# The Battle of Neighbourhoods

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### 1. Problem

- In London, there are numerous boroughs. The city is famous for its many touristic spots and is one of the most visited cities
  on the planet. In addition to tourists, many students come to London in order to study at university. As well as this, many
  people come for short term stays for e.g. internships, as well as for graduate programs.
- Finding a good place to stay is very important, somewhere that has local access to shops and supermarkets, as well as restaurants, bars and a number of other facilities.
- HS is moving to London in order to study at University College London. In addition to university life, HS also wants to get a feel
  of the city's culture, food etc. He's looking to live in a neighbourhood that is culturally vibrant but not too touristy, as he feels
  that he can visit tourist sites in his spare time, therefore the proximity of tourist attractions is not something that matters to
  him
- From the perspective of students, many factors are involved when searching for the best accommodation, which includes distance and rent. However, this project will focus on the general atmosphere of different areas of London as well as safety.
- The research carried out is expected to be of benefit to international students looking to live in London, because despite the opportunity to explore the local culture, they would very likely want to feel at home at the same time.

### 2 Data

• In the project, I will be using the following datasets to produce an outcome - Borough Level Crime, List of London boroughs and Foursquare API. After using legitimate resources to acquire them, they will be cleansed into more useful forms so that they can be analysed further.

#### (1) Borough Level Orime

- Description: London crime records classified by borough and crime type over the last 24 months.
- Size: 1569 observations x 27 columns
- Source: London Datastore, <a href="https://data.london.gov.uk/dataset/recorded\_crime\_summary">https://data.london.gov.uk/dataset/recorded\_crime\_summary</a>

:		MajorText	MinorText	LookUp_BoroughName	201807	201808	201809	201810	201811	201812	201901	 201909	201910	201911
	0	Arson and Criminal Damage	Arson	Barking and Dagenham	6	5	3	8	5	1	5	 6	9	8
	1	Arson and Criminal Damage	Criminal Damage	Barking and Dagenham	127	101	107	132	105	88	97	 109	109	97
	2	Burglary	Burglary - Business and Community	Barking and Dagenham	30	18	33	32	39	33	45	 37	30	30
	3	Burglary	Burglary - Residential	Barking and Dagenham	94	84	99	94	106	164	114	 80	97	114
	4	Drug Offences	Drug Trafficking	Barking and Dagenham	8	7	10	7	7	4	5	 7	8	13

### 2 Data

### (2) List of London Boroughs

- Description: Information 32 London boroughs, of which the variables 'Population' and 'Co-ordinates' will be used for the purpose of the project.
- Source: "List of London Boroughs", Wikipedia

• URL: <a href="https://en.wikipedia.org/wiki/List\_of\_London\_boroughs">https://en.wikipedia.org/wiki/List\_of\_London\_boroughs</a>

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Borough +	Inner +	Status \$	Local authority +	Political control \$	Headquarters +	Area (sq ¢ mi)	Population (2013 est) <sup>[1]</sup>	Co- ordinates	Nr. in			
Barking and Dagenham [note 1]			Barking and Dagenham London Borough Council	Labour	Town Hall, 1 Town Square	13.93	194,352	© 51.5607°N 0.1557°E	25			
Barnet	arnet		Barnet London Borough Council	Conservative	Barnet House, 2 Bristol Avenue, Colindale	33.49	369,088	\$1.6252°N 0.1517°W	31			
Bexley			Bexley London Borough Council	Conservative	Civic Offices, 2 Watling Street	23.38	236,687	\$1.4549°N 0.1505°E	23			
Brent			Brent London Borough Council	Labour	Brent Civic Centre, Engineers Way	16.70	317,264	\$1.5588°N 0.2817°W	12			
Bromley			Bromley London Borough Council	Conservative	Civic Centre, Stockwell Close	57.97	317,899	\$1.4039°N 0.0198°E	20			
Camden	✓		Camden London Borough Council	Labour	Camden Town Hall, Judd Street	8.40	229,719	\$1.5290°N 0.1255°W	11			
Croydon			Croydon London Borough Council	Labour	Bernard Weatherill House, Mint Walk	33.41	372,752	\$1.3714°N 0.0977°W	19			
Ealing			Ealing London Borough Council	Labour	Perceval House, 14-16 Uxbridge Road	21.44	342,494	\$1.5130°N 0.3089°W	13			
Enfield			Enfield London Borough Council	Labour	Civic Centre, Silver Street	31.74	320,524	\$1.6538°N 0.0799°W	30			
Greenwich [note 2]	[note 3]	Royal	Greenwich London Borough Council	Labour	Woolwich Town Hall, Wellington Street	18.28	264,008	\$1.4892°N 0.0648°E	22			

