

Capstone Project - The Battle of Neighborhoods

Introduction

New York City's demographics show that it is a large and ethnically diverse metropolis. It is the largest city in the United States with a long history of international immigration. New York City was home to nearly 8.5 million people in 2014, accounting for over 40% of the population of New York State and a slightly lower percentage of the New York metropolitan area, home to approximately 23.6 million. Over the last decade the city has been growing faster than the region. The New York region continues to be by far the leading metropolitan gateway for legal immigrants admitted into the United States.

Throughout its history, New York City has been a major point of entry for immigrants; the term "melting pot" was coined to describe densely populated immigrant neighborhoods on the Lower East Side. As many as 800 languages are spoken in New York, making it the most linguistically diverse city in the world. English remains the most widely spoken language, although there are areas in the outer boroughs in which up to 25% of people speak English as an alternate language, and/or have limited or no English language fluency. English is least spoken in neighborhoods such as Flushing, Sunset Park, and Corona.

With its diverse culture, comes diverse food items. There are many restaurants in New York City, each belonging to different categories like Chinese, Indian, French etc.

So as part of this project, we will list and visualize all major parts of New York City that has great Indian restaurants.

Data

For this project we need the following data :

- New York City data that contains list Boroughs, Neighborhoods along with their latitude and longitude.
 - Data source : https://cocl.us/new_york_dataset
 - Description : This data set contains the required information. And we will use this data set to explore various neighborhoods of New York city.
- Indian restaurants in each neighborhood of New York city.
 - Data source : Foursquare API
 - Description : By using this api we will get all the venues in each neighborhood. We can filter these venues to get only Indian restaurants.
- GeoSpace data
 - Data source : <https://data.cityofnewyork.us/City-Government/Borough-Boundaries/tqmj-j8zm>
 - Description : By using this geo space data we will get the New York Borough boundaries that will help us visualize choropleth map.

Approach

- Collect the New York City data from https://cocl.us/new_york_dataset (https://cocl.us/new_york_dataset)
- Using FourSquare API we will find all venues for each neighborhood.
- Filter out all venues that are Indian Restaurants.
- Find rating, tips and like count for each Indian Restaurants using FourSquare
-

API. Using rating for each restaurant , we will sort that data.

- Visualize the Ranking of neighborhoods using folium library(python)

Questions that can be asked using the above mentioned datasets

- What is best location in New York City for Indian Cuisine ?
- Which areas have potential Indian Restaurant Market ?
- Which all areas lack Indian Restaurants ?
- Which is the best place to stay if I prefer Indian Cuisine ?

Analysis

We will import the required libraries for python.

- pandas and numpy for handling data.
- request module for using FourSquare API.
- geopy to get coordinates of City of New York.
- folium to visualize the results on a map

In [206]:

```
import pandas as pd
import numpy as np
pd.set_option('display.max_columns', None)
pd.set_option('display.max_rows', None)
import requests
from bs4 import BeautifulSoup
import geocoder
import os
import folium # map rendering library
from geopy.geocoders import Nominatim # convert an address into latitude and longit #
Matplotlib and associated plotting modules
import matplotlib.pyplot as plt
import matplotlib.cm as cm
import matplotlib.colors as colors
%matplotlib inline

print('Libraries imported.')
```

Libraries imported.

Now we define a function to get the geocodes i.e latitude and longitude of a given location using geopy.

In [207]:

```
def geo_location(address):
    # get geo location of address
    geolocator = Nominatim(user_agent="ny_explorer")
    location = geolocator.geocode(address)
    latitude = location.latitude
    longitude = location.longitude
    return latitude, longitude
```

We define a function to interact with FourSquare API and get top 100 venues within a radius of 1000 metres for a given latitude and longitude. Below function will return us the venue id , venue name and category.

In [208]:

```
def get_venues(lat,lng):

    #set variables
    radius=1000
    LIMIT=100
    CLIENT_ID = os.environ['CLIENT_ID'] # your Foursquare ID
    CLIENT_SECRET = os.environ['CLIENT_SECRET'] # your Foursquare Secret
    VERSION = '20180605'
    # Foursquare API version

    #url to fetch data from foursquare api
    url = 'https://api.foursquare.com/v2/venues/explore?&client_id={}&client_secret={}&version={}&lat={}&lng={}&radius={}&limit={}'.format(CLIENT_ID, CLIENT_SECRET, VERSION, lat, lng, radius, LIMIT)

    # get all the data
    results = requests.get(url).json()
    venue_data=results["response"]["groups"][0]["items"]
    venue_details=[]
    for row in venue_data:
        try:
            venue_id=row['venue']['id']
            venue_name=row['venue']['name']
            venue_category=row['venue']['categories'][0]['name']
            venue_details.append([venue_id,venue_name,venue_category])
        except KeyError:
            pass

    column_names=['ID','Name','Category']
    df = pd.DataFrame(venue_details,columns=column_names)
    return df
```

