

# The Battle of Neighborhoods

Location for fast food restaurants in London

Coursera - IBM Data Science Specialization  
Applied Data Science Capstone

# The Ask/Problem Statement

## Location

Location is decided a London, also considering how big the city is also Narrowed down to Inner London region. But need to find a optimal location in a 319 km<sup>2</sup>

## Type of Restaurants

We are planning to a Fast Food restaurant.

## Problem statement

In a Metropolitan like London there are many places restaurant to open a Restaurants, We want to make a Data driven decision to finalize the location of pick to Borough to start the Restaurant

# Steps taken/Approach – High level

## Collecting Data

### Request and Pandas

Data is extracted from London Data Store and Wikipedia, then the indepth Location data from FourSquare.

## Cleaning Data

### Pandas

Unwanted columns are removed, columns are renamed to match the correct set and sorted to make it easy to process

## Processing Data

### SciPy and Folium

Required columns are merged, if needed additional data is added, required columns is grouped and weighted mean is calcualted

## Result Conclusion

### Pandas and Folium

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# Implementation Details

## Pre-Requirement

1. Need a Python 3.x environment
2. With the below mentioned python package
  - a. Jupyter notebook
  - b. numpy
  - c. pandas
  - d. folium
  - e. matplotlib
  - f. Scipy
  - g. Requests
3. Working Internet connection as more than half the data will be pulled from Internet
4. wget in local machine

# Implementation Details

## Collecting Data

1. Started with pulling 'london-borough-profiles.csv' from [London Data Store](#)
2. Pulled the List of London Borough and Local Authorities from [wikipedia - List of London boroughs](#)
3. Clean the unwanted columns and retain the required ones and merge both the Tables
4. Filter the data with only the Inner London Boroughs
5. Cleanup the Coordinates to correct Latitude and Longitudes
6. Collect below venues of interest from from [FourSquare API](#)
  - Two-year business survival rates
  - Population
  - Tech Startup
  - Supermarket
  - Soccer Field
  - School
  - Pub
  - Supermarket
  - Office
  - Historic Sites
  - Grocery Store
  - Factory
  - College Classroom
  - Coffee Shop
  - Café
  - Bus Stop
  - Fast Food Restaurant
  - Breakfast Spot
  - Sandwich Place

