

IDENTIFY A STRATEGIC LOCATION TO OPEN A BRAZILIAN TAKE AWAY RESTAURANT IN LONDON, UNITED KINGDOM

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Introduction

- The success of establishing a new restaurant depends on several factors: demand, brand loyalty, quality of food, competition, and so on.
- In most cases, a restaurant's location plays an essential determinant for its success.
- Whether opening your first full-service restaurant, it's important to understand what to look out for when choosing a new restaurant location.

PROBLEM STATEMENT

- Opening a Brazilian take away restaurants in London.
- Objective is to identify the optimal neighborhood location to open.
- Key factors to consider are: spending power of the London population, distribution of Brazilian restaurants, distance to public transport station.
- To identify the ideal London neighborhood clusters group, we use Foursquare API, data scraping, geopy, pycaret to build the clustering model and matplotlib/seaborn to visualize data during our EDA (Exploratory Data Analysis).

KEY CLIENTS

 Brazilian Family of chefs, highly experienced in cooking for small to medium restaurants

DATA ACQUISITION, WRANGLING AND CLEANING

- Data scraped (neighbourhoods, postal codes, geographical coordinates, population and income) will be scraped from here:
 https://www.doogal.co.uk/UKPostcodesCSV.ashx? area=London0.
- For returning the number of Brazilian restaurants in the vicinity of each neighborhood, we will use Foursquare API, more specifically, its *explore* function.

DATA CLEANING

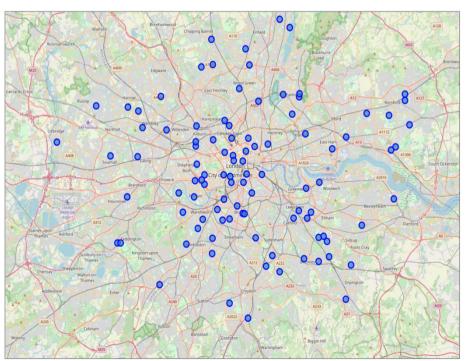
- Drop columns that do not contain useful information for the EDA (Exploratory Data Analysis) and the clustering model.
- Drop of rows with null or non existent values
- Final dataframe has been simplified and sampled to 100 rows because the folium library and the colab platform crashed several times.

Cleaned dataset

	District	Latitude	Longitude	Population	Households	Distance to station	Average Income
0	Bromley	51.404543	0.014195	34.0	21.0	0.462939	63100
1	Bromley	51.408058	0.015874	38.0	37.0	0.083058	63100
2	Bromley	51.409191	0.010068	1.0	1.0	0.489492	56100
3	Bromley	51.400462	0.016716	4.0	4.0	0.067905	63100
4	Bromley	51.401684	0.015705	14.0	6.0	0.219358	63100
140844	Hillingdon	51.623983	-0.495253	18.0	5.0	2.347240	54200
140845	Hillingdon	51.626955	-0.494143	22.0	15.0	2.048740	54200
140846	Hillingdon	51.628575	-0.499204	2.0	1.0	2.191290	54200
140847	Barnet	51.643292	-0.255958	11.0	6.0	1.960300	58000
140848	Barnet	51.642309	-0.256627	71.0	55.0	1.984360	58000