Now we will define a function to get venue details like, like count , rating , tip counts for a given venue id. This will be used for ranking.

```

def get_venue_details(venue_id):
    CLIENT_ID = os.environ['CLIENT_ID'] # your Foursquare ID
    CLIENT_SECRET = os.environ['CLIENT_SECRET'] # your Foursquare Secret
    VERSION = '20180605' # Foursquare API version

    #url to fetch data from foursquare api
    url = 'https://api.foursquare.com/v2/venues/{venue_id}?&client_id={CLIENT_ID}&client_secret={CLIENT_SECRET}&v={VERSION}'

    # get all the data
    results = requests.get(url).json()
    venue_data = results['response']['venue']
    venue_details = []
    try:
        venue_id = venue_data['id']
        venue_name = venue_data['name']
        venue_likes = venue_data['likes']['count']
        venue_rating = venue_data['rating']
        venue_tips = venue_data['tips']['count']
        venue_details.append([venue_id, venue_name, venue_likes, venue_rating, venue_tips])
    except KeyError:
        pass

    column_names = ['ID', 'Name', 'Likes', 'Rating', 'Tips']
    df = pd.DataFrame(venue_details, columns=column_names)
    return df

```

Now we define a function to get the new york city data such as Boroughs, Neighborhoods along with their latitude and longitude.

```
def get_new_york_data():
    url = 'https://cocl.us/new_york_dataset'
    resp = requests.get(url).json()
    # all data is present in features label
    features = resp['features']

    # define the dataframe columns
    column_names = ['Borough', 'Neighborhood', 'Latitude', 'Longitude']
    # instantiate the dataframe
    new_york_data = pd.DataFrame(columns=column_names)

    for data in features:
        borough = data['properties']['borough']
        neighborhood_name = data['properties']['name']

        neighborhood_latlon = data['geometry']['coordinates']
        neighborhood_lat = neighborhood_latlon[1]
        neighborhood_lon = neighborhood_latlon[0]

        new_york_data = new_york_data.append({'Borough': borough,
                                              'Neighborhood': neighborhood_name, 'Latitude':
                                              neighborhood_lat, 'Longitude':
                                              neighborhood_lon}, ignore_in

    return new_york_data
```

We will call the above function to get the new york city data.

In [211]:

```
# get new york data
new_york_data = get_new_york_data()
```

In [212]:

```
new_york_data.head()
```

Out[212]:

	Borough	Neighborhood	Latitude	Longitude
0	Bronx	Wakefield	40.894705	-73.847201
1	Bronx	Co-op City	40.874294	-73.829939
2	Bronx	Eastchester	40.887556	-73.827806
3	Bronx	Fieldston	40.895437	-73.905643
4	Bronx	Riverdale	40.890834	-73.912585

In [213]:

```
new_york_data.shape
```

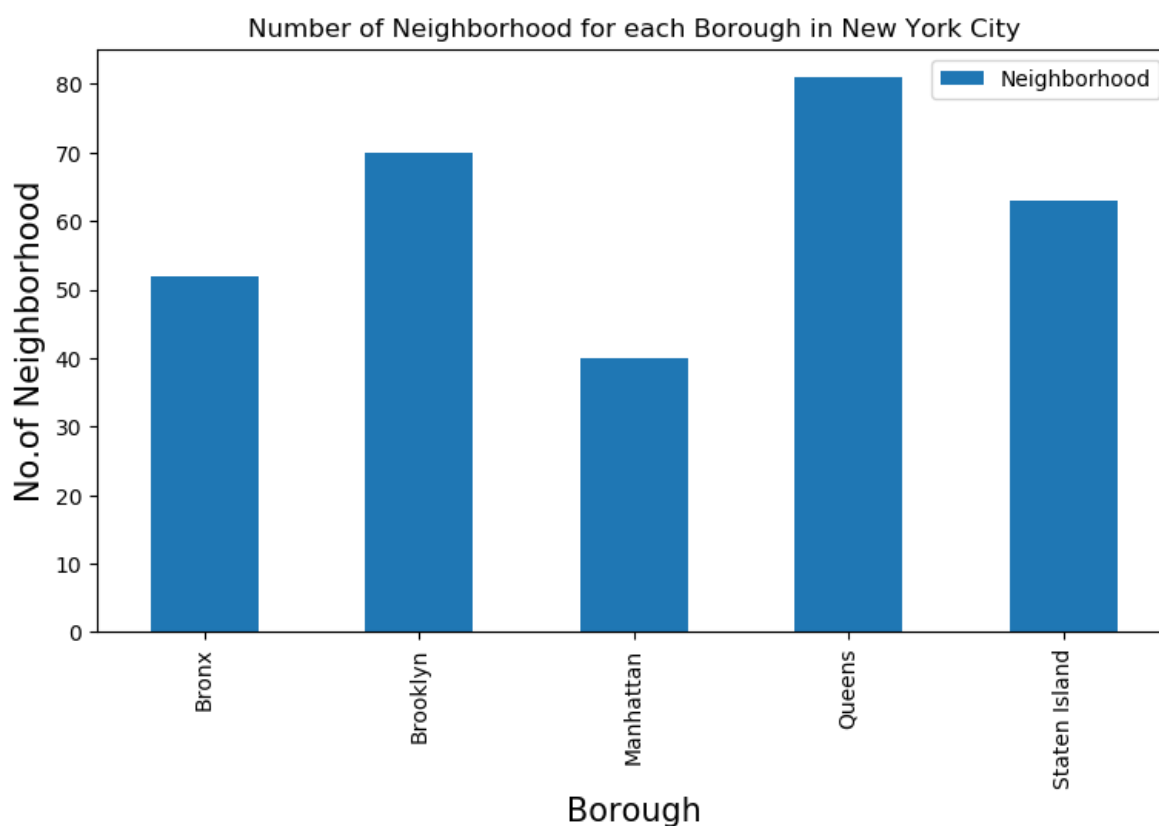
Out[213]:

