Report

Introduction

With a population of more than 10 million, the city of Bangalore in India is one of the most populous cities in India with a significant migrant population. With the blossoming number of apartments across the city, the choice of apartment to rent is determined not just by the proximity to the workplace but the quality of the neighbourhood. This project will cluster together similar neighborhoods based on the popular venue categories in each neighbourhood and map the various apartments in these neighborhoods.

This will help people who are looking to move into the city or to see if there are similar neighborhoods to their current neighborhoods which they can consider relocating.

Data

The data used for this project will be obtained from the following sources:

a) For the list of neighborhood in Bangalore, although the dataset available in Kaggle (https://www.kaggle.com/rmenon1998/bangalore-neighborhoods), was initially considered, later, the wikipedia page on neighborhoods in Bangalore which is available here:

https://en.wikipedia.org/wiki/List_of_neighbourhoods_in_Bangalore was used.

- b) For latitude and longitude details for the above neighborhoods, Google API has been used.
- c) Finally, Foursquare API will be used to collate neigborhood venue information.

Methodology

Exploratory Data Analysis: Kaggle has a dataset of Bangalore neighborhoods along with their latitude and longitude details available. Initially, the idea was to go with this available data set. An extract of the data set downloaded from Kaggle is as follows:

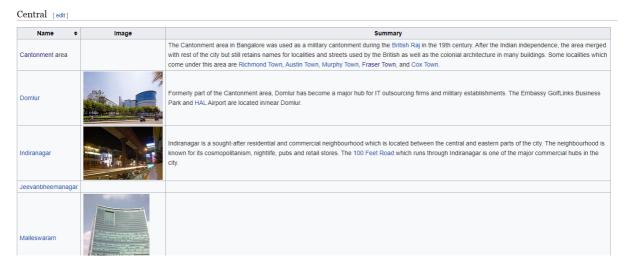
	Unnamed: 0	Neighborhood	Latitude	Longitude
0	0	Agram	45.813177	15.977048
1	1	Amruthahalli	13.066513	77.596624
2	2	Attur	11.663711	78.533551
3	3	Banaswadi	13.014162	77.651854
4	4	Bellandur	58.235358	26.683116
5	5	Bhattarahalli	13.025800	77.714279
6	6	Bidrahalli	14.577426	74.928560
7	7	Byatarayanapura	13.062074	77.596392
8	8	Devanagundi	12.973613	77.839402
9	9	Devasandra	12.757227	77.647280
10	10	Doddagubbi	14.452076	75.510799
11	11	Doddanekkundi	12.975720	77.694042

Sample of the data set available on Kaggle

There were two main issues with the above data:

- one- The neighborhoods listed here are not commonly understood as neighborhoods by residents of Bangalore. On further analysis, it was realised that this data set equates post offices as neighborhoods. While this list may be useful in some use cases, for the purposes of this Project, neighborhood is to be given a more layman interpretation- i.e., we want a list which identifies neighborhoods as commonly understood by the residents of Bangalore.
- two- the latitude and longitude data in this is very evidently incorrect. For instance, for neigborhood 'Agram', latitude is shown as 45.813 and for 'Amruthahalli', latitude is shown as 13.06. No two neighborhoods within the same city can have such vide variation in latitudes.

Accordingly, it was decided not to use this dataset, but to create a new data set of Bangalore neighbourhoods by scraping from the net. Wikipedia has a page on Bangalore neighborhoods which lists out the correct neighborhoods, as below.



snapshot of the wiki page on Bangalore neighborhoods

However, as evident above, this list does not have the latitude and longitude data. Further, the names of neighborhoods are provided in different tables in the wiki page along with images and summary which are not useful for our context.

Accordingly, I scraped only the neighborhood names from the Wikipedia page and created a new dataframe:

	Name
0	Cantonment area
1	Domlur
2	Indiranagar
3	Jeevanbheemanagar
4	Malleswaram

But as you can see, this dataframe does not have the latitude and longitude details of these neighborhoods. We would need the locational data to easily plot these neighborhoods on map.