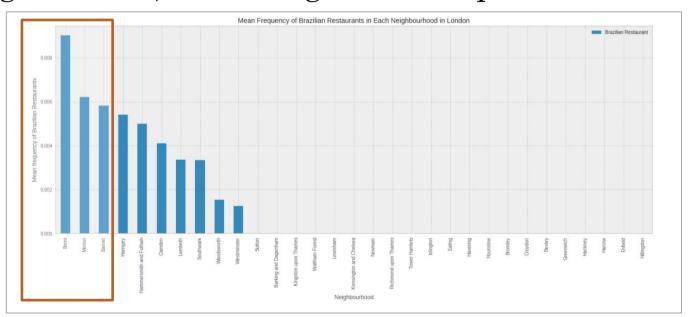
EXPLORATORY DATA ANALYSIS

• Folium mapping of city of London and neighborhoods



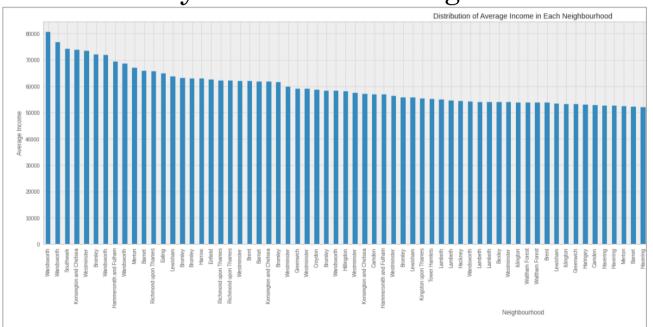
FREQUENCY DISTRIBUTION OF BRAZILIAN RESTAURANTS

• Higher number of Brazilian restaurants in a neighborhood, the stronger the competition.



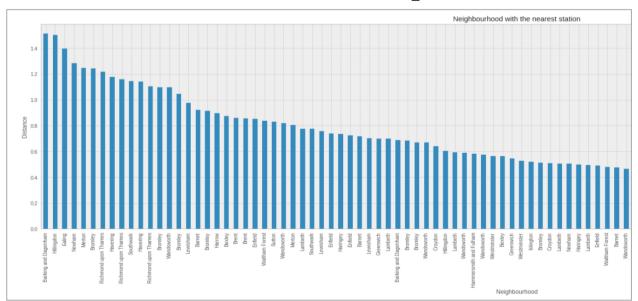
DISTRIBUTION OF MEDIAN HOUSEHOLD INCOME

 Neighborhoods distributed towards around the mean can readily afford and indulge themselves



DISTANCE TO STATION BY NEIGHBORHOODS

• Most people live in London without having a car and the public transport is quite important to reach the areas and venues of preference



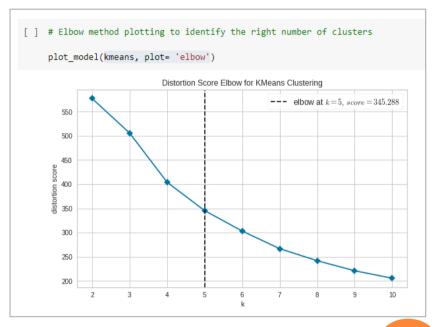
CLUSTERING MODELING

Data Pre-processing of cluster model (Pycaret):

- PyCaret is an open source low-code machine learning library in Python that aims to reduce the hypothesis to insights cycle time in a ML experiment. It enables data scientists to perform end-to-end experiments quickly and efficiently (source: https://pycaret.org).
- The library allows to implement a simple preprocessing just with few lines of codes.

K-MEANS CLUSTERING MODEL

- To implement a k-means model usually is important to assign a number of clusters.
- Pycaret highlights the suitable number of clusters calculated with the squared error as a performance metric.
- The suitable number of clusters is 5:

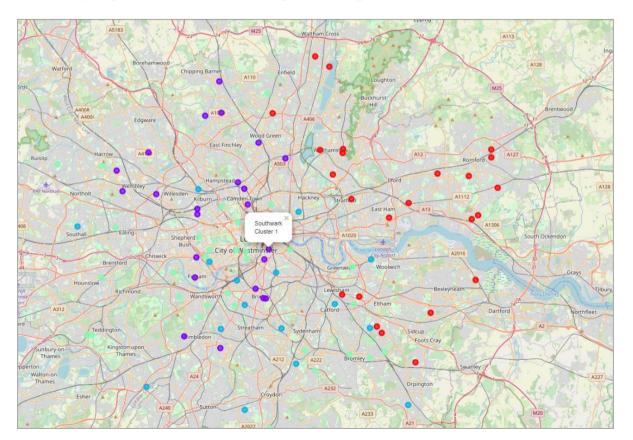


CLUSTER LABELS

• Here below we have the final data frame with the neighborhoods and the clusters (from 0 to 4 as in Python the counting starts from zero so 5 clusters: 0,1,2,3,4).

