The Battle of the Neighbourhoods Coursera-Capstone Project

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Introduction

Background

In life, people are constantly relocating and moving on. It can be easy to assume that everyone moves homes for different reasons. The best decision is to take precautions and learn what could go wrong in advance so read on the top ten moving mistakes. Safety is a top concern when moving to a new area. If you don't feel safe in your own home, you're not going to be able to enjoy living there.

Business Problem

This Capstone project aims to select the safest borough in London based on the total crimes, explore the neighbourhoods of that borough to find the 10 most common venues in each neighborhood and finally cluster the neighbourhoods using k-mean clustering.

Target Audience

People who are looking to relocate to London and people are interested to identify the safest borough in London and explore its neighbourhoods and common venues around each neighbourhood.

Data Section

Two important factors are defined which may affect the problem:

- The total number of crimes committed in each of the borough in London during the last year.
- The most common venues in each of the neighborhood in the safest borough selected.

For this project we need the following data:

- <u>London Crime Data</u> The crime statistics dataset of London found on Kaggle has crimes in each Boroughs of London from 2008 to 2016.
- <u>List of London boroughs</u> A list of local authority districts within Greater London, including 32 London boroughs and the City of London.
- <u>List of districts in the Royal Borough of Kingston upon Thames</u> A dataset created from scratch using the list of neighbourhoods available on the site.

Methodology

Exploratory Data Analysis

| Total | Violence Against the Person | Theft and Handling | Robbery | Other Notifiable Offences | Drugs | Criminal Damage | Burglary | |
|--------------|-----------------------------|--------------------|-------------|---------------------------|-------------|-----------------|-------------|-------|
| 33.000000 | 33.000000 | 33.000000 | 33.000000 | 33.000000 | 33.000000 | 33.000000 | 33.000000 | count |
| 22306.696970 | 7041.848485 | 8913.121212 | 682.666667 | 479.060606 | 1179.212121 | 1941.545455 | 2069.242424 | mean |
| 8828.228749 | 2513.601551 | 4620.565054 | 441.425366 | 223.298698 | 586.406416 | 625.207070 | 737.448644 | std |
| 178.000000 | 25.000000 | 129.000000 | 4.000000 | 6.000000 | 10.000000 | 2.000000 | 2.000000 | min |
| 16903.000000 | 5936.000000 | 5919.000000 | 377.000000 | 378.000000 | 743.000000 | 1650.000000 | 1531.000000 | 25% |
| 22730.000000 | 7409.000000 | 8925.000000 | 599.000000 | 490.000000 | 1063.000000 | 1989.000000 | 2071.000000 | 50% |
| 27174.000000 | 8832.000000 | 10789.000000 | 936.000000 | 551.000000 | 1617.000000 | 2351.000000 | 2631.000000 | 75% |
| 48330.000000 | 10834.000000 | 27520.000000 | 1822.000000 | 1305.000000 | 2738.000000 | 3219.000000 | 3402.000000 | max |

Fig. 1 Statistical description of the London crimes

Methodology

Exploratory Data Analysis

| | 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 | 2069.242424 | 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 | 16903.000000 |
| 50% | 2071.000000 | 1989.000000 | 1063.000000 | 490.000000 | 599.000000 | 8925.000000 | 7409.000000 | 22730.000000 |
| 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 |

Fig. 1 Statistical description of the London crimes

The count for each of the major categories of crime returns the value 33 which 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

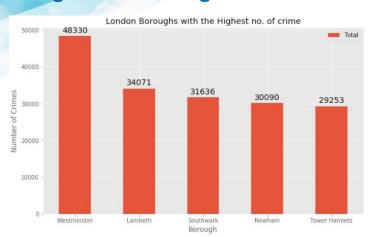


Fig. 2 Boroughs with the highest crime rates

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 Tower Hamlets. Westminster has a significantly higher crime rate than the other 4 boroughs (see Fig. 2).

