# <u>Capstone project – The battle of</u> <u>neighborhoods</u>

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**Presentation** 

## 1. Introduction

- Background: 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.
- Problem: This project aims to select the safest borough in London based on the total crimes, explore the neighborhoods of that borough to find the 10 most common venues in each neighborhood and finally cluster the neighborhoods using k-mean clustering.
- Interest: Expats who are considering to relocate to London will be interested to identify the safest borough in London and explore its neighborhoods and common venues around each neighborhood.

## 2. Data Acquisition and Cleaning

**Data Acquisition**: The data acquired for this project is a combination of data from three sources:

- The first data source of the project uses a London crime data that shows the crime per borough in London.
- The second source of data is scraped from a wikipedia page that contains the list of London boroughs. This page contains additional information about the boroughs.
- The third data source is the list of Neighborhoods in the Royal Borough of Kingston upon Thames as found on the wikipedia page.

Data Cleaning: The data cleaning process for each of the three sources of data are done separately.

selected. The major categories of crime are pivoted to get the total crimes per the boroughs for each major category.

Using this library we can extract the stata in the tabular format as shown in the website.

3. Methodology

**Data Cleaning:** The data cleaning process for each of the three sources of data are 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 crimes per the boroughs for each major category.
- 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.
- The two data sets are merged on the Borough names to form a new data set. The purpose of this data set is to visualize the crime rates in each borough and identify the borough with the least crimes recorded during the year 2016.
- After visualizing the crime in each borough we can find the borough with the lowest crime rate. The third data set is created, with the names of the neighborhoods and the name of the borough with the latitude and longitude obtained using Google Maps API geocoding.
- The new data set is used to generate the 10 most common venues for each neighborhood using the Foursquare API, finally using k means clustering algorithm to cluster similar neighborhoods together.

## 3. Methodology

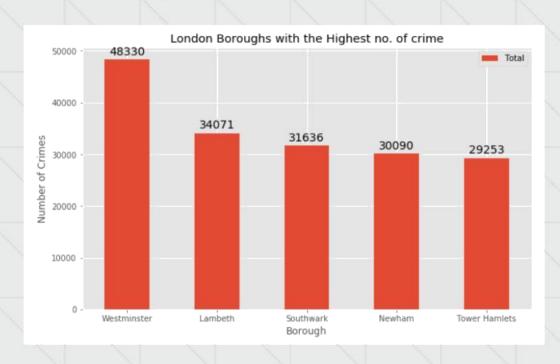
#### **Exploratory Data Analysis**

#### Statistical summary of crimes

| 33.000000<br>2069.242424 | Criminal Damage<br>33.000000  | Drugs<br>33.000000  | Other Notifiable Offences   | Robbery   | Theft and Handling   | Violence Against the Person   | Total   |
|--------------------------|---|---|---|---|--|---|---|
|                          | 33.000000   | 33.000000   | 22 000000   |   |  |   |   |
| 2069.242424              |   |   | 33.000000   | 33.000000   | 33.000000  | 33.000000   | 33.000000   |
|                          | 1941.545455   | 1179.212121   | 479.060606  | 682.666667  | 8913.121212  | 7041.848485   | 22306.696970  |
| 737.448644               | 625.207070  | 586.406416  | 223.298698  | 441.425366  | 4620.565054  | 2513.601551   | 8828.228749   |
| 2.000000                 | 2.000000  | 10.000000   | 6.000000  | 4.000000  | 129.000000   | 25.000000   | 178.000000  |
| 1531.000000              | 1650.000000   | 743.000000  | 378.000000  | 377.000000  | 5919.000000  | 5936.000000   | 16903.000000  |
| 2071.000000              | 1989.000000   | 1063.000000   | 490.000000  | 599.000000  | 8925.000000  | 7409.000000   | 22730.000000  |
| 2631.000000              | 2351.000000   | 1617.000000   | 551.000000  | 936.000000  | 10789.000000   | 8832.000000   | 27174.000000  |
| 3402.000000              | 3219.000000   | 2738.000000   | 1305.000000   | 1822.000000   | 27520.000000   | 10834.000000  | 48330.000000  |
|                          | 737.448644<br>2.000000<br>1531.000000<br>2071.000000<br>2631.000000 | 737.448644 625.207070<br>2.000000 2.000000<br>1531.000000 1650.000000<br>2071.000000 1989.000000<br>2631.000000 2351.000000 | 737.448644 625.207070 586.406416<br>2.000000 2.000000 10.000000<br>1531.000000 1650.000000 743.000000<br>2071.000000 1989.000000 1063.000000<br>2631.000000 2351.000000 1617.000000 | 737.448644         625.207070         586.406416         223.298698           2.000000         2.000000         10.000000         6.000000           1531.000000         1650.00000         743.000000         378.000000           2071.000000         1989.000000         1063.000000         490.000000           2631.000000         2351.000000         1617.000000         551.000000 | 737.448644         625.207070         586.406416         223.298698         441.425366           2.000000         2.000000         10.000000         6.000000         4.000000           1531.000000         1650.00000         743.000000         378.000000         377.000000           2071.000000         1989.00000         1063.000000         490.00000         599.000000           2631.000000         2351.000000         1617.000000         551.000000         936.000000 | 737.448644         625.207070         586.406416         223.298698         441.425366         4620.565054           2.000000         2.000000         10.000000         6.000000         4.000000         129.000000           1531.000000         1650.000000         743.000000         378.000000         377.000000         5919.000000           2071.000000         1989.000000         1663.000000         490.000000         599.000000         8925.000000           2631.000000         2351.000000         1617.000000         551.000000         936.000000         10789.000000 | 737.448644         625.207070         586.406416         223.298698         441.425366         4620.565054         2513.601551           2.000000         2.000000         10.000000         6.000000         4.000000         129.000000         25.000000           1531.000000         1650.00000         743.000000         378.000000         377.000000         5919.000000         5936.000000           2071.000000         1989.00000         1063.000000         490.000000         599.000000         8925.000000         7409.000000           2631.000000         2351.000000         1617.000000         551.000000         936.000000         10789.000000         8832.000000 |

