Methodology

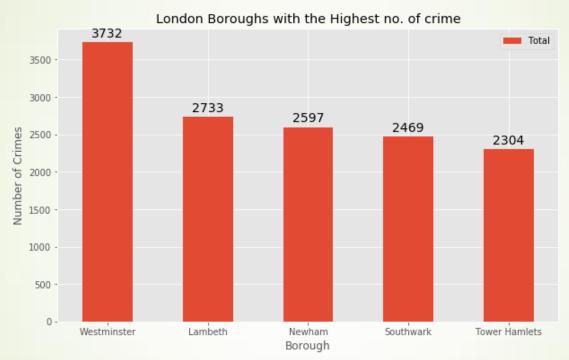
Exploratory Data Analysis

Statistical summary of crimes

| | Burglary | Criminal Damage | Drugs | Other Notifiable Offences | Robbery | Theft and Handling | Violence Against the Person | Total |
|-------|-------------|-----------------|-------------|---------------------------|-------------|--------------------|-----------------------------|--------------|
| count | 33.000000 | 33.000000 | 33.000000 | 33.000000 | 33.000000 | 33.000000 | 33.000000 | 33.000000 |
| mean | 2069242424 | 1941.545455 | 1179.212121 | 479.060606 | 682.666667 | 8913.121212 | 7041.848485 | 22306.696970 |
| std | 737.448644 | 625.207070 | 586.406416 | 223.298698 | 441.425366 | 4620.565054 | 2513.601551 | 8828.228749 |
| min | 2.000000 | 2.000000 | 10.000000 | 6.000000 | 4.000000 | 129.000000 | 25.000000 | 178.000000 |
| 25% | 1531.000000 | 1650.000000 | 743.000000 | 378.000000 | 377.000000 | 5919.000000 | 5936.000000 | 16903000000 |
| 50% | 2071.000000 | 1989.000000 | 1063.000000 | 490.000000 | 599.000000 | 8925.000000 | 7409.000000 | 22730000000 |
| 75% | 2631.000000 | 2351.000000 | 1617.000000 | 551.000000 | 936.000000 | 10789.000000 | 8832.000000 | 27174.000000 |
| max | 3402.000000 | 3219.000000 | 2738.000000 | 1305.000000 | 1822.000000 | 27520.000000 | 10834.000000 | 48330.000000 |

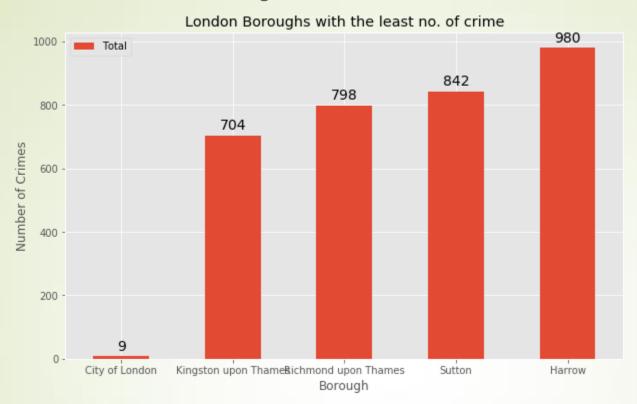
The count for each of the major categories of crime returns the value 33which is the number of London boroughs. 'Theft and Handling' is the highest reported crime during the year 2016 followed by 'Violence against the person', 'Criminal damage'. The lowest recorded crimes are 'Drugs', 'Robbery' and 'Other Notifiable offenses'

Boroughs with the highest crime rates



Comparing five boroughs with the highest crime rate during the year 2016 it is evident that Westminster has the highest crimes recorded followed by Lambeth, Southwark, Newham and Tower Hamlets. Westminster has a significantly higher crime rate than the other 4 boroughs.

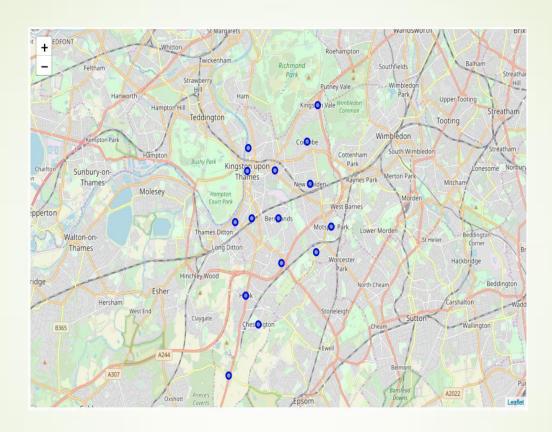
Boroughs with the lowest crime rates



Comparing five boroughs with the lowest crime rate during the year 2016, City of London has the lowest recorded crimes followed by Kingston upon Thames, Sutton, Richmond upon Thames and Merton.

- City of London has a significantly lower crime rate because it i is the 33rd principal division of Greater London but it is not a London borough. It has an area of 1.12 square miles and a population of 7000 as of 2013 which suggests that it is a small area.
- We will consider the next borough with the lowest crime rate as the safest borough in London which is Kingston upon Thames.