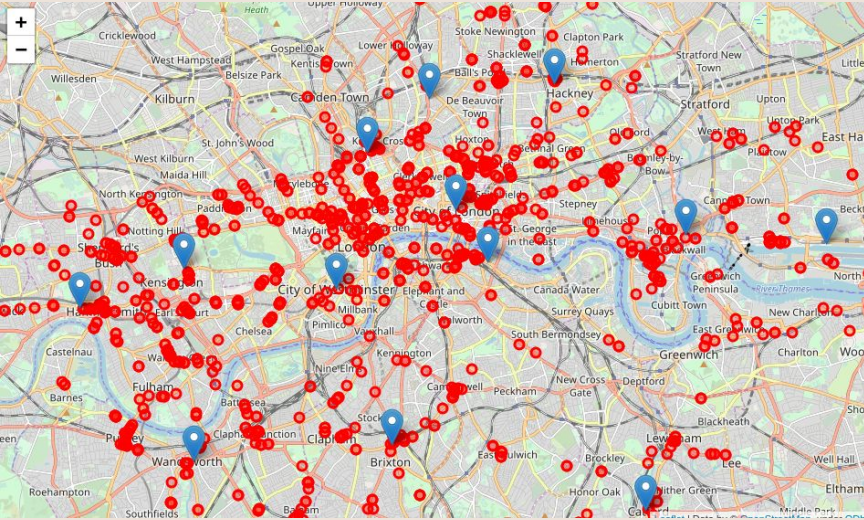


Boroughs picked for our Analysis

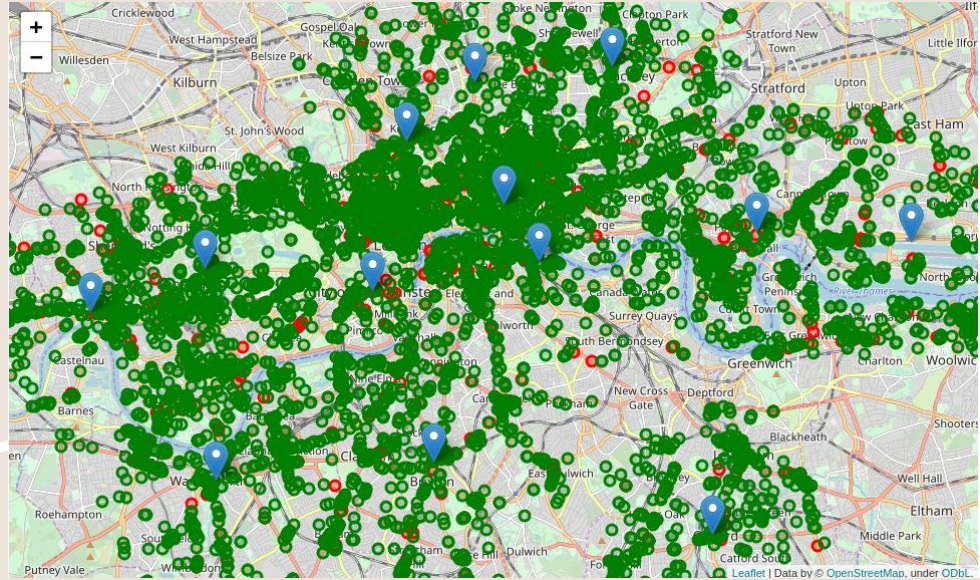
# Implementation Details

## Analysing Data

Let's begin analysis with visualizing the data and having a look at the data



Boroughs picked for our Analysis, with existing Fast Food restaurant



Boroughs picked for our Analysis with competition and potential neighbours



# Implementation Details

## Analysing Data

1. Let's start with consolidating all the existing data mapped to each of the Boroughs
2. Group them based on the name of locations based on the Borough
3. Then let's normalise the data so that the counts will overshoot when we normalise the data

Normalised data

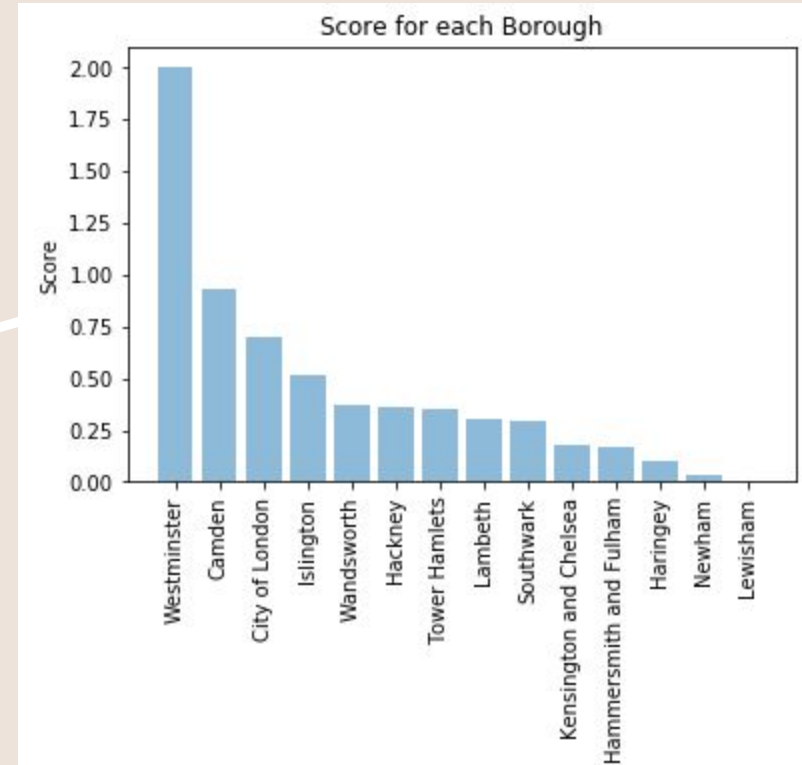
Tech Startup	Supermarket	Soccer Field	School	Residential Building (Apartment / Condo)	Pub	Professional & Other Places
0.870370	0.750000	0.8750	0.000000	0.555556	0.8	0.0
0.740741	1.000000	1.0000	0.222222	0.444444	0.6	0.0
1.000000	0.583333	0.1875	0.777778	0.666667	1.0	0.0
0.685185	0.333333	0.6250	0.666667	0.000000	0.2	0.0
0.000000	0.208333	0.1250	0.444444	0.111111	0.5	1.0
0.759259	0.875000	0.8125	0.444444	0.444444	0.8	0.0
0.685185	0.833333	0.3750	0.666667	0.444444	0.8	0.0
0.666667	0.375000	0.0000	0.888889	1.000000	0.1	0.0
0.018519	0.416667	0.0625	0.222222	0.111111	0.4	0.0
0.037037	0.250000	0.0625	0.777778	0.555556	0.0	0.0
0.851852	0.666667	0.6250	0.333333	0.777778	0.4	0.0
0.685185	0.291667	0.0625	1.000000	0.555556	0.3	0.0