Using Google Maps' API, latitude and longitude details were pulled for each of these neighborhoods and the latitude and longitude columns were added to the dataframe:

	Name	long	lat
0	Cantonment area	-87.3446	30.6054
1	Domlur	77.6387	12.961
2	Indiranagar	77.6408	12.9784
3	Jeevanbheemanagar	77.6581	12.9642
4	Malleswaram	77.5692	13.0055

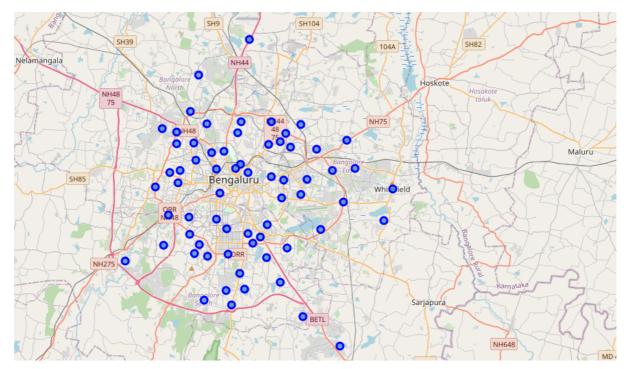
Dataframe with latitude and longitude details added

On an analysis, it turned out that even in this, a couple of longitude and latitude details were wrong. For these, further address details were added and the correct latitude and longitude details were included. For future references purposes, this dataframe was also saved as a csv file.

	Name	long	lat
0	Cantonment area	77.597800	12.993800
1	Domlur	77.638732	12.960986
2	Indiranagar	77.640836	12.978369
3	Jeevanbheemanagar	77.658076	12.964163
4	Malleswaram	77.569236	13.005511

Dataframe after correcting the latitude and longitude details

Using Folium, these neighborhoods were plotted on the map of Bangalore:



Bangalore Neighborhoods

The next step was to use Foursquare API to pull the 100 top venues within 500 metre radius of each of these neighborhoods.

(880, 7) There are 147 unique categories.												
	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category					
0	Cantonment area	12.9938	77.5978	Ujwal Bar & Restaurant	12.9923	77.5945	Indian Restaurant					
1	Cantonment area	12.9938	77.5978	Alliance Française	12.9912	77.5967	Concert Hall					
2	Cantonment area	12.9938	77.5978	Millers 46	12.9917	77.5942	Steakhouse					
3	Cantonment area	12.9938	77.5978	Jayamahal Palace Hotel	12.9968	77.5972	Indian Restaurant					
4	Cantonment area	12.9938	77.5978	Watson's	12.9942	77.5946	Pub					
5	Cantonment area	12.9938	77.5978	Jaymahal Palace Grounds	12.9953	77.5969	Music Venue					

Top venues within 500m radius of each of the neighborhoods

Next step was to generate one-hot vectors for the venues in each of these neighborhoods and generate the most common venues and get the top 10 venues.

	Neighborhood	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
0	Anjanapura	ATM	Women's Store	Electronics Store	Food	Flower Shop	Flea Market	Fast Food Restaurant	Farmers Market	Falafel Restaurant	Event Space
1	Arekere	Sporting Goods Shop	Indian Restaurant	Pizza Place	Athletics & Sports	Breakfast Spot	Liquor Store	Department Store	Business Service	Fast Food Restaurant	Bar
2	BTM Layout	Indian Restaurant	Ice Cream Shop	Bakery	Chinese Restaurant	Snack Place	Pizza Place	Fast Food Restaurant	Sandwich Place	Coffee Shop	Vegetarian / Vegan Restaurant
3	Banashankari	Clothing Store	Café	Breakfast Spot	North Indian Restaurant	Fast Food Restaurant	Indian Restaurant	Fried Chicken Joint	Pizza Place	Men's Store	Sporting Goods Shop
4	Banaswadi	Indian Restaurant	Vegetarian / Vegan Restaurant	Pharmacy	Kerala Restaurant	Women's Store	Electronics Store	Flea Market	Fast Food Restaurant	Farmers Market	Falafel Restaurant
5	Basavanagudi	Indian Restaurant	Fast Food Restaurant	Park	Sandwich Place	Food	Hookah Bar	Pharmacy	Café	Plaza	Metro Station
6	Basaveshwaranagar	Ice Cream Shop	Fast Food Restaurant	Indian Restaurant	Juice Bar	Burger Joint	Miscellaneous Shop	Gastropub	Sporting Goods Shop	Pizza Place	Gym
7	Begur	Supermarket	Mobile Phone Shop	Women's Store	Flower Shop	Flea Market	Fast Food Restaurant	Farmers Market	Falafel Restaurant	Event Space	Electronics Store
8	Bellandur	Hotel Bar	Soccer Field	Fast Food Restaurant	Shopping Mall	Indian Restaurant	Hookah Bar	Eastern European Restaurant	Hotel	Farmers Market	Falafel Restaurant
9	Bidadi	Indian Restaurant	Breakfast Spot	Women's Store	Electronics Store	Flower Shop	Flea Market	Fast Food Restaurant	Farmers Market	Falafel Restaurant	Event Space
10	Bommanahalli	Indian Restaurant	Department Store	Women's Store	Electronics Store	Flower Shop	Flea Market	Fast Food Restaurant	Farmers Market	Falafel Restaurant	Event Space