(306, 4)

So there are a total of 306 different Neighborhoods in New York

In [219]:

```
plt.figure(figsize=(9,5), dpi = 100)
# title
plt.title('Number of Neighborhood for each Borough in New York City')
#On x-axis
plt.xlabel('Borough', fontsize = 15)
#On y-axis
plt.ylabel('No.of Neighborhood', fontsize=15)
#giving a bar plot
new_york_data.groupby('Borough')['Neighborhood'].count().plot(kind='bar')
#legend
plt.legend()
#displays the plot
plt.show()
```



We see that Queens has the highest number of

neighborhoods Now we will collect Indian restaurants for

each Neighborhood

```

# prepare neighborhood list that contains indian restaurants
column_names=['Borough', 'Neighborhood', 'ID', 'Name']
indian_rest_ny=pd.DataFrame(columns=column_names)
count=1
for row in new_york_data.values.tolist():
    Borough, Neighborhood, Latitude, Longitude=row
    venues = get_venues(Latitude,Longitude)
    indian_restaurants=venues[venues['Category']=='Indian Restaurant']
    print('(',count,'/',len(new_york_data),')','Indian Restaurants in '+Neighborhood for
    restaurant_detail in indian_restaurants.values.tolist():
        id, name , category=restaurant_detail
        indian_rest_ny = indian_rest_ny.append({'Borough': Borough,
                                                'Neighborhood': Neighborhood,
                                                'ID': id,
                                                'Name' : name
                                                }, ignore_index=True)

count+=1

```

```

( 288 / 306 ) Indian Restaurants in Egbertville, Staten Island:0
( 289 / 306 ) Indian Resturants in Roxbury, Queens:0
( 290 / 306 ) Indian Restaurants in Homecrest, Brooklyn:0
( 291 / 306 ) Indian Resturants in Middle Village, Queens:0
( 292 / 306 ) Indian Resturants in Prince's Bay, Staten Island:0
( 293 / 306 ) Indian Resturants in Lighthouse Hill, Staten Island:0
( 294 / 306 ) Indian Resturants in Richmond Valley, Staten Island:0
( 295 / 306 ) Indian Resturants in Malba, Queens:0
( 296 / 306 ) Indian Resturants in Highland Park, Brooklyn:0
( 297 / 306 ) Indian Resturants in Madison, Brooklyn:0
( 298 / 306 ) Indian Resturants in Bronxdale, Bronx:0 Allerton,
( 299 / 306 ) Indian Resturants in Bronx:0
( 300 / 306 ) Indian Resturants in Kingsbridge Heights, Bronx:0
( 301 / 306 ) Indian Resturants in Erasmus, Brooklyn:1
( 302 / 306 ) Indian Resturants in Hudson Yards, Manhattan:0
( 303 / 306 ) Indian Resturants in Hammels, Queens:0
( 304 / 306 ) Indian Resturants in Bayswater, Queens:0
( 305 / 306 ) Indian Resturants in Queensbridge, Queens:2
( 306 / 306 ) Indian Resturants in Fox Hills, Staten Island:1

```

Now that we have got all the indian resturants in new york city , we will analyze it

In [232]:

```
indian_rest_ny.head()
```

Out[232]:

