# Coursera Capstone IBM Applied Data Science Capstone

# Analysis of cafes within Colombo, Sri Lanka

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**Introduction and Business Problem** 

## Introduction

Sri Lanka is a relatively small island directly south of India, with a population of more than 21 Million residents (as of 2018) of which around 5.6 Million populate the capital of Colombo. Colombo is also the biggest city within the island and serves as the financial capital and a tourist destination. As such Colombo serves as a hunting ground for up and coming businesses such as cafes and restaurants. It a can daunting for a person to search for the best place to eat/drink with the various options presented in this ever-growing city.

## **Business Problem**

For this capstone I will be creating a guide utilizing the Foursquare API, to categorize cafes within Colombo, after which I will be clustering the cafes according the amount of likes they have received. My aim is to build a guide to traverse through the various coffee shops of Colombo and select the best ones.

## **Data**

To solve this problem, I will be utilizing the following data,

- A list of the inter-city limits of Colombo.
- The latitudes and longitudes of those areas for visualization.
- Venue Data utilizing the foursquare API.

# Data gathering

To obtain the list of inter-city limits I will be extracting the content from the Wikipedia Page (<a href="https://en.wikipedia.org/wiki/Colombo">https://en.wikipedia.org/wiki/Colombo</a>) which contains the inter-city limits. I will be using web scraping techniques to extract the data with the help of python libraries such as beautiful soup and requests. The Latitudes and Longitudes of the Inter-city Limits will be obtained using pythons Geocoder package. After which I will be utilizing the Foursquare API to get venue Data of the relevant locations.

# Methodology

When starting this project my thought process behind it was that the more likes they receive the better the coffee shop is. While this is an inaccurate measurement for measuring quality due to the restrictions of the free tier account from Foursquare, I was un-able to retrieve rating. This error has been somewhat rectified by binning the data into categorical variables.

I will take the gathered data (see above in Data Acquisition Approach and Data Required sections) and will create a k-means clustering algorithm that groups restaurants into 4 clusters so that people looking for a coffee shop can easily see which coffee shops are the best and where in Hoboken they can look to eat.

# **Results**

Running my clustering algorithm, I was able to generate four clusters of coffee shops. These are as follows:

