Applied Data Science Capstone Project The Battle of Neighborhoods

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1. Introduction

1.1 Introduction / Business Problem

Change is the only constant in today's modern age. People move from place to place to find comfort in life whether it is due to safety reasons, transportation or better food availability or job opportunities. But it's hard to relocate to different places and experience their conditions by actually living there. If it can be done virtually or with the help of data analysis then a lot of money and time will be saved and a relocator will definitely find the best place to live which fulfills all his/her needs.

So here in this project best and safest places to relocate in London are found out with analysis. The London crime data is obtained from Kaggle. Different boroughs in London are studied based on the total crimes, and neighbourhoods are explored for safest borough and analysed using k-mean clustering.

1.2 Interest

This report is targeted to people who are relocating to London considering safety as top priority. The crime data is taken into consideration to select safest neighbourhoods in London. Furthermore, for the safest borough, the neighbourhoods are clustered based on most common venues which are obtained from Foursquare API. Based on clusters a person can find best neighbourhood to relocate.

2 Data Loading

2.1 Data Acquisition

The data acquired from following data sources:

- The dataset consisting of the crime statistics of each borough in London obtained from Kaggle
- Co-ordinate of neighborhood obtained using Google Maps API geocoding
- Most common venues obtained by Foursquare API

The <u>London crime data</u> is obtained from Kaggle. The dataset contains the following columns:

- Isoa_code: code for Lower Super Output Area in Greater London.
- borough: Common name for London borough.
- major_category: High level categorization of crime
- minor_category: Low level categorization of crime within major category.
- value: monthly reported count of categorical crime in given borough
- year: Year of reported counts, 2008-2016
- month: Month of reported counts, 1-12

For the safest borough the coordinate of neighborhood obtained using Google Maps API geocoding. For that the neighborhood list is obtained from <u>Wikipedia</u>.

For the different neighborhoods the most common venues are obtained from Foursquare API.

2.2 Data Cleaning

The London crime data is loaded from Kaggle and is shown in fig 2.1

| | Isoa_code | borough | major_category | minor_category | value | year | month |
|---|-----------|------------|-----------------------------|-----------------------------|-------|------|-------|
| 0 | E01001116 | Croydon | Burglary | Burglary in Other Buildings | 0 | 2016 | 11 |
| 1 | E01001646 | Greenwich | Violence Against the Person | Other violence | 0 | 2016 | 11 |
| 2 | E01000677 | Bromley | Violence Against the Person | Other violence | 0 | 2015 | 5 |
| 3 | E01003774 | Redbridge | Burglary | Burglary in Other Buildings | 0 | 2016 | 3 |
| 4 | E01004563 | Wandsworth | Robbery | Personal Property | 0 | 2008 | 6 |

Fig 2.1 London crime data before preprocessing

The data of year 2016 is only selected. As shown in fig 2.2

| | Isoa_code | borough | major_category | minor_category | value | year | month |
|---|-----------|-----------|--------------------|----------------------------------|-------|------|-------|
| 0 | E01004177 | Sutton | Theft and Handling | Theft/Taking of Pedal Cycle | 1 | 2016 | 8 |
| 1 | E01000733 | Bromley | Criminal Damage | Criminal Damage To Motor Vehicle | 1 | 2016 | 4 |
| 2 | E01003989 | Southwark | Theft and Handling | Theft From Shops | 4 | 2016 | 8 |
| 3 | E01002276 | Havering | Burglary | Burglary in a Dwelling | 1 | 2016 | 8 |
| 4 | E01003674 | Redbridge | Drugs | Possession Of Drugs | 2 | 2016 | 11 |

Fig 2.2 London crime data 2016

Later for year 2016 only Borough, Major crime category and Number of crimes is selected. As shown in fig 2.3

| | Borough | Major_Category | No_of_Crimes |
|---|-----------|--------------------|--------------|
| 0 | Sutton | Theft and Handling | 1 |
| 1 | Bromley | Criminal Damage | 1 |
| 2 | Southwark | Theft and Handling | 4 |
| 3 | Havering | Burglary | 1 |
| 4 | Redbridge | Drugs | 2 |