	Borough	Neighborhood	ID	Name
0	Bronx	Riverdale	4c04544df423a593ac83d116	Cumin Indian Cuisine
1	Bronx	Kingsbridge	4c04544df423a593ac83d116	Cumin Indian Cuisine
2	Bronx	Woodlawn	4c0448d9310fc9b6bf1dc761	Curry Spot
3	Bronx	Parkchester	4c194631838020a13e78e561	Melanies Roti Bar And Grill
4	Bronx	Spuyten Duyvil	4c04544df423a593ac83d116	Cumin Indian Cuisine


```
indian_rest_ny.shape
```

```
In [233]:  
Out[233]:
```

```
(153, 4)
```

We got 153 Indian Restaurants across New York City

```
In [234]:
```

```
plt.figure(figsize=(9,5), dpi = 100)  
# title  
plt.title('Number of Indian Restaurants for each Borough in New York City')  
#On x-axis  
plt.xlabel('Borough', fontsize = 15)  
#On y-axis  
plt.ylabel('No.of Indian Restaurants', fontsize=15)  
#giving a bar plot  
indian_rest_ny.groupby('Borough')['ID'].count().plot(kind='bar') #legend  
plt.legend()  
#displays the plot  
plt.show()
```

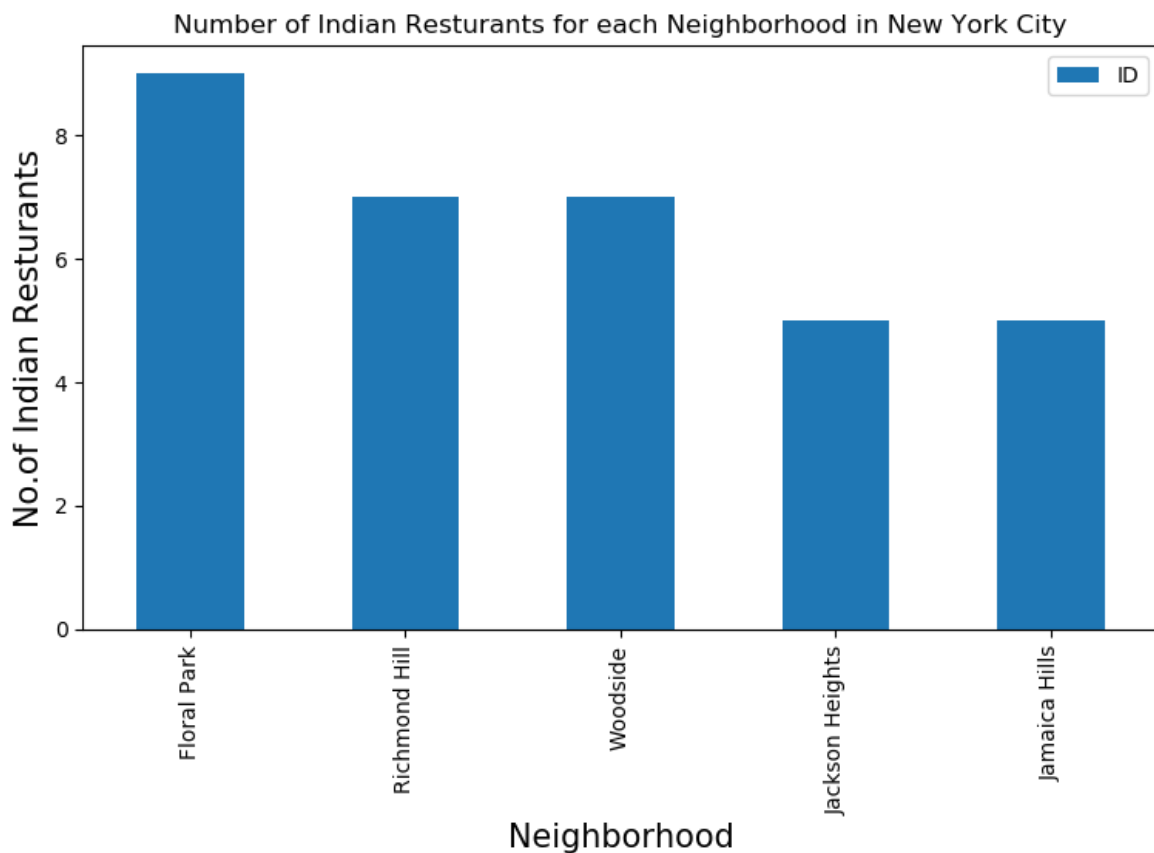


We see that Queens has the largest number of indian resturants

```

plt.figure(figsize=(9,5), dpi = 100)
# [245]:
plt.title('Number of Indian Resturants for each Neighborhood in New York City')
# On x-axis
plt.xlabel('Neighborhood', fontsize = 15)
# On y-axis
plt.ylabel('No.of Indian Resturants', fontsize=15)
#giving a bar plot
indian_rest_ny.groupby('Neighborhood')['ID'].count().nlargest(5).plot(kind='bar') #legend
plt.legend()
#displays the plot
plt.show()

