Company's New Branch in Bangalore

B.M.SURAJ

Project Report

a. Introduction:

i. Background:

This project is implemented based on Bangalore, its locality's boundaries, real estate prices and venues associated with each locality. So let me introduce you to the city before we move forward. Bangalore, also called Bengaluru, is the capital of Karnataka. The city is also known as the "Garden City" and the "Silicon Valley of India".

The city of Bangalore is situated in the south-eastern part of the state in the centre of the Mysore Plateau at an average elevation of 2953 ft. The city of Bangalore is the 3rd most populated city in India and has a population of 8,425,970.

Bangalore is rapidly becoming a major financial and cultural hub and is ranked as the best city to live among all the five metro cities of India. The quality of life and climate in Bangalore, both are much better than the rest of the Indian metropolitan cities.

Bangalore with a population of over eight million is growing fast as an employment hub. It is home to the most reputed Indian software companies such as Infosys, Wipro, Polaris, and HCL along with the development facilities of fortune 500 companies such as Apple,

Microsoft, IBM, Oracle, SAP etc. Having a branch office in Bangalore has become a symbol of reputation for various MNCs. IT Park alone houses many Indian and foreign IT companies. It is truly an IT and BPO hub at present.

For more than a decade many Information Technology enabled Services (I.T.E.S) and Business Process Outsourcing (BPO) companies have established their offices in Bangalore. This has formed an active demand for commercial and residential properties in Bangalore. In the commercial segment, office spaces are in demand and the retail segment is also strong. The Residential segment, developers are building unique properties to attract consumer. The Real Estate in Bangalore is also witnessing new trends and is expanding to new centres. Thus, offering lots of opportunities in the real estate sector.

ii. Business Problem:

This project is for a company which is already established where there headquarters is located. Now they want to expand their business by constructing a new unit in Bangalore, INDIA for which they need to find a plot which is reasonable but also surrounded by venues which will attract more talented employees and keeps the productivity high.

Finding a plot in a populated city like Bangalore is not an easy job. Locality average prices, its basic features needs to be considered. This project will help them in zeroing out the probable locations and then they may get into further details based on their specific needs.

This project can also be used in future for similar other companies who will be interested in opening new branches in Bangalore and are searching for plots.

b. Data description:

In this project we have used data from various sources for extracting different information.

- Firstly, we extracted data for area names in Bangalore, INDIA with its average price per sqft in rupees. (https://www.makaan.com/price-trends/property-rates-for-buy-in-bangalore)
 - We used plot section from this site as our project is about constructing new company.
- Secondly, we used foursquare api for extracting neighbourhood venues.
- And finally we used a json file for boundaries of each area in Bangalore from here-(https://raw.githubusercontent.com/datameet/PincodeBoundary/master/Bangalore/boundary.geojson).

c. Methodology:

i. Data extraction for average price per sqft in rupees for plots in each locality of Bangalore.

We used a website mentioned above to extract data about locality names and its average price per sqft of plots.

We also had to make necessary changes in certain rows for entered data in it and also to add some more meaningful locality names and its average plot price so we used excel for extraction, made necessary changes, saved it in csv format and imported it in python.

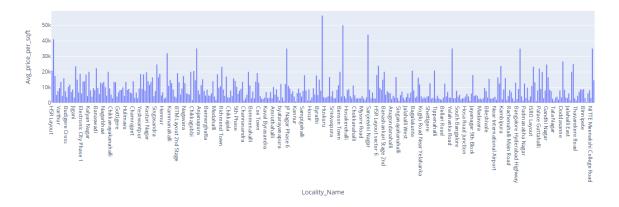
ii. Extracting Longitudes and Latitudes for localities mentioned in the dataframe. For this we have used geopy and Nominatim.

In this part we extracted Longitudes and Latitudes with the help of geopy and nominatim according to the names present in "Locality_Name". We also dropped null values and created a data frame shown below.

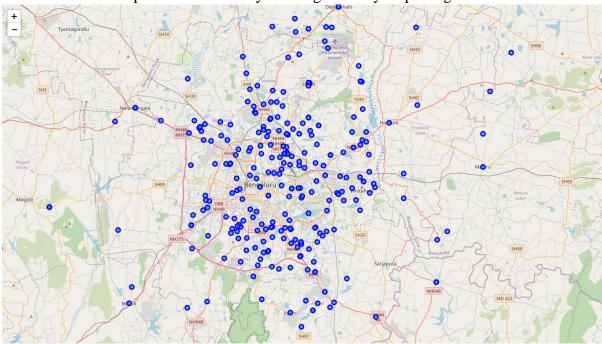
	Locality_Name	Avg_price_per_sqft	Longitude	Latitude	
0	HSR Layout	20833.33	77.638862	12.911623	
1	Koramangala	40959.90	77.751926	13.292399	
2	Whitefield	17408.84	-71.611858	44.373058	
3	Krishnarajapura	5126.25	76.805633	12.032522	
4	Begur	7571.43	3.208835	41.954192	
5	Varthur	9534.09	77.746994	12.940615	
6	Devanahalli	13611.17	77.713438	13.248350	
7	Chandapura	4473.16	77.433391	17.443639	
8	Mahadevapura	15922.58	77.692416	12.993498	
9	Whitefield Hope Farm Junction	7500.00	77.752044	12.983974	
10	Budigere Cross	3723.50	77.750307	13.046410	
11	Bellandur	10391.92	77.666761	12.935772	
12	Jakkur	11820.00	77.606894	13.078474	
13	Marathahalli	6999.50	77.698416	12.955257	
14	Yelahanka	8638.38	77.596345	13.100698	
15	Jigani	3565.19	77.638093	12.785253	
16	Indira Nagar	23531.75	77.640467	12.973291	
17	Murugeshpalya	15000.00	77.655593	12.958948	
18	Horamavu	6692.31	77.660151	13.027331	
19	BTM Layout	18750.00	77.610282	12.915177	