Fig 2.3 London crime data after preprocessing

Further the data is pivoted across number of crimes of different crimes as shown in fig 2.4

| | 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 | 33482 |
| 1 | Barnet | 3402 | 2183 | 906 | 499 | 464 | 9731 | 7499 | 49368 |
| 2 | Bexley | 1123 | 1673 | 646 | 294 | 209 | 4392 | 4503 | 25680 |
| 3 | Brent | 2631 | 2280 | 2096 | 536 | 919 | 9026 | 9205 | 53386 |
| 4 | Bromley | 2214 | 2202 | 728 | 417 | 369 | 7584 | 6650 | 40328 |

Fig 2.4 London crime data further processing

Later for the safest borough in London the neighborhoods are obtained from Wikipedia and the coordinates are obtained from Google API geocoder. Final data frame is shown in fig 2.5

| | Neighborhood | Borough | Latitude | Longitude | |
|---|--------------|----------------------|----------|-----------|--|
| 0 | Berrylands | Kingston upon Thames | 51.3938 | -0.284802 | |
| 1 | Canbury | Kingston upon Thames | 51.4175 | -0.305553 | |
| 2 | Chessington | Kingston upon Thames | 51.3583 | -0.298622 | |
| 3 | Coombe | Kingston upon Thames | 51.4194 | -0.265398 | |
| 4 | Hook | Kingston upon Thames | 51.3679 | -0.307145 | |

Fig 2.5 Neighborhoods of the safest borough along with coordinates

For the neighborhoods best 10 most common venues are obtained using the Foursquare API,

| 10th Mos Common Venue | 9th Most Common Venue | 8th Most Common Venue | 7th Most Common Venue | 6th Most Common Venue | 5th Most Common Venue | 4th Most Common Venue | 3rd Most Common Venue | 2nd Most Common Venue | 1st Most Common Venue | Neighborhood | |
|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-------------------------|---|
| French Restaurant | Fish & Chips Shop | Fast Food Restaurant | Farmers Market | Food | Wine Shop | Bus Stop | Café | Park | Gym / Fitness Center | Berrylands | 0 |
| Supermarket | Spa | Park | Fish & Chips Shop | Indian Restaurant | Plaza | Café | Shop & Service | Hotel | Pub | Canbury | 1 |
| Food | Fast Food Restaurant | Farmers Market | Electronics Store | French Restaurant | Wine Shop | Indian Restaurant | Fish & Chips Shop | Supermarket | Bakery | Hook | 2 |
| Food | Fish & Chips Shop | Fast Food Restaurant | Farmers Market | Electronics Store | Wine Shop | Soccer Field | Bar | Grocery Store | Sandwich Place | Kingston Vale | 3 |
| Electronics | French Restaurant | Furniture / Home Store | Gift Shop | Asian Restaurant | Burger Joint | Café | Sushi Restaurant | Pub | Coffee Shop | Kingston upon Thames | 4 |

Fig 2.6 Neighborhoods with most common places

3. Methodology

3.1 Data Analysis

Statistical summary

The statistical summary is shown in fig 3.1

| | 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 | 44613.393939 |
| std | 737.448644 | 625.207070 | 586.406416 | 223.298698 | 441.425366 | 4620.565054 | 2513.601551 | 17656.457498 |
| min | 2.000000 | 2.000000 | 10.000000 | 6.000000 | 4.000000 | 129.000000 | 25.000000 | 356.000000 |
| 25% | 1531.000000 | 1650.000000 | 743.000000 | 378.000000 | 377.000000 | 5919.000000 | 5936.000000 | 33806.000000 |
| 50% | 2071.000000 | 1989.000000 | 1063.000000 | 490.000000 | 599.000000 | 8925.000000 | 7409.000000 | 45460.000000 |
| 75% | 2631.000000 | 2351.000000 | 1617.000000 | 551.000000 | 936.000000 | 10789.000000 | 8832.000000 | 54348.000000 |
| max | 3402.000000 | 3219.000000 | 2738.000000 | 1305.000000 | 1822.000000 | 27520.000000 | 10834.000000 | 96660.000000 |