### Cluster 1

- Characteristics
  - o Poor quality food

	Neighborhood	Latitude	Longitude	Venueld	VenueName	VenueLatitude	VenueLongitude	VenueCategory	total likes	total likes_cat	label
0	Slave Island	6.926520	79.850200	4d8a1106d85f37044a2ff5db	The Dining Room	6.929095	79.849568	Coffee Shop	22	poor	0
2	Slave Island	6.926520	79.850200	51dcff7e498e94b774bb1a0e	Pebbles cafe, Taj Samudra	6.923323	79.846273	Coffee Shop	11	poor	0
4	Slave Island	6.926520	79.850200	4c8b43d575ba9c749e81575c	Goodies	6.928913	79.849598	Coffee Shop	19	poor	0
5	Slave Island	6.926520	79.850200	5155636fe4b08e31444f9e39	Cafe Seventy Seven	6.920389	79.853801	Coffee Shop	5	poor	0
6	Slave Island	6.926520	79.850200	4de0e9d11f6ece647385c007	The Verandah	6.920024	79.845865	Coffee Shop	14	poor	0
8	Union Place	6.918760	79.861130	50cc7812e4b05e0e4e80e1e9	Odel Boulevard	6.916773	79.865148	Coffee Shop	14	poor	0
9	Union Place	6.918760	79.861130	590d92a2cad1b67f0cbf302e	Life Food	6.916240	79.853326	Coffee Shop	8	poor	(
11	Union Place	6.918760	79.861130	4e2959f952b1c37d35caca13	M3 Cafe	6.914540	79.857053	Coffee Shop	14	poor	(
13	Union Place	6.918760	79.861130	5155636fe4b08e31444f9e39	Cafe Seventy Seven	6.920389	79.853801	Coffee Shop	5	poor	(
4	Union Place	6.918760	79.861130	5a609723ee628b483065da85	Caramel Pumpkin	6.916296	79.869043	Coffee Shop	13	poor	(
15	Kollupitiya	6.912680	79.851350	553915b3498eb10d08ee7371	The Coffee Bean & Tea Leaf	6.911989	79.851029	Coffee Shop	24	poor	(
17	Kollupitiya	6.912680	79.851350	590d92a2cad1b67f0cbf302e	Life Food	6.916240	79.853326	Coffee Shop	8	poor	(
8	Kollupitiya	6.912680	79.851350	59da19494928143ec58f0c33	Tea Avenue	6.908325	79.849255	Coffee Shop	9	poor	(
22	Kollupitiya	6.912680	79.851350	4e2959f952b1c37d35caca13	M3 Cafe	6.914540	79.857053	Coffee Shop	14	poor	(
23	Kollupitiya	6.912680	79.851350	5489298a498e3141e039ed7d	Delifrance	6.906575	79.853310	Coffee Shop	6	poor	(
24	Bambalapitiya	6.904710	79.855070	55e5188a498e34b61c727637	The Coffee Bean & Tea Leaf	6.903622	79.852232	Coffee Shop	18	poor	(
25	Bambalapitiya	6.904710	79.855070	5aae6995e65d0c06df889c1d	The Embazzy	6.906572	79.858438	Coffee Shop	14	poor	(
30	Bambalapitiya	6.904710	79.855070	527b8ca211d2fbb60218d34c	Mitsi's Delicacies	6.902005	79.857477	Coffee Shop	21	poor	(
31	Bambalapitiya	6.904710	79.855070	59da19494928143ec58f0c33	Tea Avenue	6.908325	79.849255	Coffee Shop	9	poor	(
2	Bambalapitiya	6.904710	79.855070	5489298a498e3141e039ed7d	Delifrance	6.906575	79.853310	Coffee Shop	6	poor	(
33	Bambalapitiya	6.904710	79.855070	553915b3498eb10d08ee7371	The Coffee Bean & Tea Leaf	6.911989	79.851029	Coffee Shop	24	poor	(
34	Bambalapitiya	6.904710	79.855070	4fcf6db0e4b0f54f2917ec83	Cloud Cafe	6.897779	79.856392	Coffee Shop	15	poor	(
8	Havelock Town	6.887580	79.862560	59cba4328d0a536b6c9a39a8	Cafe Noir Blanc	6.892419	79.869752	Coffee Shop	11	poor	(

## Cluster 2

- Characteristics
  - o Good quality food

	Neighborhood	Latitude	Longitude	Venueld	VenueName	VenueLatitude	VenueLongitude	VenueCategory	total likes	total likes_cat	label
3	Slave Island	6.926520	79.850200	4c7192ccf4d476b035bd5acf	Coffee Stop - Cinnamon Grand	6.917826	79.848547	Coffee Shop	133	great	1
7	Union Place	6.918760	79.861130	4c0771cb3cbed13a7af40bc0	Coco Veranda	6.916424	79.868338	Coffee Shop	571	great	1
19	Kollupitiya	6.912680	79.851350	4c7192ccf4d476b035bd5acf	Coffee Stop - Cinnamon Grand	6.917826	79.848547	Coffee Shop	133	great	1
20	Kollupitiya	6.912680	79.851350	4c0b3fd5340720a191f28793	The Commons	6.908486	79.858051	Coffee Shop	286	great	1
26	Bambalapitiya	6.904710	79.855070	4b991f3df964a520ca6235e3	The Gallery Café	6.898672	79.854599	Coffee Shop	269	great	1
27	Bambalapitiya	6.904710	79.855070	4c0b3fd5340720a191f28793	The Commons	6.908486	79.858051	Coffee Shop	286	great	1
9	Bambalapitiya	6.904710	79.855070	4cb9f8f543ec6dcb891aa131	The Barefoot Cafe	6.896644	79.853765	Coffee Shop	199	great	1
86	Havelock Town	6.887580	79.862560	4efaaa3df790731254d52c57	Java Lounge	6.892233	79.867115	Coffee Shop	318	great	1
9	Kirulapana	6.878604	79.873269	4c305b767cc0c9b69b40ed9a	Barista	6.877262	79.873039	Coffee Shop	207	great	1
45	Kirilapone North	6.881350	79.875880	4c305b767cc0c9b69b40ed9a	Barista	6.877262	79.873039	Coffee Shop	207	great	1
55	Pamankada	6.879540	79.869370	4c305b767cc0c9b69b40ed9a	Barista	6.877262	79.873039	Coffee Shop	207	great	1
61	Cinnamon Gardens	6.909510	79.867750	4b991ccbf964a5201e6235e3	The Coffee Bean & Tea Leaf	6.911452	79.866747	Coffee Shop	346	great	1
64	Cinnamon Gardens	6.909510	79.867750	4c0771cb3cbed13a7af40bc0	Coco Veranda	6.916424	79.868338	Coffee Shop	571	great	1
71	Cinnamon Gardens	6.909510	79.867750	4f0d7c39e4b0281b8c4e48ff	Choco Luv	6.909549	79.874280	Coffee Shop	139	great	1
75	Borella	6.916690	79.877620	4f0d7c39e4b0281b8c4e48ff	Choco Luv	6.909549	79.874280	Coffee Shop	139	great	1
В0	Maradana	6.924730	79.865700	4c0771cb3cbed13a7af40bc0	Coco Veranda	6.916424	79.868338	Coffee Shop	571	great	1