	Neighbourhood	Latitude	Longitude	Population	Households	Distance to station	Average Income	Brazilian Restaurant	Cluster
0	Westminster	51.528918	-0.206048	71.0	20.0	0.565596	44500	0.00125	Cluster 1
1	Westminster	51.496019	-0.137685	46.0	27.0	0.433721	57500	0.00125	Cluster 3
2	Westminster	51.522106	-0.149017	3.0	1.0	0.316775	73500	0.00125	Cluster 3
3	Westminster	51.513050	-0.131427	5.0	3.0	0.294475	54000	0.00125	Cluster 3
4	Westminster	51.535685	-0.171829	6.0	2.0	0.180020	59900	0.00125	Cluster 3

CLUSTERS ON THE LONDON MAP



- Cluster 0:
- Low spending power
- No presence of competition
- mid/high distance to the station

	Neighbourhood	Latitude	Longitude	Population	Households	Distance to station	Average Income	Brazilian Restaurant	Cluster
8	Bromley	51.425513	0.052632	34.0	17.0	1.098220	45800	0.0	0
9	Bromley	51.399587	0.100309	71.0	28.0	0.683057	42300	0.0	0
10	Bromley	51.430848	0.045822	35.0	11.0	1.045150	45800	0.0	0
32	Waltham Forest	51.584746	-0.033925	30.0	11.0	0.409922	50700	0.0	0
33	Waltham Forest	51.582277	-0.002204	89.0	31.0	0.478209	53700	0.0	0
34	Waltham Forest	51.563486	-0.015653	6.0	6.0	0.837416	46700	0.0	0
35	Waltham Forest	51.585813	-0.001676	58.0	29.0	0.097137	53700	0.0	0
40	Enfield	51.657057	-0.020592	1.0	1.0	0.723715	38700	0.0	0
42	Enfield	51.666158	-0.040539	76.0	30.0	0.853476	38300	0.0	0
43	Enfield	51.616572	-0.100123	16.0	5.0	0.737417	46200	0.0	0
47	Newham	51.542189	0.010360	36.0	15.0	0.503985	50200	0.0	0
48	Newham	51.525888	0.063026	95.0	33.0	1.283330	40400	0.0	0
66	Lewisham	51.458596	-0.002975	40.0	17.0	0.702157	53500	0.0	0
70	Havering	51.585085	0.206607	69.0	31.0	0.356473	52600	0.0	0
71	Havering	51.527812	0.188181	83.0	30.0	1.176440	44000	0.0	0
72	Havering	51.551881	0.215761	100.0	47.0	0.351363	52100	0.0	0
73	Havering	51.578145	0.206487	56.0	23.0	0.414984	52600	0.0	0
74	Havering	51.524680	0.179122	171.0	57.0	1.142380	44000	0.0	0
86	Greenwich	51.457183	0.021233	60.0	30.0	0.699302	53200	0.0	0
92	Bexley	51.443029	0.121412	88.0	39.0	0.874050	54000	0.0	0
93	Bexley	51.471565	0.185028	69.0	21.0	0.565264	44200	0.0	0
94	Barking and Dagenham	51.561947	0.175161	26.0	11.0	1.512370	43700	0.0	0
95	Barking and Dagenham	51.538752	0.090788	116.0	35.0	0.688178	49900	0.0	0
96	Barking and Dagenham	51.564448	0.131013	61.0	24.0	0.423293	40800	0.0	0

- Cluster 1:
- Mid spending power
- Med/high presence of competition
- Medium distance to the station