Fig 3.1. Statistical description of the London crimes

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

Top five boroughs with the highest crime rate are Westminster, Lambeth, Southwark, Newham, Tower Hamlets. From fig 3.2

| | Borough | Burglary | Criminal Damage | Drugs | Other Notifiable Offences | Robbery | Theft and Handling | Violence Against the Person | Total |
|----|---------------|----------|-----------------|-------|---------------------------|---------|--------------------|-----------------------------|-------|
| 32 | Westminster | 3218 | 2179 | 2049 | 708 | 1822 | 27520 | 10834 | 96660 |
| 21 | Lambeth | 3087 | 2764 | 2738 | 635 | 1196 | 13155 | 10496 | 68142 |
| 27 | Southwark | 2946 | 2621 | 1838 | 494 | 1317 | 12946 | 9474 | 63272 |
| 24 | Newham | 2115 | 2496 | 1684 | 713 | 1472 | 11964 | 9646 | 60180 |
| 29 | Tower Hamlets | 2794 | 2357 | 1629 | 678 | 1234 | 10953 | 9608 | 58506 |

Fig 3.2 Boroughs with the highest crime rates

Boroughs with the lowest crime rates

Top low crime rate borough are Kingston upon Thames, Sutton, Richmond upon Thames and Merton from fig 3.3

| | Borough | Burglary | Criminal Damage | Drugs | Other Notifiable Offences | Robbery | Theft and Handling | Violence Against the Person | Total |
|----|----------------------|----------|-----------------|-------|---------------------------|---------|--------------------|-----------------------------|-------|
| 25 | Redbridge | 1997 | 1650 | 1017 | 381 | 599 | 7447 | 6411 | 39004 |
| 15 | Havering | 1826 | 1804 | 718 | 389 | 311 | 5919 | 5936 | 33806 |
| 0 | Barking and Dagenham | 1287 | 1949 | 919 | 378 | 534 | 5607 | 6067 | 33482 |
| 14 | Harrow | 1994 | 1212 | 473 | 267 | 377 | 4537 | 4293 | 26306 |
| 2 | Bexley | 1123 | 1673 | 646 | 294 | 209 | 4392 | 4503 | 25680 |
| 23 | Merton | 1419 | 1418 | 466 | 249 | 283 | 4894 | 4026 | 25510 |
| 26 | Richmond upon Thames | 1359 | 1148 | 320 | 217 | 106 | 4769 | 3155 | 22148 |
| 28 | Sutton | 1233 | 1316 | 461 | 253 | 165 | 3516 | 3714 | 21316 |
| 20 | Kingston upon Thames | 879 | 1054 | 743 | 189 | 121 | 3803 | 3194 | 19966 |

Fig 3.3 Boroughs with the lowest crime rates

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 in fig 3.4

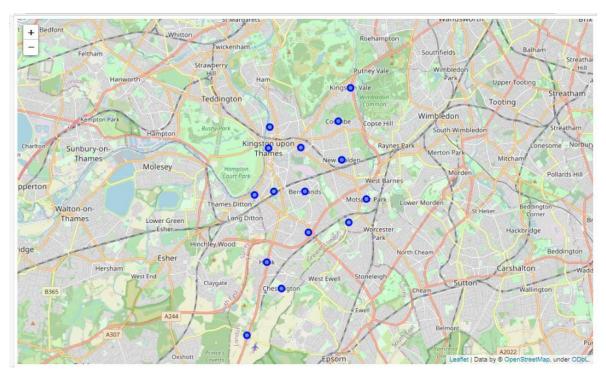


Fig 3.4 Neighborhoods in Kingston upon Thames

3.2 Modelling

For neighborhoods in Kingston upon Thames all the venues within a 500 m radius of are obtained from Foursquare API. This returns a json file containing all the venues in each neighborhood which is converted to a pandas data frame. As shown in fig 3.5

| | 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 | Canbury Gardens | 51.417409 | -0.305300 | Park |

Fig 3.5 Venue details of each Neighborhood

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.