#### Cluster 3

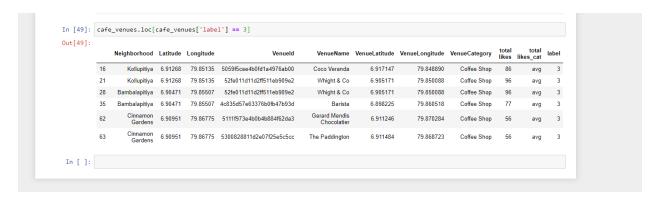
#### Characteristics

Below average quality food

cafe\_venues.loc[cafe\_venues['label'] == 2] total likes total Neighborhood Latitude Longitude Venueld VenueName VenueLatitude VenueLongitude VenueCategory label likes\_cat below Slave Island 6.92652 79.85020 4b6fa5cdf964a52006f92ce3 Barista 6.932601 79.843855 Coffee Shop 38 2 below 10 Union Place 6.91876 79.86113 56716368498ee66e925b8ffa Black Cat Cafe 6.915373 79.869426 Coffee Shop 2 avg Laksala Museum Gallery Café below 12 Union Place 6.91876 79.86113 521d962c498e314c4ac39d7a 6.910536 79.861873 Coffee Shop 29 2 avg below 37 Havelock Town 6.88758 79.86256 53c7cc97498e210e88986473 6.882134 79.858934 Coffee Shop 42 2 Java Lounge below 51 Wellawatte 6.87924 79.85989 53c7cc97498e210e88986473 Java Lounge 6.882134 79.858934 Coffee Shop 2 avg That Cafe in Colombo below 53 Wellawatte 6.87924 79.85989 52e79d2211d21513c70f616a 6.875635 79.861391 Coffee Shop 30 2 avg That Cafe in below 60 Pamankada 6.87954 6.875635 79.861391 79.86937 52e79d2211d21513c70f616a Coffee Shop 30 2 London House of below 65 6.90951 79.86775 562e0cc6498e85c97a0b9753 6.909475 79.866491 Coffee Shop 31 2 Gardens Coffee avq Cinnamon Gardens below 66 6.90951 79.86775 56716368498ee66e925b8ffa Black Cat Cafe 6.915373 79.869426 Coffee Shop 31 2 avg Cinnamon 6.908334 below 67 6.90951 79.86775 4d3e8f813ec9a35dd2525b81 Aviraté Café 79.865088 Coffee Shop 36 2 Gardens Cinnamon Laksala Museum below 68 6.90951 79.86775 521d962c498e314c4ac39d7a 6.910536 79.861873 Coffee Shop 29 2 Gardens Gallery Café Cinnamon below 70 6.90951 79.86775 4e608ec4b993678bb574f058 Caffè Pascucci 6.910128 79.871694 Coffee Shop 41 2 Gardens avq Cinnamon below Tea Breeze 6 905846 79.863597 72 6.90951 79 86775 52cbe6e8498e966d6dea55b7 Coffee Shop 26 2 Gardens avg below 74 Borella 6.91669 79.87762 56716368498ee66e925b8ffa Black Cat Cafe 6.915373 79.869426 Coffee Shop 31 2 avg below Coffee Shop 76 Borella 6.91669 79.87762 4e608ec4b993678bb574f058 Caffè Pascucci 6.910128 79.871694 41 2 avg

#### Cluster 4

- Characteristics
  - Average quality food



#### **Map of Clusters for Users**