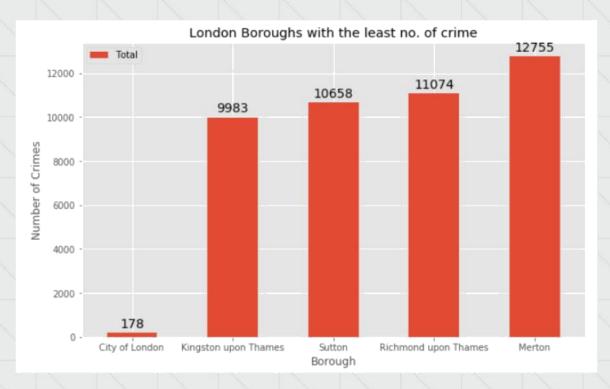
The count for each of the major categories of crime returns the value 33 which is the number of London boroughs. 'Theft and Handling' is the highest reported crime during the year 2016 followed by 'Violence against the person', 'Criminal damage'. The lowest recorded crimes are 'Drugs', 'Robbery' and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

#### Boroughs with the highest crime rates



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.

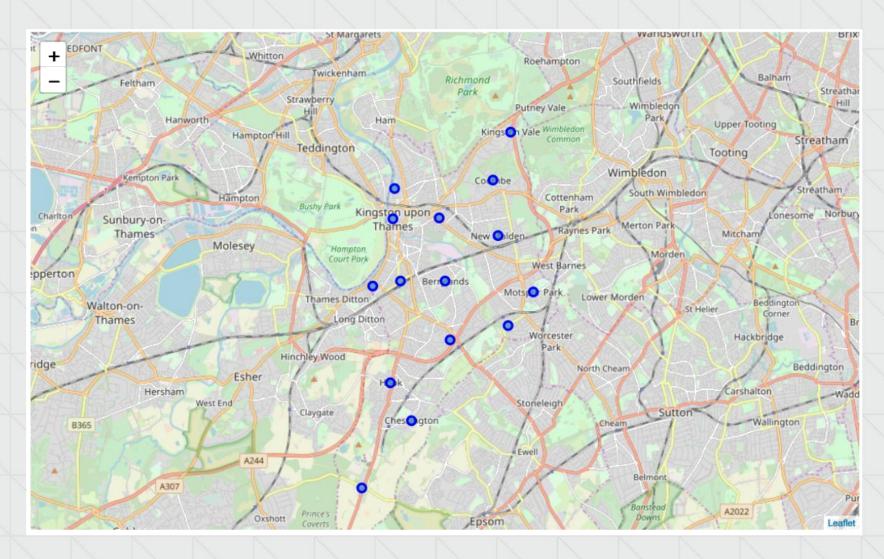
#### Boroughs with the lowest crime rates