The neighborhoods are clustered using K - means clustering and visualized data using color coding the clusters.

4. Results

After K-means clustering the neighborhoods are classified into 5 different clusters

| | | 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 Mos Commo Venu |
|---|----|-------------------|----------------------------|----------|-----------|-------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|--------------------------|
| | 6 | Kingston Vale | Kingston upon Thames | 51.4318 | -0.258138 | 0 | Sandwich Place | Grocery Store | Bar | Soccer Field | Wine Shop | Electronics Store | Farmers Market | Fast Food Restaurant | Fish Chip Sho |
| | 7 | Malden Rushett | Kingston upon Thames | 51.3411 | -0.319076 | 0 | Grocery Store | Pub | Garden Center | Restaurant | Wine Shop | Fish & Chips Shop | Department Store | Electronics Store | Farmer Marke |
| | 8 | Motspur Park | Kingston upon Thames | 51.391 | -0.248898 | 0 | Gym | Park | Restaurant | Soccer Field | Fish & Chips Shop | Department Store | Electronics Store | Farmers Market | |
| 1 | 14 | Tolworth | Kingston upon Thames | 51.3789 | -0.28286 | 0 | Grocery Store | Pharmacy | Bowling Alley | Coffee Shop | Italian Restaurant | Pizza Place | Hotel | Café | Bus Sto |

Fig 4.1 Cluster 1

The 1st cluster has 4 of 15 neighborhoods in the borough Kingston upon Thames. The most common venues are Gym, sandwich place and grocery stores, pharmacy, park. .

| | 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 | |
|----|--------------|----------------------------|----------|-----------|-------------------|-------------------------------|--------------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-------------------------|
| 11 | Old Malden | Kingston upon Thames | 51.3825 | -0.25909 | 1 | Construction & Landscaping | Train Station | Food | Deli / Bodega | Department Store | Electronics Store | Farmers Market | Fast Food Restaurant | Fish & Chips Shop |
| 4 | | | | | | | | | | | | | | + |

Fig 4.2 Cluster 2

The second cluster has one neighborhood with Construction & Landscaping as most common venue.

| | Neighborhood | Borough | Latitude | Longitude | Cluster Labels | 1st Most Common Venue | | 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue | 6th Most Common Venue | 7th Most Common Venue | | 9th Most Common Venue | (|
|---|--------------|----------------------------|----------|-----------|-------------------|-----------------------------|-------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-------------------|-----------------------------|---|
| | 4 Hook | Kingston upon Thames | 51.3679 | -0.307145 | 2 | Bakery | Supermarket | Fish & Chips Shop | Indian Restaurant | Wine Shop | French Restaurant | | Farmers Market | Fast Food Restaurant | |
| 4 | | | | | | | | | | | | | | + | |

Fig 4.3 Cluster 3

The third cluster has one neighborhood with Bakery as most common venue.

| | Neighborhood | Borough | Latitude | Longitude | Cluster Labels | 1st Most Common Venue | Most Common Venue | | | 5th Most Common Venue | | | | 9th Most Common Venue | 10th M Comn Ve |
|---|--------------|----------------------------|----------|-----------|-------------------|-----------------------------|-------------------------|------|----------|-----------------------------|------|-------------------|-------------------------|-----------------------------|----------------------|
| 0 | Berrylands | Kingston upon Thames | 51.3938 | -0.284802 | 3 | Gym / Fitness Center | Park | Café | Bus Stop | Wine Shop | Food | Farmers Market | Fast Food Restaurant | Fish & Chips Shop | Fre Restau |
| 4 | | | | | | | | | | | | | | | - |

Fig 4.4 Cluster 4

The fourth cluster has one neighborhood with Gym / Fitness Center as most common venue.