	Borough	Tech Startup	Supermarket	Soccer Field	School	Residential Building (Apartment / Condo)	Pub	Professional & Other Places	Office	Historic Site	Grocery Store	Factory	College Classroom	Coffee Shop	Café	Bus Stop
0	City of London	56.0	30.0	17.0	6.0	48.0	50.0	0.0	69.0	30.0	57.0	78.0	42.0	65.0	32.0	50.0
1	Camden	49.0	36.0	19.0	8.0	47.0	48.0	0.0	77.0	31.0	52.0	82.0	43.0	51.0	43.0	50.0
2	Hackney	63.0	26.0	6.0	13.0	49.0	52.0	0.0	47.0	19.0	50.0	72.0	48.0	58.0	75.0	50.0
3	Hammersmith and Fulham	46.0	20.0	13.0	12.0	43.0	44.0	0.0	84.0	22.0	49.0	12.0	29.0	55.0	57.0	49.0
4	Haringey	9.0	17.0	5.0	10.0	44.0	47.0	2.0	61.0	16.0	44.0	17.0	5.0	51.0	49.0	50.0
5	Islington	50.0	33.0	16.0	10.0	47.0	50.0	0.0	76.0	36.0	53.0	80.0	48.0	55.0	56.0	50.0
6	Kensington and Chelsea	46.0	32.0	9.0	12.0	47.0	50.0	0.0	91.0	47.0	46.0	10.0	40.0	44.0	48.0	50.0
7	Lambeth	45.0	21.0	3.0	14.0	52.0	43.0	0.0	52.0	9.0	47.0	14.0	3.0	69.0	46.0	50.0
8	Lewisham	10.0	22.0	4.0	8.0	44.0	46.0	0.0	48.0	3.0	44.0	8.0	0.0	45.0	53.0	49.0
9	Newham	11.0	18.0	4.0	13.0	48.0	42.0	0.0	35.0	5.0	28.0	47.0	28.0	41.0	52.0	48.0
10	Southwark	55.0	28.0	13.0	9.0	50.0	46.0	0.0	76.0	35.0	55.0	70.0	43.0	61.0	38.0	50.0
11	Tower Hamlets	46.0	19.0	4.0	15.0	48.0	45.0	0.0	99.0	13.0	43.0	54.0	49.0	48.0	41.0	50.0
12	Wandsworth	61.0	12.0	10.0	15.0	46.0	48.0	0.0	64.0	11.0	48.0	14.0	15.0	48.0	59.0	48.0
13	Westminster	59.0	27.0	17.0	10.0	48.0	46.0	0.0	73.0	33.0	50.0	32.0	47.0	51.0	47.0	50.0

# Implementation Details

## Analysing Data

1. Then a weighing amount is applied on the each category based on the importance that we feel
2. Weighing amount for Competition in that same are also applied but are given negative score
3. Normalised the weighted mean some for all the categories are grouped by Borough are sorted and plotted in a Bar-chart
4. We can see a clear trend that the preferred location is Westminster





# Conclusion

**Borough of Westminster** is the best place to start a Fast Food Restaurant

As the Data is so obvious when we do a weighted score of the Boroughs we have not proceeded with K-Mean algorithm. When we add more parameters like, Restriction and benefits in each of Borough Council, Fresh data on the transport and tourists stops, Safety in each of the council , Cost of running in each Council.