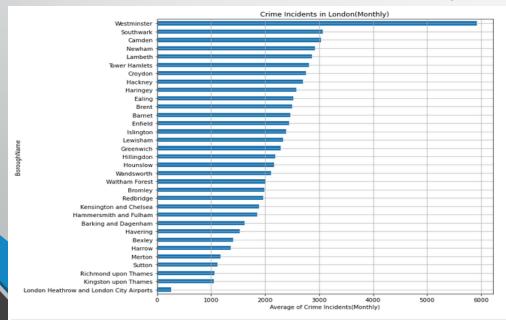
### 2 Data

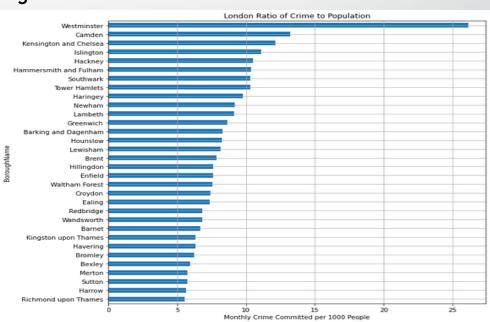
### (3) Foursquare API

- Description: List of the 50 most popular places in each borough. However, it's possible that some boroughs may have less than 50 registered on Foursquare
- Source: Foursquare API
- URL: <a href="https://api.foursquare.com">https://api.foursquare.com</a>

	BoroughName	Borough Latitude	Borough Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category
0	Barking and Dagenham	51.5607	0.1557	Central Park	51.559560	0.161981	Park
1	Barking and Dagenham	51.5607	0.1557	Crowlands Heath Golf Course	51.562457	0.155818	Golf Course
2	Barking and Dagenham	51.5607	0.1557	Robert Clack Leisure Centre	51.560808	0.152704	Martial Arts Dojo
3	Barking and Dagenham	51.5607	0.1557	Bea contree Heath Leisure Centre	51.560997	0.148932	Gym / Fitness Center
4	Barking and Dagenham	51.5607	0.1557	Becontree Heath Bus Station	51.561065	0.150998	Bus Station

- (1) Exploratory analysis Which borough has higher levels of crime?
- To have a better look at the datasets, they were each cleansed into more useful resources.
- By combining the Borough Level Crime data and the List of London Boroughs, the number of crimes per 1000 people was calculated.
- While visualising the boroughs according to total level of crimes and by the ratio of crime per 1000 people, it can be seen that both Westminster and Camden still remain 2 of the most dangerous boroughs to live.



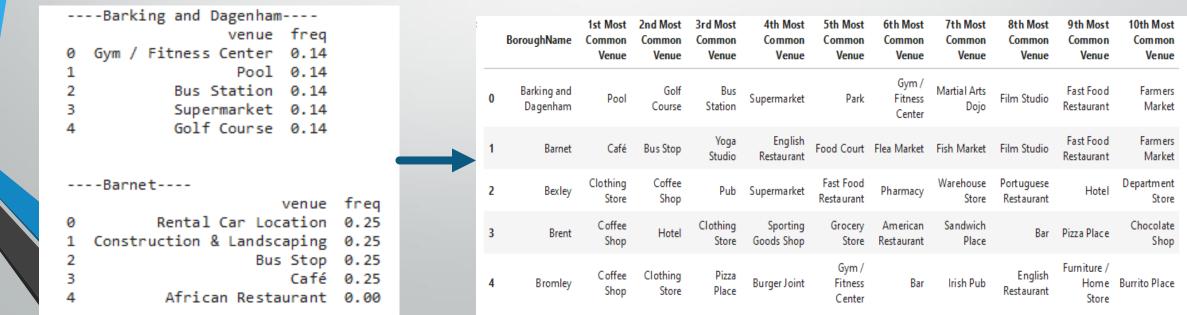


- (2) Exploratory analysis: London Map
- Using the coordinates of each of the London boroughs, I was able to visualise each of the boroughs on a map to be able to get more familiar with London's geography.



#### (3) Quster analysis

 With the 50 most popular venues acquired from Foursquare, I used onehot encoding to find out the most popular venue categories. Venues from the same boroughs were grouped by borough names and the popular categories by frequency.



#### (4) Cluster analysis: K-means clustering

Based on common venue categories, the boroughs were grouped into 5 different clusters, which are represented

on the map by different coloured dats.