```



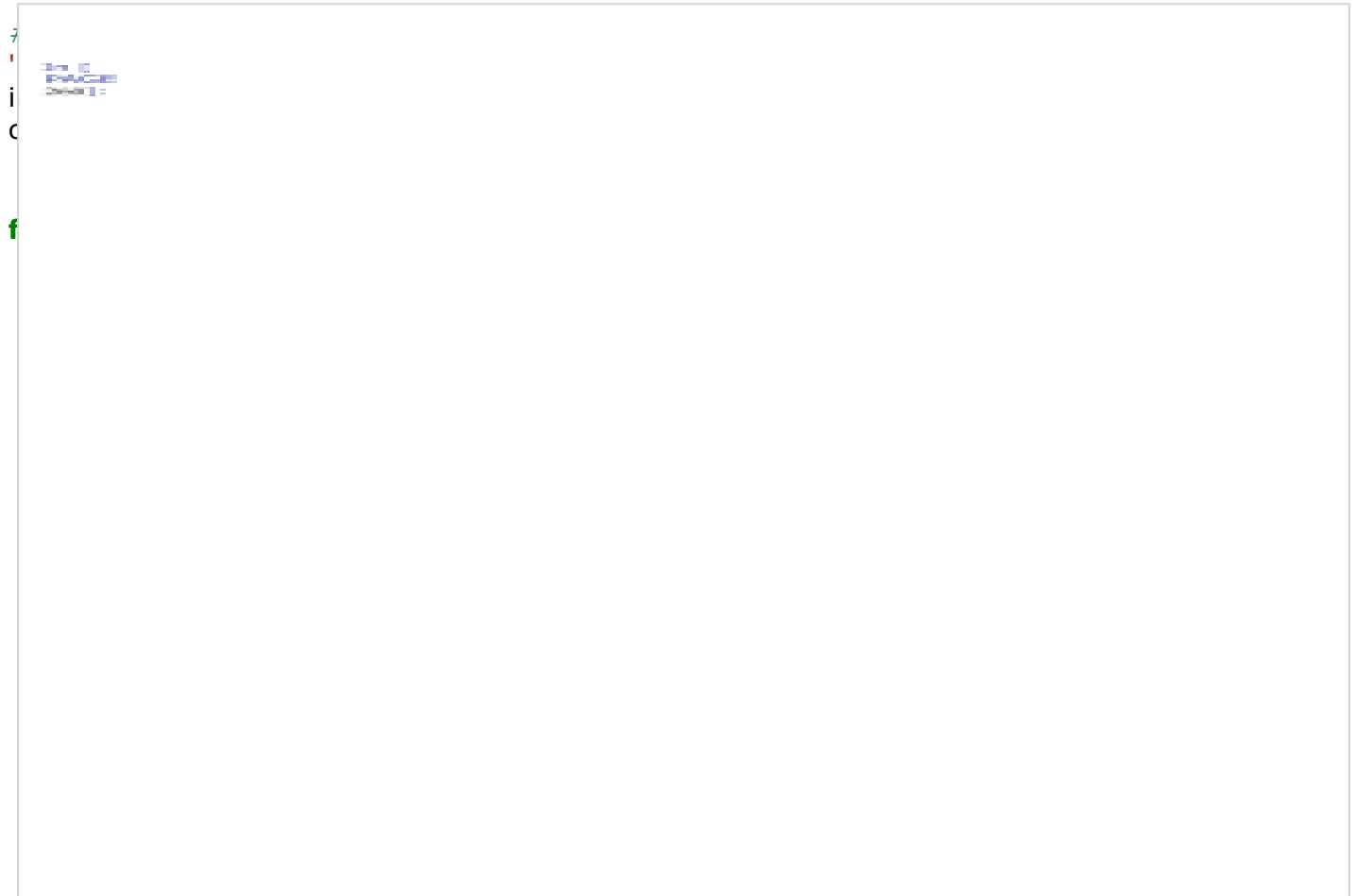
```
indian_rest_ny[indian_rest_ny['Neighborhood']=='Floral Park']
```

Out[238]:

	Borough	Neighborhood	ID	Name
103	Queens	Floral Park	527ffc0811d2d329d5e49abd	Jackson Diner
104	Queens	Floral Park	4b647b56f964a520c4b62ae3	Usha Foods & Usha Sweets
105	Q			
106	Q			
107	Q			
108	Q			
109	Q			
110	Q			
111	Q			

