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| **Restaurant recommendation system** |
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| 21/10/2020 | IBM Capstone Project |

Restaurant recommendation system is a machine learning model, developed to demonstrate as a capstone project to IBM through Coursera. It recommends restaurants based on users’ likes and dislikes, Boroughs, income, Neighbourhood, population etc. and their previous interest data. **Table of contents**

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1. **Introduction:**

## The Restaurant Recommendation Problem

**Problem background:**

#### Bangalore is the economic capital of the Indian state of Gujarat. It is also a World Heritage Site

#### The diversity of the cuisine available is reflective of the social and economic diversity of Bangalore. Roadside vendors, tea stalls, South Indian, North Indian, Chinese and Western fast food are all very popular in the city.

### Problem description:

#### Food being a big part of the city's culture, it is really important to have a functional and accurate reccomendation system for the people who love eating out. Following are the key questions that can be asked:

**KEY QUESTIONS:**

#### How many types of foods are available in the RESTAURANT?

#### which one is the nearest to me with a good RATING?

#### How many "similar" restaurants are available nearby ME?

#### Do the "similar" restaurants cost more? If yes, what speciality do they have that make them BETTER?

#### Expectations from this recommendation system is to get answer for the questions, and in such a way that it uncovers all the perspective of managing recommendations. It is sighted to SHOW:

#### What types of restaurants are present in a particular AREA?

#### How do different restaurants rank with respect to my PREFERENCES?

**Target audience:**

#### Target audiences for this project is not limited to a person who keeps travelling but to everyone. People could simply decide to look for a similar restaurant all the time because they are addicted to a specific category of food. People who rarely use restaurants would prefer to have the most rated restaurants nearby them and all this could be easily handed by our recommendation system. So, target for this project is basically everyone who is exploring different places or similar places.

**Success rate:**

With restaurants evolving, new food categories emerge, hybrid food starts to be more popular, we need a system that could help us access vast number of food varieties. It is impossible for a person to ask each one about their visit to a place and not everyone remembers everything. On the other hand, computers are good at remembering things, and with Machine learning at its peak, it high time technology will by our personal guidance and help us specifically based on our likes and dislikes. So, people would care about this project as their personal assistance and success rate could certainly increase with time.

1. **Data:**

**Data requirements:**

To find a solution to the questions and build a recommendation model, we need lots and lots of data. Data can answer questions which are unimaginable and non-answerable by humans because humans do not have the capability to analyze such large dataset and produce analytics to find a solution.

Let's consider the base scenario:

Suppose I want to find a restaurant, then logically, I need 3 things:

1. Its geographical coordinates (latitude and longitude) to find out where exactly it is located.

2. Population of the neighbourhood where the restaurant is located.

3. Average income of neighborhood to know how much the restaurant is worth.

Let’s take a closer look at each of these:

1. To access location of a restaurant, it’s Latitude and Longitude is to be known so that we can point at its coordinates and create a map displaying all the restaurants with its labels respectively.

2. Population of a neighborhood is very important factor in determining a restaurant's growth and amount of customers who turn up to eat. Logically, the more the population of a neighborhood, the more people will be interested to walk openly into a restaurant and less the population, less number of people frequently visit a restaurant. Also if more people visit, better the restaurant is rated because it is accessed by different people with different taste. Hence is very important factor.

3. Income of a neighborhood is also very important factor as population was. Income is directly proportional to richness of a neighborhood. If people in a neighborhood earns more than an average income, then it is very much possible that they will spend more however not always true with very less probability. So a restaurant assessment is proportional to income of a neighborhood.

**Data collection:**

1. Collecting geographical coordinates is not difficult but after googling for more than 2 days, it was not available on open source data websites such as Wikipedia, India gov website, census report websites etc. So I decided to use Google maps API to fetch latitude and longitude but google API has limited number of calls that I could make with my free account. So it would take around 15 - 20 days to fetch location of all the neighborhoods in Bangalore.

Initially I scrapped list of neighbor's using beautifulSoup4 from [wikipedia](https://en.wikipedia.org/wiki/List\_of\_neighbourhoods\_in\_Bangalore). The table headings becoming the boroughs and data becoming the neighborhoods. Bangalore has 8 boroughs and 64 neighborhoods. So i manually googled each neighborhood to find its corresponding latitude and longitude. After doing so, I produced the following data frame.



2. Population by neighborhood is again easy to find out given that it’s readily available. But in case of Bangalore, it is again not the case. i was able to find population data for few cities. [Here is the link](https://indikosh.com/dist/655489/Bangalore). Rest other neighborhood population is assumed and may be inaccurate but since this is a demonstrating project, the main idea to get the working model. The data frame for Bangalore neighborhood population looks like:



3. Income by neighborhood is again easy to find out given that it’s readily available. But in case of Bangalore, it is again not the case. i was able to find Income data for main city. [Here is the link](https://en.wikipedia.org/wiki/List\_of\_Indian\_cities\_by\_GDP\_per\_capita). Neighborhood Income is assumed and may be inaccurate but since this is a demonstrating project, the main idea to get the working model. The data frame for Bangalore neighborhood population looks like:



4. Foursquare API:

Use of foursquare is focused to fetch nearest venue locations so that we can use them to form a cluster. Foursquare API leverages the power of finding nearest venues in a radius (in my case: 500mts) and also corresponding coordinates, venue location and names. After calling, the following data frame is created:



1. **Methodology:**

**Exploratory analysis:**

Scrapping the data from different sources and then combining it to form a single-ton dataset is a difficult task. To do so, we need to explore the current state of dataset and then list up all the features needed to be fetched.

Exploring the dataset is important because it gives you initial insights and may help you to get partial idea of the answers that you are looking to find out from the data.

While exploring the dataset, I found out that Indiranagar has most number of venues while Varthur has the least.



Also while producing graph for number of cluster, I produced a graph to explore all the values for n\_clusters and then finding the best by exploring the elbow graph.



**Inferential analysis:**

Most important factors while building the recommendation system were population and income. They are the most import factor because they have a nonlinear relationship according to our dataset.

It needed to make some inferential analysis to understand this nonlinear relationship. As the amount of population increases, it does not necessarily mean that average income of a neighborhood will also increase. It is true to most of the case but also many cases differ to follow this trend. Similarly, a neighborhood with less number of people may not necessarily have less average income. It is possible to have less number of people and more income and vice versa. This can be inferred from the following graph:



1. **Result:**

The result of the recommendation system is that it produces a list of top restaurants and the most common venue item that the user can enjoy. During the runtime of the model, a simulation was done by taking ‘Whitefield’ as the neighborhood and then processed through our model so that it could recommend neighborhoods with similar characters as that of ‘Whitefield’.

The following image shows the result:



1. **Discussion:**

Since there was a nonlinear relationship between income and population, it can be concluded that we must always perform inferential approach to find relationship among different set of features. Also during clustering, similar neighborhoods must be dumped into the right cluster.

The following graph shows the clusters:



Another observation that we can make is that choosing number of clustering could produce very diverse results. Some may be over fitted or some may be under fitted. Hence analysis of number of clusters must be done. Ref. elbow\_graph in the Methodology section.

1. **Conclusion :**

The recommendation system is a system that considers factors such as users’ likes and dislikes, Boroughs, income, Neighbourhood, population etc. and makes use of Foursquare API to determine a restaurant that is most suitable for a user. It is a powerful data driven model whose efficiency and accuracy will increase with more data. It will help users in finding a restaurant that is catered to their needs.