| 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 | |
|-------------------------|----------------------------|----------|-----------|-------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|------------------------------|-----------------------------|----|
| Canbury | Kingston upon Thames | 51.4175 | -0.305553 | 4 | Pub | Hotel | Shop & Service | Café | Plaza | Indian Restaurant | Fish & Chips Shop | Park | Spa | St |
| Kingston upon Thames | Kingston upon Thames | 51.4096 | -0.306262 | 4 | Coffee Shop | Pub | Sushi Restaurant | Café | Burger Joint | Asian Restaurant | Gift Shop | Furniture / Home Store | French Restaurant | |
| New Malden | Kingston upon Thames | 51.4053 | -0.263407 | 4 | Gym | Gastropub | Sushi Restaurant | Supermarket | Bar | Indian Restaurant | Korean Restaurant | Food | Electronics Store | |
| Norbiton | Kingston upon Thames | 51.41 | -0.287396 | 4 | Pub | Italian Restaurant | Indian Restaurant | Food | Platform | Wine Shop | Coffee Shop | Hotel | Hardware Store | |
| Seething Wells | Kingston upon Thames | 51.3926 | -0.314366 | 4 | Indian Restaurant | Coffee Shop | Pub | Café | Gym | Restaurant | Park | Pet Café | Fast Food Restaurant | - |
| Surbiton | Kingston upon Thames | 51.3938 | -0.30331 | 4 | Coffee Shop | Pub | Grocery Store | Italian Restaurant | Pharmacy | French Restaurant | Train Station | Gym / Fitness Center | Hotel | |
| 4 | | | | | | | | | | | | | | - |

Fig 4.5 Cluster 5

The fifth cluster is biggest cluster with 6/15 neighborhood which consists of Venues such as Pub, Coffee shop, gym, indian restaurant

Visualising the clustered neighborhoods on a map using the folium library fig 4.6.

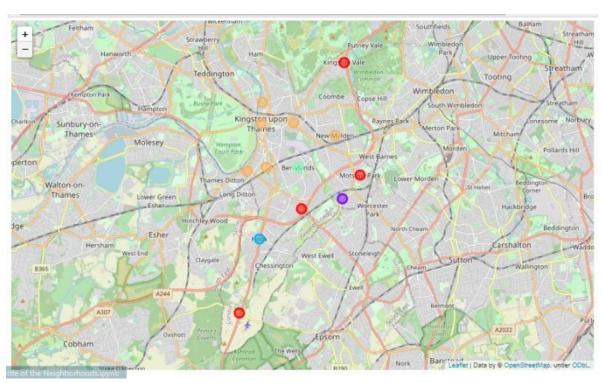


Fig 4.6 Clustered neighborhoods in the Borough of Kingston upon Thame

5. Discussion

The aim of this project is to find safest borough in London for people who are relocating to London. For the safest borough the neighbourhood need to be analyzed so that person will be able to enjoy most common venues and could enjoy life along with safety. Most important venues include pharmacy, gym, restaurants, ease of transportation and grocery stores.

From the analysis of neighborhoods we can see that 1st cluster includes venues like Gym, sandwich place and grocery stores, pharmacy, park. 2nd cluster includes Construction & Landscaping, Train Station, Food, Deli / Bodega, Department Store, Electronics Store, Farmers Market. 3rd cluster includes Bakery, Supermarket, Fish & Chips Shop, Indian Restaurant, Wine Shop, French Restaurant, Electronics Store. 4th cluster includes Gym / Fitness Center, Park, Café, Bus Stop, Wine Shop, Food, Farmers Market. 5th cluster includes Pub, Coffee shop, gym, Indian restaurant, pharmacy, mostly restaurants and eatery.

For a family 1st cluster is more suitable dues to the common venues in that cluster includes most of the essential services. For a foodie person I think 5th cluster is more suitable.

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6. Conclusion

Relocation is most difficult and hectic process and it involves moving to new unknown location where we don't about food, transportation services and neighboring area and safety. so due to new techniques in data science we can analyze different neighborhoods as if we are virtually present there and we can choose best area to live which is safe for our family and also fulfills all necessary needs. Here in this project we analyzed different borough of London based on crime rates and shortlisted some neighborhood of safe borough and finally categorized different clusters based on most common venues. this can be helpful for person relocating to London and choosing best neighborhood to live.