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

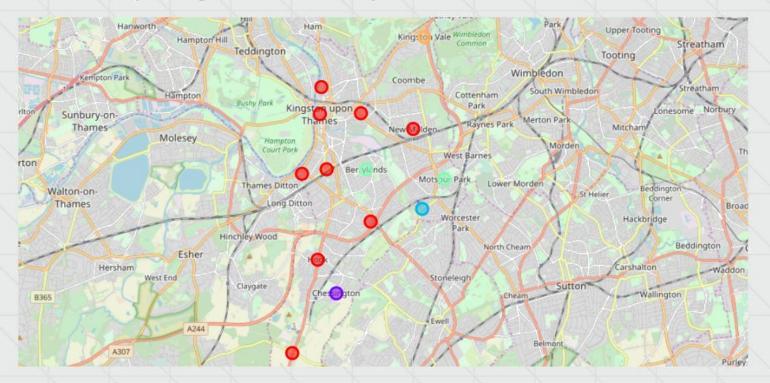
 Using the final data set containing the neighborhoods in Kingston upon Thames along with the latitude and longitude, we can find all the venues within a 500 meter radius of each neighborhood by connecting to the Foursquare API.

|   | Neighborhood | 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                  |

- 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

|    | Neighborhood            | Borough                    | Latitude  | Longitude | Cluster<br>Labels | 1st Most<br>Common<br>Venue | 2nd Most<br>Common<br>Venue | 3rd Most<br>Common<br>Venue | 4th Most<br>Common<br>Venue | 5th Most<br>Common<br>Venue | 6th Most<br>Common<br>Venue | 7th Most<br>Common<br>Venue | 8th Most<br>Common<br>Venue |
|----|-------------------------|----------------------------|-----------|-----------|-------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|
| 1  | Canbury                 | Kingston<br>upon<br>Thames | 51.417499 | -0.305553 | 0                 | Pub                         | Café                        | Plaza                       | Fish & Chips<br>Shop        | Supermarket                 | Spa                         | Shop &<br>Service           | Park                        |
| 4  | Hook                    | Kingston<br>upon<br>Thames | 51.367898 | -0.307145 | 0                 | Bakery                      | Convenience<br>Store        | Indian<br>Restaurant        | Fish & Chips<br>Shop        | Wine Shop                   | Food                        | Electronics<br>Store        | Farmers<br>Market           |
| 5  | Kingston upon<br>Thames | Kingston<br>upon<br>Thames | 51.409627 | -0.306262 | 0                 | Coffee Shop                 | Café                        | Burger Joint                | Sushi<br>Restaurant         | Pub                         | Record<br>Shop              | Cosmetics<br>Shop           | Market                      |
| 7  | Malden<br>Rushett       | Kingston<br>upon<br>Thames | 51.341052 | -0.319076 | 0                 | Convenience<br>Store        | Pub                         | Garden<br>Center            | Restaurant                  | Fast Food<br>Restaurant     | Discount<br>Store           | Dry<br>Cleaner              | Electronics<br>Store        |
| 9  | New Malden              | Kingston<br>upon<br>Thames | 51.405335 | -0.263407 | 0                 | Gastropub                   | Gym                         | Sushi<br>Restaurant         | Supermarket                 | Korean<br>Restaurant        | Indian<br>Restaurant        | Fish &<br>Chips<br>Shop     | Dry<br>Cleaner              |
| 10 | Norbiton                | Kingston<br>upon<br>Thames | 51.409999 | -0.287396 | 0                 | Indian<br>Restaurant        | Pub                         | Food                        | Italian<br>Restaurant       | Platform                    | Grocery<br>Store            | Farmers<br>Market           | Dry<br>Cleaner              |
| 12 | Seething Wells          | Kingston<br>upon<br>Thames | 51.392642 | -0.314366 | 0                 | Indian<br>Restaurant        | Coffee Shop                 | Italian<br>Restaurant       | Pub                         | Café                        | Wine<br>Shop                | Fast Food<br>Restaurant     | Chinese<br>Restaurant       |
| 13 | Surbiton                | Kingston<br>upon<br>Thames | 51.393756 | -0.303310 | 0                 | Coffee Shop                 | Pub                         | Supermarket                 | Breakfast<br>Spot           | Grocery<br>Store            | Gastropub                   | French<br>Restaurant        | Train<br>Station            |
| 14 | Tolworth                | Kingston<br>upon<br>Thames | 51.378876 | -0.282860 | 0                 | Grocery<br>Store            | Pharmacy                    | Furniture /<br>Home Store   | Train Station               | Pizza Place                 | Discount<br>Store           | Coffee<br>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.