	Neighbourhood	Latitude	Longitude	Population	Households	Distance to station	Average Income	Brazilian Restaurant	Cluster
0	Westminster	51.528918	-0.206048	71.0	20.0	0.565596	44500	0.001250	1
18	Lambeth	51.455669	-0.113461	9.0	5.0	0.777299	54000	0.003367	
21	Lambeth	51.455202	-0.109609	63.0	27.0	0.506102	54000	0.003367	
22	Lambeth	51.489647	-0.112059	2.0	1.0	0.495755	54600	0.003367	
23	Lambeth	51.463865	-0.124353	22.0	9.0	0.402040	54900	0.003367	
24	Brent	51.582646	-0.274947	57.0	19.0	0.353669	50900	0.009009	
25	Brent	51.533333	-0.206048	92.0	46.0	0.101633	53700	0.009009	
26	Brent	51.566621	-0.319504	11.0	5.0	0.860764	61900	0.009009	
27	Brent	51.549053	-0.310892	69.0	24.0	0.457962	47800	0.009009	
28	Brent	51.546484	-0.263667	62.0	19.0	0.857033	40900	0.009009	
44	Barnet	51.614525	-0.194904	70.0	23.0	0.715983	65800	0.005814	
45	Barnet	51.644459	-0.175013	44.0	21.0	0.474490	61700	0.005814	
46	Barnet	51.617341	-0.172074	37.0	16.0	0.923050	52200	0.005814	
57	Southwark	51.498460	-0.105932	26.0	14.0	0.410530	48300	0.003333	
64	Haringey	51.591314	-0.120335	43.0	16.0	0.734368	53100	0.005405	
65	Haringey	51.577388	-0.082158	77.0	28.0	0.498445	44400	0.005405	
75	Merton	51.422477	-0.224531	19.0	8.0	1.247450	67000	0.006211	
76	Merton	51.411943	-0.173447	85.0	32.0	0.805762	52500	0.006211	
79	Camden	51.518421	-0.132706	1.0	1.0	0.269255	46800	0.004098	
81	Camden	51.537440	-0.135533	24.0	11.0	0.407740	48500	0.004098	
82	Camden	51.551023	-0.139781	23.0	10.0	0.071352	57000	0.004098	
83	Camden	51.556426	-0.148735	34.0	15.0	0.185730	52900	0.004098	
98	Hammersmith and Fulham	51.491253	-0.207484	6.0	5.0	0.154938	56900	0.005000	
99	Hammersmith and Fulham	51.473862	-0.209332	39.0	11.0	0.581015	69400	0.005000	

- Cluster 2:
- Mid spending power
- Low presence of competition
- Medium/high distance to the station

	Neighbourhood	Latitude	Longitude	Population	Households	Distance to station	Average Income	Brazilian Restaurant	Cluster
14	Bromley	51.362141	0.090192	73.0	38.0	1.241690	61600	0.000000	2
19	Lambeth	51.449614	-0.139185	83.0	29.0	0.697894	49700	0.003367	
20	Lambeth	51.429065	-0.088232	102.0	36.0	0.594626	50100	0.003367	
38	Kensington and Chelsea	51.487289	-0.188915	109.0	81.0	0.458343	57100	0.000000	
50	Wandsworth	51.428941	-0.172968	148.0	53.0	0.363703	51300	0.001538	
52	Wandsworth	51.471192	-0.150287	77.0	36.0	0.465168	54100	0.001538	
56	Southwark	51.478168	-0.095108	127.0	45.0	1.143710	44000	0.003333	
61	Sutton	51.357886	-0.139529	117.0	38.0	0.832058	44600	0.000000	
67	Lewisham	51.429587	0.036631	144.0	63.0	0.976703	44200	0.000000	
68	Lewisham	51.450695	-0.012703	173.0	66.0	0.756707	55700	0.000000	
78	Ealing	51.518183	-0.373996	108.0	27.0	1.395690	47700	0.000000	
80	Camden	51.550415	-0.204611	186.0	105.0	0.385689	51400	0.004098	
88	Greenwich	51.489310	0.038067	146.0	56.0	0.546876	52000	0.000000	
90	Croydon	51.398141	-0.067555	188.0	85.0	0.509361	45200	0.000000	
91	Kingston upon Thames	51.378398	-0.277722	173.0	70.0	0.209291	55400	0.000000	
97	Tower Hamlets	51.530888	-0.063582	118.0	50.0	0.452014	55100	0.000000	

- Cluster 3:
- Mid/high spending power
- Low presence of competition
- Medium distance to the station