Neighborhoods in Kingston upon Thames



There are 15 neighborhoods in the royal borough of Kingston upon Thames, they are visualised on a map using folium on python.

Modelling

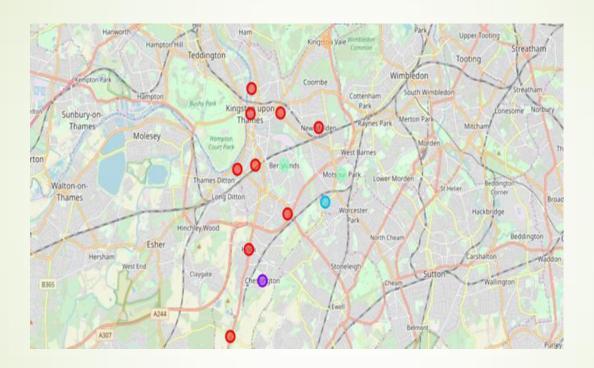
 Using the final data set containing the neighborhoods in Kingston upon Thames along with the latitude and longitude, we can find all the venues within a 500 meter radius of each neighborhood by connecting to the Foursquare API.

| | Neighborhood | Neighborh ood Latitude | Neighborhood Longitude | Venue | Venue Latitude | Venue Longitude | Venue Category |
|---|--------------|------------------------|------------------------|--------------------------------|----------------|-----------------|------------------|
| 0 | Benylands | 51.393781 | -0.284802 | Surbiton Racket & Fitness Club | 51.392676 | -0.290224 Gym | I Fitness Center |
| | Benylands | 51.393781 | -0.284802 | Alexandra Park | 51.394230 | -0.281206 | Park |
| 2 | Benylands | 51.393781 | -0.284802 | K2 Bus Stop | 51.392302 | -0.281534 | Bus Stop |
| 3 | Benylands | 51.393781 | -0.284802 | Cafe Rosa | 51.390175 | -0.282490 | Cafe |
| 4 | Canbury | 51.417499 | -0.305553 | The Boater's Inn | 51.418546 | -0.305915 | Pub |

- One hot encoding is done on the venues data. The Venues data is then grouped by the Neighborhood and the mean of the 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 a cluster size of 5 for this project that will cluster the 15 neighborhoods into 5 clusters. 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 of the five clusters. Visualizing the clustered neighborhoods on a map using the folium library.



Each cluster is color coded for the ease of presentation, we can see that majority of the neighborhood falls in the red cluster which is the first cluster. Three neighborhoods have their own cluster (Blue, Purple and Yellow), these are clusters two three and five. The green cluster consists of two neighborhoods which is the 4th cluster.

Cluster 1: Looking into the neighborhoods in the first cluster

| N | eighborhood | Borough | la titude l | _ongitude | Cluster Labels | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Mo st Common Venue | 5th Most Common Venue | 6th Most Common Venue | 7th Most Common Venue | 8th Most Common Venue |
|------|-------------------------|----------------------------|-------------|------------|-------------------|-----------------------------|-----------------------------|-----------------------------|------------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|
| | Canbury | Kingston upon Thames | 51.417499 - | 0.305553 | 0 | Pub | Cafe | Plaza | Fish & Chips Shop | Supermarket | Spa | Shop& Service | Park |
| 4 | Hook | Kingston upon Thames | 51.367898 - | 0.307145 | 0 | Bakery | Convenience Store | Indian Restaurant | Fish&Chlps Shop | Wine Shop | Food | Electroincs Store | Farmers Market |
| 5 k | Kingston upon Thames | Kingston upon Thames | 51.409627 - | 0.306262 | 0 Cc | offee Shop | Cafe | Burger Joint | Sush Restaurant | Pub | Record Shop | Cosmetics Shop | Market |
| 7 | Malden Rushett | Kingston upon Thames | 51.341052 | -0.319076 | 0 | Convenience Store | Pub | Garden Center | Restaurant | Fast Food Restaurant | Discount Store | Dry E Cleaner | Electronics Store |
| 9 | New Malden | Kingston upon Thames | 51.405335 | -0.263407 | 0 | Gastropub | Gym | Sushi Restaurant | Supermarket | Korean Restaurant | Indian Restaurant | Fish & Chip s Shop | Dry Cleaner |
| 10 | Norbiton | Kingston upon Thames | 51.409999 | -0.287396 | 0 | Indian Restaurant | Pub | Food | Italian Restaurant | Platform | Grocery Store | Farmers Mark et | Dry Cleaner |
| 12 / | Seething Wells | Kingston upon Thames | 51.392642 | .0.314366 | 0 | Indian Restaurant | CoffeeShop | Italian Restaurant | Pub | Cafe | | Fast Food Restaurant Ro | Chinese estaurant |
| 13 | Surbiton | Kingston upon Thames | 51.393756 | -0.303310 | 0 | CoffeeShop | Pub | Supermarket | Breakfast Spot | Grocery Store | Gastropub | French Restaurant | Trah Station |
| 14 | Tolworth | Kingston upon Thames | 51.378876 | . 0.282860 | 0 | Grocery Store | Pharmacy _F | Furniture / Home Stora | ain Station | Pizza Place | Discount Stora | Coffee Shop | Bus Stop |