Boroughs with the lowest crime rates

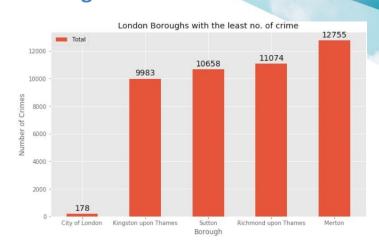


Fig. 3 Boroughs with the lowest crime rates

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 (see fig.3).

Neighbourhoods in Kingston upon Thames

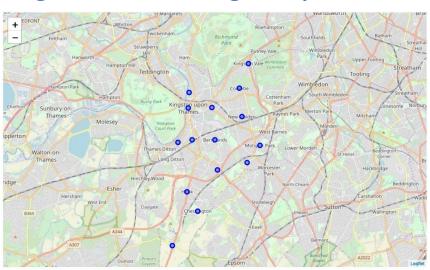


Fig 4 Neighbourhoods in Kingston upon Thames

There are 15 neighborhoods in the royal borough of Kingston upon Thames, they are visualised on a map using folium on python (see fig. 4).

Modelling

Using the final dataset containing the neighbourhoods in Kingston upon Thames along with the latitude and longitude, we can find all the venues within a 500 meter radius of each neighbourhood by connecting to the Foursquare API. This returns a json file containing all the venues in each neighbourhood which is converted to a pandas dataframe. This data frame contains all the venues along with their coordinates and category (see fig. 5).

| | Neighborhood | Neighborhood Latitude | Neighborhood Longitude | Venue | Venue Latitude | Venue Longitude | Venue Category |
|---|--------------|-----------------------|------------------------|--------------------------------|----------------|-----------------|----------------------|
| 0 | Berrylands | 51.393781 | -0.284802 | Surbiton Racket & Fitness Club | 51.392676 | -0.290224 | Gym / Fitness Center |
| 1 | Berrylands | 51.393781 | -0.284802 | Alexandra Park | 51.394230 | -0.281206 | Park |
| 2 | Berrylands | 51.393781 | -0.284802 | K2 Bus Stop | 51.392302 | -0.281534 | Bus Stop |
| 3 | Berrylands | 51.393781 | -0.284802 | Cafe Rosa | 51.390175 | -0.282490 | Café |
| 4 | Canbury | 51.417499 | -0.305553 | The Boater's Inn | 51.418546 | -0.305915 | Pub |

Fig 5 Venue details of each Neighborhood

Results

Visualising the clustered neighborhoods on a map using the folium library (see fig 4.11).

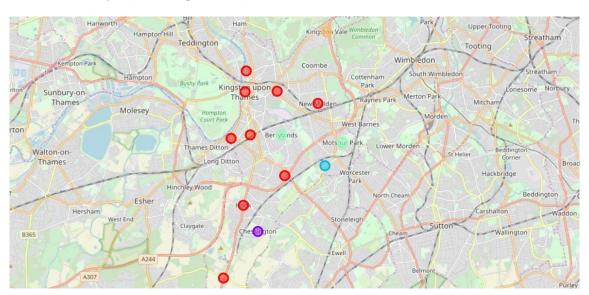


Fig 4.11 Clustered neighborhoods in the Borough of Kingston upon Thames

Cluster 1

| | Neighborhood | Borough | Latitude | Longitude | Cluster Labels | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue | 6th Most Common Venue | 7th Most Common Venue | 8th Most Common Venue |
|----|-------------------------|----------------------------|-----------|-----------|-------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|
| 1 | Canbury | Kingston upon Thames | 51.417499 | -0.305553 | 0 | Pub | Café | 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 & Chips Shop | Wine Shop | Food | Electronics Store | Farmers Market |
| 5 | Kingston upon Thames | Kingston upon Thames | 51.409627 | -0.306262 | 0 | Coffee Shop | Café | Burger Joint | Sushi 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 Cleaner | Electronics Store |
| 9 | New Malden | Kingston upon Thames | 51.405335 | -0.263407 | 0 | Gastropub | Gym | Sushi Restaurant | Supermarket | Korean Restaurant | Indian Restaurant | Fish & Chips Shop | Dry Cleaner |
| 10 | Norbiton | Kingston upon Thames | 51.409999 | -0.287396 | 0 | Indian Restaurant | Pub | Food | Italian Restaurant | Platform | Grocery Store | Farmers Market | Dry Cleaner |
| 12 | Seething Wells | Kingston upon Thames | 51.392642 | -0.314366 | 0 | Indian Restaurant | Coffee Shop | Italian Restaurant | Pub | Café | Wine Shop | Fast Food Restaurant | Chinese Restaurant |
| 13 | Surbiton | Kingston upon Thames | 51.393756 | -0.303310 | 0 | Coffee Shop | Pub | Supermarket | Breakfast Spot | Grocery Store | Gastropub | French Restaurant | Train Station |
| 14 | Tolworth | Kingston upon Thames | 51.378876 | -0.282860 | 0 | Grocery Store | Pharmacy | Furniture / Home Store | Train Station | Pizza Place | Discount Store | 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