So Floral

Now we



	ID	Name	Likes	Rating	Tips
0	5b931ea69d7468002c3b1382	Adda	71	9.2	20

(149 / 153) processed

	ID	Name	Likes	Rating	Tips
0	564d283d498e6e851df79d87	Great Indian Curry	3	6.7	2

(150 / 153) processed

	ID	Name	Likes	Rating	Tips
0	4b1b341bf964a5208af923e3	Five Star Banquet	29	7.4	31

(151 / 153) processed

	ID	Name	Likes	Rating	Ti
ps					
0	50a287a7e4b0033f830f06db	Raj's Indian Kitchen	21	7.2	

9
(152 / 153) processed

Empty DataFrame

Columns: [ID, Name, Likes, Rating, Tips]

Index: []

No data available for id= 4b65f2e3f964a5206e0a2be3 (153 / 153) processed

```
indian_rest_stats_ny.head()
```

Out[261]:

Name Likes Rating Tips				
Cumin Indian Cuisine	13	6.6	9	
Cumin Indian Cuisine	13	6.6	9	
Curry Spot	4	7.7	10	
Melanies Roti Bar And Grill	3	6.1	2	
Cumin Indian Cuisine	13	6.6	9	

```
indian_rest_stats_ny.shape
```

Out[265]:

(153, 7)

In [266]:

```
indian_rest_ny.shape
```

Out[266]:

(153, 4)

So we got data for all restaurants. Now let's save this data to a csv sheet. In case we by mistake modify it. As the number of calls to get details for venue are premium call and have limit of 500 per day, we will refer to saved data sheet csv if required.

In [267]:

```
indian_rest_stats_ny.to_csv('indian_rest_stats_ny.csv', index=False)
```

Let's verify the data from saved csv file.

In [268]:

```
indian_rest_stats_ny_csv=pd.read_csv('indian_rest_stats_ny.csv')
```

In [269]:

```
indian_rest_stats_ny_csv.shape
```

Out[269]:

(153, 7)

```
indian_rest_stats_ny_csv.head()
```

In [270]:
Out[270]:

Name Likes Rating Tips				
Cumin Indian Cuisine	13	6.6	9	
Cumin Indian Cuisine	13	6.6	9	
Curry Spot	4	7.7	10	
Melanies Roti Bar And Grill	3	6.1	2	
Cumin Indian Cuisine	13	6.6	9	

```
indian_rest_stats_ny.info()
```

```
<class  
'pandas.core.frame.DataFrame  
> RangeIndex: 153 entries,  
0 to 152 Data columns (total  
7 columns): Borough      153  
non-null object Neighborhood  
153  
non-null object ID 153  
non-null object  
Name      153  
non-null object  
Likes      153  
non-null object  
Rating      153  
non-null float64  
Tips      153  
non-null object dtypes:  
float64(1), object(6)  
memory usage: 8.4+ KB
```

We see that values like Likes, Tips are strig values. We would need to convert them into float for further analysis

In [279]:

```
indian_rest_stats_ny['Likes']=indian_rest_stats_ny['Likes'].astype('float64')
```

In [280]:

```
indian_rest_stats_ny['Tips']=indian_rest_stats_ny['Tips'].astype('float64')
```

```
indian_rest_stats_ny.info()
```

```
In [283]:  
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 153 entries, 0 to 152 Data  
columns (total 7 columns): Borough    153  
non-null object Neighborhood    153  
non-null object ID    153 non-null object  
Name    153 non-null object  
Likes    153 non-null float64  
Rating    153 non-null float64  
Tips    153 non-null float64  
dtypes: float64(3), object(4)  
memory usage: 8.4+ KB
```

Now the data types looks correct

In [286]:

```
# Resturant with maximum Likes  
indian_rest_stats_ny.iloc[indian_rest_stats_ny['Likes'].idxmax()]
```

Out[286]:

Borough	Manhattan
Neighborhood	Midtown
ID	49d91c12f964a520015e1fe3
Name	The Kati Roll Company
Likes	819
Rating	9
Tips	257

Name: 43, dtype: object

In [287]:

```
# Resturant with maximum Rating  
indian_rest_stats_ny.iloc[indian_rest_stats_ny['Rating'].idxmax()]
```

Out[287]:

Borough	Manhattan
Neighborhood	Tribeca
ID	4bbb9dbded7776b0e1ad3e51
Name	Tamarind TriBeCa
Likes	566
Rating	9.2
Tips	141

Name: 45, dtype: object


```
# Resturant with maximum Tips
indian_rest_stats_ny.iloc[indian_rest_stats_ny['Tips'].idxmax()]
```

Out[288]:

```
Borough          Manhattan
Neighborhood      Midtown
ID              49d91c12f964a520015e1fe3
Name             The Kati Roll Company
Likes            819
Rating           9
Tips            257
Name: 43, dtype: object
```