iii. Plotting each column data on map and bar plot for average price in INR with respect to locality name.

Here, we used plotly for plotting barplot for each locality and its average price in INR. Average price per Square feet for plots in INR for each area



And then we plotted each locality on Bangalore city map using Folium.

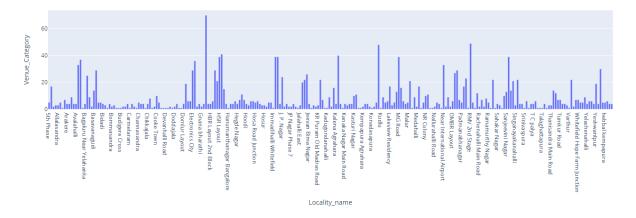


iv. Collection of venues for each locality using Foursquare and creating a table for venue categories.

In this section we have used foursquare api for extracting venues for each neighbourhood and then used one hot encoding for each venue based on locality which we later used for modelling.

We also used value counts of each venue according to locality and placed it in a data frame to create a bar plot for venue category shown below:

Venue category counts for each locality of Bengaluru

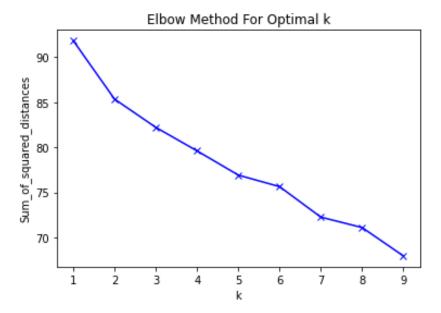


We also placed the top ten venues of each locality, converted into columns and placed it as a table with locality name column.

10th Most Common Venue	9th Most Common Venue	8th Most Common Venue	7th Most Common Venue	6th Most Common Venue	5th Most Common Venue	4th Most Common Venue	3rd Most Common Venue	2nd Most Common Venue	1st Most Common Venue	Locality_name	
Fabric Shop	Farmers Market	Fast Food Restaurant	Field	Fish Market	Doner Restaurant	Women's Store	Bus Station	Snack Place	Indian Restaurant	5th Phase	0
Diner	Tea Room	South Indian Restaurant	Coffee Shop	Udupi Restaurant	Breakfast Spot	Salon / Barbershop	Bakery	Department Store	Indian Restaurant	AECS Layout	1
Fabric Shop	Farmers Market	Fast Food Restaurant	Field	Fish Market	Fishing Store	Flower Shop	Women's Store	Kerala Restaurant	Lake	AECS Layout A Block Singasandra	2
Fabric Shop	Farmers Market	Fast Food Restaurant	Field	Fish Market	Fishing Store	Doner Restaurant	Indie Movie Theater	Bakery	Business Service	Abbigere	3
Event Service	Fabric Shop	Farmers Market	Fast Food Restaurant	Field	Fish Market	Fishing Store	Flea Market	Doner Restaurant	ATM	Allalasandra	4

v. Kmeans Modelling.

We used one hot coded data to use it for kmeans modelling for which first we found the best k using elbow curve method and chose k as 5 and modelled.

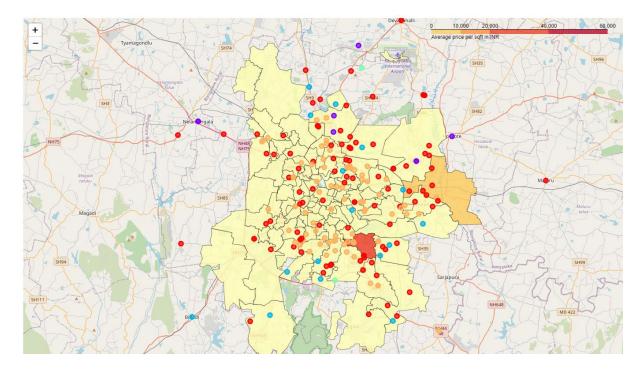


Then we merged all the data for including columns with Locality_Name, average price, Longitudes, latitudes and top ten venues associated.