The cluster one is the biggest cluster with 9 of the 15 neighborhoods in the borough kingston upon Thames. Upon closely examining these neighborhoods we can see that the most common venues in these neighborhoods are Restaurants, Pubs, Cafe, Supermarkets, and stores

Cluster 2: Looking into the neighborhoods in the second cluster.

| | Neighborhood E | Borough | Latitude Longitude | Cluster Labels | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | | | st 7th Most Common Venue | | 9th Most Common Venue |
|---|----------------|----------------------------|---------------------|-------------------|-----------------------------|--------------------------------|-----------------------------|-------------------------------|-----------|------------------|--------------------------------|---------------------------|-----------------------------|
| 2 | C hessington | Kingston upon Thames | 51.358336 -0.298622 | | Fast Food Restaurant | Wine Shop | Golf Course | German Restau ra nt | Gastropub | Garden Center | Furniture /Home Store | Fried Chicken Joint | French Restaurant |

The second cluster has one neighborhood which consists of Venues such as Restaurants, Golf courses, and wine shops.

Cluster 3: Looking into the neighborhoods in the third cluster.

| $/_{_}$ | Neighborhood E | Borough | Latitude | Longitude | Cluster Labe | 1st Most Common Is Venue | 2nd Most Common Venue | 3rd Most Common Venue | | | st 6th Most n Common Venue | | | 9th Most Common Venue |
|----------|----------------|----------------------------|-----------|-----------|-----------------|--------------------------------|--------------------------------|-----------------------------|----------|------------------|----------------------------------|---------------------------|---------------------|-----------------------------|
| 11 | Old Malden | Kingston upon Thames | 51.382484 | -0.25909 |) | Train 2 Station | Pub | Food G | astropub | Garden Center | Furniture / Home Store | Fried Chicken Joint | French estaurant | Deli/ Bod ega |

The third cluster has one neighborhood which consists of Venues such as Train stations, Restaurants, and Furniture shops.

Cluster 4: Looking into the neighborhoods in the fourth cluster.

| | Neighborhood E | Borough | Latitude | Longitude | Cluster Labels | 1st Most Common Venue | 2nd Most ommon Venue | | 4th Most Common Venue | 5th Most Common Venue | | | | 9th Most Common Venue |
|---|----------------|----------------------------|-----------|-----------|-------------------|-----------------------------|-------------------------------|------------|-----------------------------|-----------------------------|------------------------|--------------------------|--|-----------------------------|
| 0 | Benylands | Kingston upon Thames | 51.393781 | -0.284802 | 3 | Gy m / Fitness Center | Park | Cate | Bus Stop | Wine Shop | Fish& Chips Shop | Electronics Store | Farmers F Mar1 <et r<="" th=""><th></th></et> | |
| 8 | Motspur Par1< | Kingston upon Thames | | -0.248898 | 3 | Park | Gym | Restaurant | Socce Field | er Bus Stop | Wii Shop | ne FastFoo Restaurant | od Dr Cleaner | y Electronics Store |

The fourth cluster has two neighborhoods in it, these neighborhoods have common venues such as Parks, Gym/Fitness centers, Bus Stops, Restaurants, Electronics Stores and Soccer fields etc.

Cluster 5: Looking into the neighborhoods in the fourth cluster.

| $/$ _ | Neighborhood B | Borough | Latitude | Longitude ^C | uster _{Con} Labels | 1 st Most nmon Venue | 2nd Most Common Venue | | | | st 6th Most Common Venue | | 8th Most Common Venue | 9th Most Common Venue |
|-------|----------------|----------------------------|----------|------------------------|--------------------------------|----------------------------|--------------------------------|-----------------------|-----------------|------------------|--------------------------------|---------------------------|-----------------------------|-----------------------------|
| 6 | Kingston Vale | Kingston upon Thames | 51.43185 | -0.258138 | 4 | Grocery Store | Bar | Italian Restaurant | Soccer Field | Garden Center | Furniture / Home Store | Fried Chicken Joint | French D Restaurant | epartment Store |

The fifth cluster has one neighborhood which consists of Venues such as Grocery shops, Bars, Restaurants, Furniture shops, and Department stores.

5. Discussion

- The aim of this project is to help people who want to relocate to the safest borough in London, expats can chose 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 Clusters 3 and 4 have Train stations and Bus stops as the most common venues.
- If a person is looking for a neighborhood with stores and restaurants in a close proximity then the neighborhoods in the first cluster is suitable.
- For a family I feel that the neighborhoods in Cluster 4 are more suitable dues to the common venues in that cluster, these neighborhoods have common venues such as Parks, Gym/Fitness centers, Bus Stops, Restaurants, Electronics Stores and Soccer fields which is ideal for a family.
- The preference of venues may vary from person to person, they can select a neighborhood based on ones priorities.

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 safest borough of London. The future of this project includes taking other factors such as cost of living in the areas into consideration to shortlist the borough, such as filtering areas based on a predefined budget.