Now lets visualize neighborhood with maximum average rating of resturants

In [374]:

```
ny_neighborhood_stats=indian_rest_stats_ny.groupby('Neighborhood',as_index=False).m
ny_neighborhood_stats.columns=['Neighborhood','Average Rating']
```

In [375]:

```
ny_neighborhood_stats.sort_values(['Average Rating'],ascending=False).head(10)
```

Out[375]:

	Neighborhood	Average Rating
0	Astoria	9.200000
71	Sunnyside	9.200000
75	Tribeca	9.200000
5	Blissville	9.200000
11	Civic Center	9.200000
47	Midtown	9.000000
48	Midtown South	9.000000
30	Gramercy	8.866667
66	Roosevelt Island	8.700000
53	North Side	8.700000

Above are the top neighborhoods with top average rating of Indian resturants

In [376]:

```
ny_borough_stats=indian_rest_stats_ny.groupby('Borough',as_index=False).mean()['Bo
ny_borough_stats.columns=['Borough','Average Rating']
```

```
ny_borough_stats.sort_values(['Average Rating'],ascending=False).head()
```

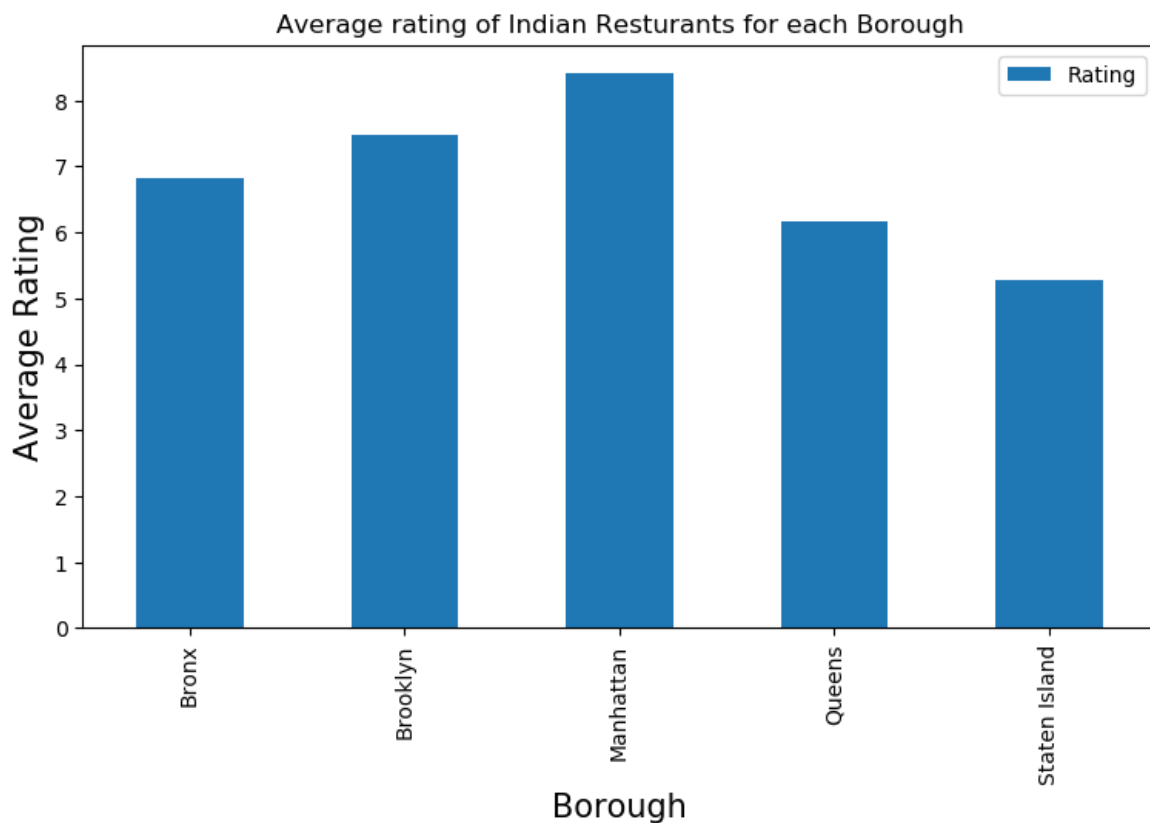
Out[377]:

	Borough	Average Rating
2	Manhattan	8.414706
1	Brooklyn	7.478571
0	Bronx	6.812500
3	Queens	6.155844
4	Staten Island	5.266667

Similarly these are the average rating of Indian Resturants for each Borough

Lets visualize it

```
plt.figure(figsize=(9,5), dpi = 100)
# In [265]:
plt.title('Average rating of Indian Resturants for each Borough')
# On x-axis
plt.xlabel('Borough', fontsize = 15)
# On y-axis
plt.ylabel('Average Rating', fontsize=15)
# giving a bar plot
indian_rest_stats_ny.groupby('Borough').mean()['Rating'].plot(kind='bar') #legend
plt.legend()
# displays the plot
plt.show()
```