| | Neighborhood | Borough | Latitude | Longitude | Cluster Labels | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue | 6th Most Common Venue | 7th Most Common Venue | 8th Most Common Venue | 9th Most Common Venue | |
|---|--------------|----------------------------|-----------|-----------|-------------------|-----------------------------|--------------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|------------------------------|-----------------------------|-----------------------------|-------|
| 2 | Chessington | Kingston upon Thames | 51.358336 | -0.298622 | 1 | Fast Food Restaurant | Wine Shop | Golf Course | German Restaurant | Gastropub | Garden Center | Furniture / Home Store | Fried Chicken Joint | French Restaurant | 10000 |

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

Cluster 3

| | Neighborhood | Borough | Latitude | Longitude | Cluster Labels | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | | 6th Most Common Venue | 7th Most Common Venue | 8th Most Common Venue | 9th Most Common Venue |
|----|--------------|----------------------------|-----------|-----------|-------------------|-----------------------------|--------------------------------|-----------------------------|-----------------------------|------------------|------------------------------|-----------------------------|-----------------------------|-----------------------------|
| 11 | Old Malden | Kingston upon Thames | 51.382484 | -0.25909 | 2 | Train Station | Pub | Food | Gastropub | Garden Center | Furniture / Home Store | Fried Chicken Joint | French Restaurant | Deli / Bodega |

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

Cluster 4

| | Neighborhood | Borough | Latitude | Longitude | Cluster Labels | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | | | 7th Most Common Venue | 8th Most Common Venue | 9th Most Common Venue |
|---|--------------|----------------------------|-----------|-----------|-------------------|-----------------------------|--------------------------------|-----------------------------|-----------------------------|--------------|-------------------------|-----------------------------|-----------------------------|-----------------------------|
| 0 | Berrylands | Kingston upon Thames | 51.393781 | -0.284802 | 3 | Gym / Fitness Center | Park | Café | Bus Stop | Wine Shop | Fish & Chips Shop | Electronics Store | Farmers Market | Fast Food Restaurant |
| 8 | Motspur Park | Kingston upon Thames | 51.390985 | -0.248898 | 3 | Park | Gym | Restaurant | Soccer Field | Bus Stop | Wine Shop | Fast Food Restaurant | Dry Cleaner | 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

| | Neighborhood | Borough | Latitude | Longitude | Cluster Labels | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | | 6th Most Common Venue | 7th 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 Restaurant | Department Store |

The fifth cluster has one neighborhood which consists of Venues such as Grocery shops, Bars, Restaurants, Furniture shops, and Department stores. We will look into the neighbourhoods in the fourth cluster.

Discussion

The aim of this project is to help people who want to relocate to the safest borough in London, expats can choose the neighbourhoods to which they want to relocate based on the most common venues in it. For example, if a person is looking for a neighbourhood 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 neighbourhood with stores and restaurants in a close proximity, then the neighbourhoods in the first cluster is suitable. For a family I feel that the neighbourhoods in Cluster 4 are more suitable dues to the common venues in that cluster, these neighbourhoods have common venues such as Parks, Gym/Fitness centres, Bus Stops, Restaurants, Electronics Stores and Soccer fields which is ideal for a family. The choices of neighbourhoods may vary from person to person.

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.