| 1 | Neighborhood | Borough                    | Latitude  | Longitude | Cluster<br>Labels | 1st Most<br>Common<br>Venue | 2nd<br>Most<br>Common<br>Venue | 3rd Most<br>Common<br>Venue | 4th Most<br>Common<br>Venue |           | 6th Most<br>Common<br>Venue | 7th Most<br>Common<br>Venue  | 8th Most<br>Common<br>Venue | 9th Most<br>Common<br>Venue |
|---|--------------|----------------------------|-----------|-----------|-------------------|-----------------------------|--------------------------------|-----------------------------|-----------------------------|-----------|-----------------------------|------------------------------|-----------------------------|-----------------------------|
| 2 | Chessington  | Kingston<br>upon<br>Thames | 51.358336 | -0.298622 | ì                 | Fast Food<br>Restaurant     | Wine<br>Shop                   | Golf<br>Course              | German<br>Restaurant        | Gastropub | Garden<br>Center            | Furniture<br>/ Home<br>Store | Fried<br>Chicken<br>Joint   | French<br>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 | Borough                    | Latitude  | Longitude | Cluster<br>Labels | 1st Most<br>Common<br>Venue | 2nd<br>Most<br>Common<br>Venue | 3rd Most<br>Common<br>Venue |           | 5th Most<br>Common<br>Venue |                              |                           | 8th Most<br>Common<br>Venue |                  |
|----|--------------|----------------------------|-----------|-----------|-------------------|-----------------------------|--------------------------------|-----------------------------|-----------|-----------------------------|------------------------------|---------------------------|-----------------------------|------------------|
| 11 | Old Malden   | Kingston<br>upon<br>Thames | 51.382484 | -0.25909  | 2                 | Train<br>Station            | Pub                            | Food                        | Gastropub | Garden<br>Center            | Furniture<br>/ Home<br>Store | Fried<br>Chicken<br>Joint | French<br>Restaurant        | Deli /<br>Bodega |

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 | Borough                    | Latitude  | Longitude | Cluster<br>Labels | 1st Most<br>Common<br>Venue | 2nd<br>Most<br>Common<br>Venue | 3rd Most<br>Common<br>Venue |                 |              |                         | 7th Most<br>Common<br>Venue | 8th Most<br>Common<br>Venue | 9th Most<br>Common<br>Venue |
|---|--------------|----------------------------|-----------|-----------|-------------------|-----------------------------|--------------------------------|-----------------------------|-----------------|--------------|-------------------------|-----------------------------|-----------------------------|-----------------------------|
| 0 | Berrylands   | Kingston<br>upon<br>Thames | 51.393781 | -0.284802 | 3                 | Gym /<br>Fitness<br>Center  | Park                           | Café                        | Bus Stop        | Wine<br>Shop | Fish &<br>Chips<br>Shop | Electronics<br>Store        | Farmers<br>Market           | Fast Food<br>Restaurant     |
| 8 | Motspur Park | Kingston<br>upon<br>Thames | 51.390985 | -0.248898 | 3                 | Park                        | Gym                            | Restaurant                  | Soccer<br>Field | Bus Stop     | Wine<br>Shop            | Fast Food<br>Restaurant     | Dry<br>Cleaner              | Electronics<br>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  | Borough                    | Latitude | Longitude | Cluster<br>Labels | 1st Most<br>Common<br>Venue | 2nd<br>Most<br>Common<br>Venue | 3rd Most<br>Common<br>Venue |                 | 5th Most<br>Common<br>Venue |                              | 7th Most<br>Common<br>Venue | 8th Most<br>Common<br>Venue | 9th Most<br>Common<br>Venue |
|---|---------------|----------------------------|----------|-----------|-------------------|-----------------------------|--------------------------------|-----------------------------|-----------------|-----------------------------|------------------------------|-----------------------------|-----------------------------|-----------------------------|
| 6 | Kingston Vale | Kingston<br>upon<br>Thames | 51.43185 | -0.258138 | 4                 | Grocery<br>Store            | Bar                            | Italian<br>Restaurant       | Soccer<br>Field | Garden<br>Center            | Furniture<br>/ Home<br>Store | Fried<br>Chicken<br>Joint   | French<br>Restaurant        | Department<br>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.