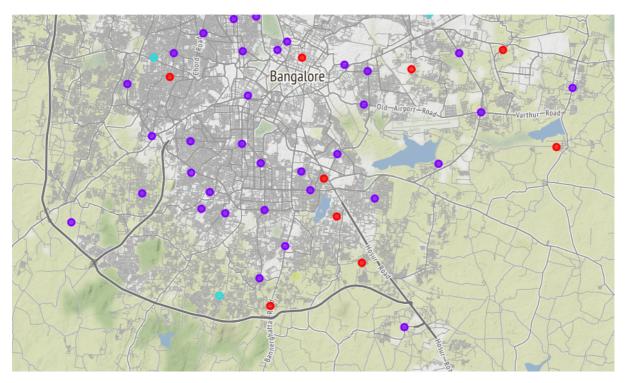
Top ten venues in each neighborhood

Next step is to segment these neighborhoods based on similarities in their common venues. For this, I have used K-means clustering. I have set the clusters as four.

	Neighborhood	long	lat	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	
0	Cantonment area	77.597800	12.993800	1	Indian Restaurant	Bakery	Music Venue	Gym / Fitness Center	Concert Hall	Tea Room	Pub	
1	Domlur	77.638732	12.960986	1	Indian Restaurant	Café	BBQ Joint	Rajasthani Restaurant	Chinese Restaurant	Asian Restaurant	Sandwich Place	F
2	Indiranagar	77.640836	12.978369	1	Indian Restaurant	Café	Pub	Bakery	Vegetarian / Vegan Restaurant	Fast Food Restaurant	Chinese Restaurant	
3	Malleswaram	77.569236	13.005511	1	Ice Cream Shop	Coffee Shop	Chinese Restaurant	Fast Food Restaurant	South Indian Restaurant	Donut Shop	Café	
4	Pete area	77.577000	12.965600	1	Indian Restaurant	Plaza	Market	Flower Shop	Historic Site	Food Truck	Diner	
5	Sadashivanagar	77.581285	13.006818	1	Coffee Shop	Department Store	Ice Cream Shop	Café	Indian Restaurant	Seafood Restaurant	Cafeteria	
6	Seshadripuram	77.574044	12.988905	1	Ice Cream Shop	Indian Restaurant	Clothing Store	Fast Food Restaurant	Bakery	Electronics Store	Donut Shop	
7	Shivajinagar	77.605693	12.985650	0	Indian Restaurant	Clothing Store	Donut Shop	Fast Food Restaurant	Tea Room	Market	South Indian Restaurant	
8	Ulsoor	77.628415	12.981700	1	Café	Burger Joint	Bakery	Event Space	Flower Shop	Flea Market	Fast Food Restaurant	
9	Vasanth Nagar	77.592795	12.989619	1	Indian Restaurant	Coffee Shop	Chinese Restaurant	Italian Restaurant	Dessert Shop	Pizza Place	Sandwich Place	
10	Bellandur	77.678404	4 12.930428 1		Hotel Bar	Soccer Field	Fast Food Restaurant	Shopping Mall	Indian Restaurant	Hookah Bar	Eastern European Restaurant	
11	CV Raman Nagar	77.664184	12.979325	0	Indian Restaurant	Coffee Shop	Food Court	Café	Department Store	Dessert Shop	Dim Sum Restaurant	
12	Hoodi	77.712657	12.989555	0	Indian Restaurant	Bus Station	Women's Store	Electronics Store	Flower Shop	Flea Market	Fast Food Restaurant	

Each neighborhoods with Cluster labels

Finally, these clusters are plotted on the map to get a spatial information about the relative locations of these similar neighborhoods.