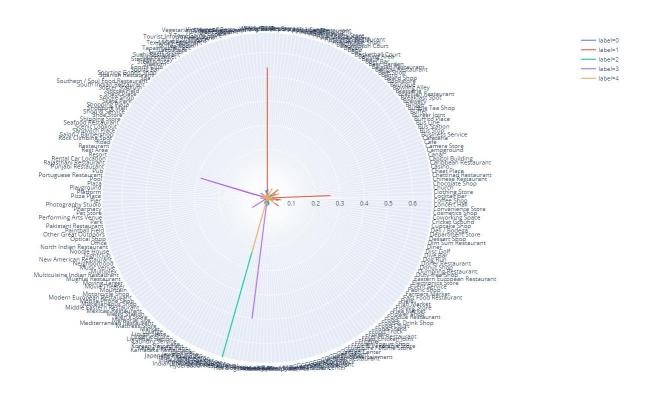
	Locality_Name	Avg_price_per_sqft	Longitude	Latitude	Cluster_Labels	1st Most Common Venue	2nd Most Common Venue	Ommon Common Venue	4th Most Common Venue	Sth Most Common Venue	6th Most Common Venue	7th Most Common Venue	Ommon Venue	9th Most Common Venue
0	HSR Layout	20833.33	77.638862	12.911623	4.0	Indian Restaurant	Ice Cream Shop	Café	Snack Place	Liquor Store	Farmers Market	Seafood Restaurant	Food Court	Coffee Shop
1	Whitefield	17408.84	-71.611858	44.373058	0.0	Pizza Place	Sandwich Place	Donut Shop	Gun Shop	Grocery Store	Convenience Store	Women's Store	Electronics Store	Dumpling Restaurant
2	Begur	7571.43	3.208835	41.954192	0.0	Mediterranean Restaurant	Hotel	Tapas Restaurant	Spanish Restaurant	Cocktail Bar	Pub	Athletics & Sports	Bar	Bakery
3	Varthur	9534.09	77.746994	12.940615	2.0	Indian Restaurant	Women's Store	Dog Run	Fishing Store	Fish Market	Field	Fast Food Restaurant	Farmers Market	Fabric Shop
4	Devanahalli	13611.17	77.713438	13.248350	0.0	Campground	Women's Store	Dog Run	Fishing Store	Fish Market	Field	Fast Food Restaurant	Farmers Market	Fabric Shop
5	Mahadevapura	15922.58	77.692416	12.993498	0.0	Movie Theater	Coffee Shop	Women's Store	Bar	Italian Restaurant	French Restaurant	Fast Food Restaurant	Multiplex	Noodle House
6	Whitefield Hope Farm Junction	7500.00	77.752044	12.983974	4.0	Intersection	Eastern European Restaurant	Vegetarian / Vegan Restaurant	Indian Restaurant	Breakfast Spot	Restaurant	Department Store	Doner Restaurant	Donut Shop
7	Budigere Cross	3723.50	77.750307	13.046410	0.0	Bus Stop	Women's Store	Fishing Store	Fish Market	Field	Fast Food Restaurant	Farmers Market	Fabric Shop	Event Service
8	Marathahalli	6999.50	77.698416	12.955257	4.0	Indian Restaurant	Clothing Store	Asian Restaurant	Shoe Store	South Indian Restaurant	Kerala Restaurant	Sporting Goods Shop	Restaurant	Bakery
9	Yelahanka	8638.38	77.596345	13.100698	4.0	ATM	Train Station	Indian Restaurant	Mobile Phone Shop	Food Truck	Farmers Market	Fabric Shop	Fast Food Restaurant	Doner Restaurant

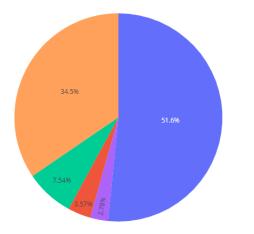
vi. Plotting the clusters on map using json data for boundaries of locality in Bangalore and choropleth for Average price per sqft of plots.

We finally used a json data of Bangalore consisting of boundary coordinates of each locality and used it plotting heat map of the city based on average price per sqft and used clusters as markers on top it.



Lastly, we used Line polar for visualizing which cluster has the maximum frequency per venue and also used a pie chart to check the number of localities in each cluster.





0 4 2 1 3

d. Result discussion:

Finally we will conclude with results in this section:

We need to consider two facts here:

- Average price per sqft in INR for each locality
- Number of venues available at each locality

With respect to above observations we may say that prices are almost equal around the city other than couple of locality where prices are at peak.

We can also say that cluster 4 considers the highest venues with help of line polar and pie plots. Cluster 4 can be seen in the maximum around south, east and central-north Bangalore.

Any of these places might be suitable for new office space considering many other established companies and start-ups are also set up here.

Please note that plot selection of different companies may depend on many other factors, so other deep research is also required based on company requirements.

e. Conclusion:

In this project we have searched data available from all possible sources available and selected the best suitable data for analysing results.

Here we have identified locality names, average price per sqft, its longitudes, latitudes, best possible general venues. Then used this data to plot on map with both heat map of average prices and marking clusters based on venues. Finally discussed about the possible results.

This project as used for a company to open their new branch office in best possible location of Bangalore, we can say that this project may also be used with similar scenarios when required.