	BoroughName	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue
0	Barking and Dagenham	Pool	Golf Course	Bus Station	Supermarket
1	Barnet	Café	Bus Stop	Yoga Studio	English Restaurant
2	Bexley	Clothing Store	Coffee Shop	Pub	Supermarket
3	Brent	Coffee Shop	Hotel	Clothing Store	Sporting Goods Shop
4	Bromley	Coffee Shop	Clothing Store	Pizza Place	Burger Joint



#### (4) Cluster analysis: K-means clustering

• Each of the clusters were named based on their individual characteristics

BoroughNam e	CrimePer1000	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	10th Most Common Venue
Barking and Dagenham	8.315668	0	Pool	Golf Course	Bus Station	Supermarket	Park	Gym / Fitness Center	Martial Arts Dojo	Film Studio	Fast Food Restaurant	Farmers Market

Cluster 0: Healthy area (Gym, pool)

BoroughName	CrimePer1000	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
Brent	7.869498	1	Coffee Shop	Hotel	Clothing Store	Sporting Goods Shop	Grocery Store	American Restaurant	Sandwich Place	Bar	Pizza Place
Camden	13.198009	1	Hotel	Café	Coffee Shop	Bakery	Breakfast Spot	Burger Joint	Italian Restaurant	Garden	Train Station
Kensington and Chelsea	12.148069	1	Café	Juice Bar	Restaurant	Clothing Store	Gym / Fitness Center	Burger Joint	Bakery	Sporting Goods Shop	Lebanese Restaurant

Cluster 1: Lively area (Pubs, restaurants)

#### (4) Cluster analysis: K-means clustering

• Each of the clusters were named based on their individual characteristics

BoroughName	CrimePer1000	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
Bexley	5.931181	2	Clothing Store	Coffee Shop	Pub	Supermarket	Fast Food Restaurant	Pharmacy	Warehouse Store
Bromley	6.240452	2	Coffee Shop	Clothing Store	Pizza Place	Burger Joint	Gym / Fitness Center	Bar	Irish Pub
Croydon	7.390638	2	Pub	Coffee Shop	Portuguese Restaurant	Asian Restaurant	Clothing Store	Park	Spanish Restaurant

Cluster 2: Shopping area (Coffee shops, clothing stores

Borough <b>Na</b> me	CrimePer1000	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	10th Most Common Venue
Barnet	6.684607	3	Café	Bus Stop	Yoga Studio	English Restaurant	Food Court	Flea Market	Fish Market	Film Studio	Fast Food Restaurant	Farmers Market

**Quster 3: Market area** 

#### (4) Cluster analysis: K-means clustering

• Each of the clusters were named based on their individual characteristics

BoroughName	CrimePer1000	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
Hounslow	8.259275	4	Metro Station	Park	Bed & Breakfast	Café	Yoga Studio	English Restaurant	Food Court	Flea Market	Fish Market

**Cluster 4: Traveller area** 

### 4. Results

- Using different analysis, I was able to discover the best areas to live based on the criteria of general atmosphere and safety.
   Now, we will look back at all the analysis performed in the project before making a conclusion on which area would be the best for HS to live in.
- For safety, I decided to normalise the crime to population video and reversed the safety score so that the closes value to 1
  represents the area that has the lowest crime per head.
- With regards to atmosphere, I decided to give an arbitrary value to each cluster on the basis of personal preference, since
  preference is difficult to quantify without being subjective. The highest value was given to the Shopping Area (Cluster 2)
  while the lowest was given to the Traveller Area (Cluster 4).

Finally, by adding the Safety score to the Atmosphere value, I was able to find the best area to live in, which was Richmond upon Thames, scoring 2.0.

BoroughName	CrimePer 1000	Cluster Labels	Safety	Atmosphere
Barking and Dagenham	8.315668	0.8	0.864607	0.0
Barnet	6.684607	3.0	0.943901	0.7
Bexley	5.931181	2.0	0.980529	1.0
Brent	7.869498	1.0	0.886298	0.9
Bromley	6.240452	2.0	0.965494	1.0

BoroughNam e	Safety	Atmosphere	Score
Richmond upon Thames	1.000000	1.0	2.000000
Harrow	0.995407	1.0	1.995407
Sutton	0.990868	1.0	1.990868
Merton	0.988647	1.0	1.988647
Bexley	0.980529	1.0	1.980529

### 5. Conclusion

- Based on the analysis, I have found that the 5 boroughs below would be the best places for H5 to live according on the criteria that he set out, based on general atmosphere and safety. They all belong to the Shopping Area cluster, which contains many clothing shops and coffee shops, as well as a number of restaurants that serve a range of cuisines. The only thing that sets them apart is the safety score.
- From the map, we can see that all of the top 5 boroughs are located far out from the city, which is due to the limitations held by the research carried out.
- As well as taking into account crime rates and the types of stores that
  exist in each area, I could've perhaps also taken into account the number
  of each type of store that exists in each borough, as well as other factors
  such as house/rental prices.
- Nevertheless, performing this research was still incredible enjoyable as I was able to explore each borough in a lot of depth.



### References

- "Borough Level Orime", London Datastore
- "List of London Boroughs"
- Foursquare API
- IBM Data Science Professional Certificate notes, Coursera