Neighborhoods clustered

Results

First Cluster

	Neighborhood	long	lat	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 Mos Commo Venu
7	Shivajinagar	77.605693	12.985650	0	Indian Restaurant	Clothing Store	Donut Shop	Fast Food Restaurant	Tea Room	Market	South Indian Restaurant	Event Space	Flowe Sho
11	CV Raman Nagar	77.664184	12.979325	0	Indian Restaurant	Coffee Shop	Food Court	Café	Department Store	Dessert Shop	Dim Sum Restaurant	Diner	Donu Sho
12	Hoodi	77.712657	12.989555	0	Indian Restaurant	Bus Station	Women's Store	Electronics Store	Flower Shop	Flea Market	Fast Food Restaurant	Farmers Market	Falafe Restaurar
13	Krishnarajapuram	77.704433	13.016999	0	Indian Restaurant	Bakery	Tibetan Restaurant	Women's Store	Electronics Store	Flower Shop	Flea Market	Fast Food Restaurant	Farmer Marke
18	Banaswadi	77.648194	13.010376	0	Indian Restaurant	Vegetarian / Vegan Restaurant	Pharmacy	Kerala Restaurant	Women's Store	Electronics Store	Flea Market	Fast Food Restaurant	Farmer Marke
27	Mathikere	77.563976	13.033419	0	Indian Restaurant	Fast Food Restaurant	Ice Cream Shop	Department Store	Pizza Place	Bus Station	Convenience Store	Halal Restaurant	Histori Sit
32	Bommanahalli	77.624194	12.902980	0	Indian Restaurant	Department Store	Women's Store	Electronics Store	Flower Shop	Flea Market	Fast Food Restaurant	Farmers Market	Falafe Restaurar
33	Bommasandra	77.697437	12.816730	0	Indian Restaurant	Women's Store	Electronics Store	Flower Shop	Flea Market	Fast Food Restaurant	Farmers Market	Falafel Restaurant	Ever Spac

Extract from the first cluster

Second Cluster

	Neighborhood	long	lat	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	8 C
0	Cantonment area	77.597800	12.993800	1	Indian Restaurant	Bakery	Music Venue	Gym / Fitness Center	Concert Hall	Tea Room	Pub	F
1	Domlur	77.638732	12.960986	1	Indian Restaurant	Café	BBQ Joint	Rajasthani Restaurant	Chinese Restaurant	Asian Restaurant	Sandwich Place	Pizz
2	Indiranagar	77.640836	12.978369	1	Indian Restaurant	Café	Pub	Bakery	Vegetarian / Vegan Restaurant	Fast Food Restaurant	Chinese Restaurant	Desse
3	Malleswaram	77.569236	13.005511	1	Ice Cream Shop	Coffee Shop	Chinese Restaurant	Fast Food Restaurant	South Indian Restaurant	Donut Shop	Café	Came
4	Pete area	77.577000	12.965600	1	Indian	Plaza	Market	Flower	Historic	Food Truck	Diner	

Extract from the Second cluster

Third Cluster

	Neighborhood	long	lat	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
	Ramamurthy Nagar	77.673703	13.008489	2	Shoe Store	ATM	Electronics Store	Food	Flower Shop	Flea Market	Fast Food Restaurant	Farmers Market	Falafel Restaurant	Event Space
	Anjanapura	77.561594	12.861652	2	ATM	Women's Store	Electronics Store	Food	Flower Shop	Flea Market	Fast Food Restaurant	Farmers Market	Falafel Restaurant	Event Space
	Kamakshipalya	77.526883	12.985524	2	ATM	Warehouse Store	Electronics Store	Food	Flower Shop	Flea Market	Fast Food Restaurant	Farmers Market	Falafel Restaurant	Event Space
	Doddaballapura / i\nHoskote\n\nMagad	77.536391	13.295714	2	ATM	Women's Store	Electronics Store	Food	Flower Shop	Flea Market	Fast Food Restaurant	Farmers Market	Falafel Restaurant	Event Space

Extract from the Third Cluster

Discussion

Looking at the clustered data, the most common venue across the first and second cluster is Indian restaurants. Therefore, while this may not be helpful in distinguishing between the first and the second neighbourhood, if you are looking to stay in a neigborhood with enough number of eating out places, then neighborhoods in the third cluster may not be your ideal choice.

Between the first cluster and the second cluster, if you are looking a place with more number of neighborhood cafes, pubs and fitness centres, any of the neighborhoods in the second cluster will be your ideal choice.

Conclusion

This project looked at the qualitative factors in choosing a neighborhood. Rather than looking at price and proximity to work place, this project looks at the common venues in a neighborhood and using unsupervised k-cluster machine learning algorithm, segmented the similar neighborhoods into clusters.

If you are new to the city and if you prefer a neighborhood with a lot of cafes, pubs and fitness centres, neighborhoods in the second cluster will be your ideal choice.