	Neighbourhood	Latitude	Longitude	Population	Households	Distance to station	Average Income	Brazilian Restaurant	Cluster
1	Westminster	51.496019	-0.137685	46.0	27.0	0.433721	57500	0.001250	3
2	Westminster	51.522106	-0.149017	3.0	1.0	0.316775	73500	0.001250	3
3	Westminster	51.513050	-0.131427	5.0	3.0	0.294475	54000	0.001250	3
4	Westminster	51.535685	-0.171829	6.0	2.0	0.180020	59900	0.001250	3
5	Westminster	51.512436	-0.113124	4.0	1.0	0.160563	59000	0.001250	3
6	Westminster	51.497664	-0.151261	32.0	17.0	0.526160	56300	0.001250	3
7	Westminster	51.489636	-0.136115	9.0	6.0	0.208241	61900	0.001250	3
11	Bromley	51.406526	0.007654	36.0	17.0	0.353919	63100	0.000000	3
12	Bromley	51.403588	0.036137	55.0	21.0	0.669057	62900	0.000000	3
13	Bromley	51.408445	0.056699	26.0	11.0	0.321041	72000	0.000000	3
15	Bromley	51.392147	-0.040713	7.0	6.0	0.913357	58300	0.000000	3
16	Bromley	51.409708	-0.055161	52.0	20.0	0.511303	55700	0.000000	3
17	Hackney	51.528557	-0.089805	11.0	10.0	0.315468	54300	0.000000	3
29	Richmond upon Thames	51.423487	-0.360588	65.0	25.0	1.158880	62200	0.000000	3
30	Richmond upon Thames	51.423439	-0.354513	35.0	16.0	1.217460	62200	0.000000	3
31	Richmond upon Thames	51.477996	-0.239360	20.0	8.0	1.104870	65600	0.000000	3
36	Kensington and Chelsea	51.497809	-0.188800	52.0	18.0	0.376699	73900	0.000000	3
37	Kensington and Chelsea	51.508964	-0.204267	2.0	1.0	0.256720	61700	0.000000	3
39	Kensington and Chelsea	51.491967	-0.194232	13.0	8.0	0.068090	49800	0.000000	3
41	Enfield	51.634244	-0.107997	38.0	16.0	0.489925	62600	0.000000	3
49	Wandsworth	51.456981	-0.231109	80.0	25.0	1.096980	68500	0.001538	3
51	Wandsworth	51.459925	-0.163482	71.0	25.0	0.668297	80600	0.001538	3
53	Wandsworth	51.448371	-0.153267	29.0	16.0	0.574968	71800	0.001538	3
54	Wandsworth	51.449184	-0.192121	32.0	12.0	0.819897	58200	0.001538	3
55	Wandsworth	51.439621	-0.201794	69.0	31.0	0.588494	76700	0.001538	3
58	Southwark	51.498154	-0.075531	38.0	16.0	0.775141	74200	0.003333	3
59	Harrow	51.571271	-0.339808	35.0	13.0	0.897249	62900	0.000000	3
60	Hounslow	51.473786	-0.341075	97.0	35.0	0.294194	51600	0.000000	3
62	Hillingdon	51.572291	-0.402240	61.0	20.0	0.603354	58100	0.000000	3
69	Lewisham	51.450751	0.014012	74.0	35.0	0.116572	63700	0.000000	3
77	Ealing	51.517266	-0.320989	35.0	11.0	0.413129	64800	0.000000	3
84	Islington	51.523194	-0.107313	8.0	5.0	0.366736	53800	0.000000	3
85	Islington	51.537215	-0.102420	22.0	15.0	0.520079	53300	0.000000	3

- Cluster 4:
- Low spending power
- No presence of competition
- Highest distance to the station

	Neighbourhood	Latitude	Longitude	Population	Households	Distance to station	Average Income	Brazilian Restaurant	Cluster
63	Hillingdon	51.533017	-0.479713	1129.0	0.0	1.50396	45800	0.0	4

CONCLUSION

- The most promising group of neighborhoods for opening an Brazilian Restaurant, with a niche in Brazilian cuisine, appears to be 'Cluster Label 3'.
- The medium high spending power of the neighborhoods allows them to readily afford the slightly upscaled prices of the client's Brazilian restaurant menu.
- The number of competitors is pretty low, that would definitely help build a brand in the area for people curious to try Brazilian cuisine.
- Our client could more specifically consider Richmond and some areas of Kensington and Chelsea characterized by a medium high spending average.
- However cluster 3 shows a medium distance to the station and that probably could be considered as an insight to activate a solid promotion to the locals to make sure there is a significant and remunerative client base to limit any potential issue of the less easy to reach of the restaurant from people of other areas of London.
- In conclusion, the extensive analysis above would greatly increase the likelihood of the restaurant's success.