	Borough	Neighborhood	ID
0	Bronx	Riverdale	4c04544df423a593ac83d116
1	Bronx	Kingsbridge	4c04544df423a593ac83d116
2	Bronx	Woodlawn	4c0448d9310fc9b6bf1dc761
3	Bronx	Parkchester	4c194631838020a13e78e561
4	Bronx	Spuyten Duyvil	4c04544df423a593ac83d116

In [265]:

	Borough	Neighborhood	ID
0	Bronx	Riverdale	4c04544df423a593ac83d116
1	Bronx	Kingsbridge	4c04544df423a593ac83d116

```
2    Bronx    Woodlawn    4c0448d9310fc9b6bf1dc761
3    Bronx    Parkchester  4c194631838020a13e78e561
4    Bronx    Spuyten Duyvil  4c04544df423a593ac83d116
```

In [277]:

We will consider all the neighborhoods with average rating greater or equal 9.0 to visualize on map

In [472]:

```
ny_neighborhood_stats=ny_neighborhood_stats[ny_neighborhood_stats['Average Rating']
```

In [473]:

Out[473]:

	Borough_x	Neighborhood	Latitude_x	Longitude_x	Average	Rating	Label Borough_y La	
0	Queens	Astoria	40.768509	-73.915654	9.2	Queens(9.2)	Queens	40
1	Queens	Blissville	40.737251	-73.932442	9.2	Queens(9.2)	Manhattan	40
2	Manhattan	Civic Center	40.715229	-74.005415	9.2	Manhattan(9.2)	Manhattan	40
3	Manhattan	Midtown	40.754691	-73.981669	9.0	Manhattan(9.0)	Midtown	
4	Manhattan	Midtown South	40.748510	-73.988713	9.0	South, Manhattan(9.0)	Manhattan	40
5	Queens	Sunnyside	40.740176	-73.926916	9.2	Queens(9.2)	Queens	40
6	Queens	Sunnyside	40.740176	-73.926916	9.2	Queens(9.2)	Staten Island	40
7		Staten Island				Sunnyside, Staten Island(9.2)	Queens	40
8		Staten Island				Sunnyside, Staten Island(9.2)	Staten Island	40
9	Manhattan					Tribeca, Manhattan(9.2)	Manhattan	40

We will join this dataset to original new york data to get lonitude and latitude

In [474]:

```
ny_neighborhood_stats=pd.merge(ny_neighborhood_stats,new_york_data, on='Neighborhood')
```

In [475]:

```
ny_neighborhood_stats=ny_neighborhood_stats[['Borough' 'Neighborhood' 'Latitude' 'Longitude' 'Rating' 'Label Borough_y La']]
```

In [473]:

Out[473]:
1:

In [476]:

	gh	Neighborhood	Latitude	Longitude	Average Rating
0	Queens	Astoria	40.768509	-73.915654	9.2
1	Queens	Blissville	40.737251	-73.932442	9.2
2	Manhattan	Civic Center	40.715229	-74.005415	9.2
3	Manhattan	Midtown	40.754691	-73.981669	9.0
4	Manhattan	Midtown South	40.748510	-73.988713	9.0
5	Queens	Sunnyside	40.740176	-73.926916	9.2
6	Staten Island	Sunnyside	40.612760	-74.097126	9.2
7	Queens	Sunnyside	40.740176	-73.926916	9.2
8	Staten Island	Sunnyside	40.612760	-74.097126	9.2
9	Queens	Sunnyside	40.740176	-73.926916	9.2
10	Staten Island	Sunnyside	40.612760	-74.097126	9.2
11	Queens	Sunnyside	40.740176	-73.926916	9.2
12	Staten Island	Sunnyside	40.612760	-74.097126	9.2
13	Manhattan	Tribeca	40.721522	-74.010683	9.2

Now we will show this data on a map

In [477]:

```
# create map and display it
ny_map = folium.Map(location=geo_location('New York'), zoom_start=12)
```

In [478]:

```
# instantiate a feature group for the incidents in the dataframe
incidents = folium.map.FeatureGroup()

# loop through the 100 crimes and add each to the incidents feature group
for lat, lng, in ny_neighborhood_stats[['Latitude', 'Longitude']].values:
    incidents.add_child(
        folium.CircleMarker(
            [lat, lng],
            radius=10, # define how big you want the circle markers to be
            color='yellow', fill=True,
            fill_color='blue',
            fill_opacity=0.6
        )
    )
```

Lets add a new field to dataframe for labeling purpose

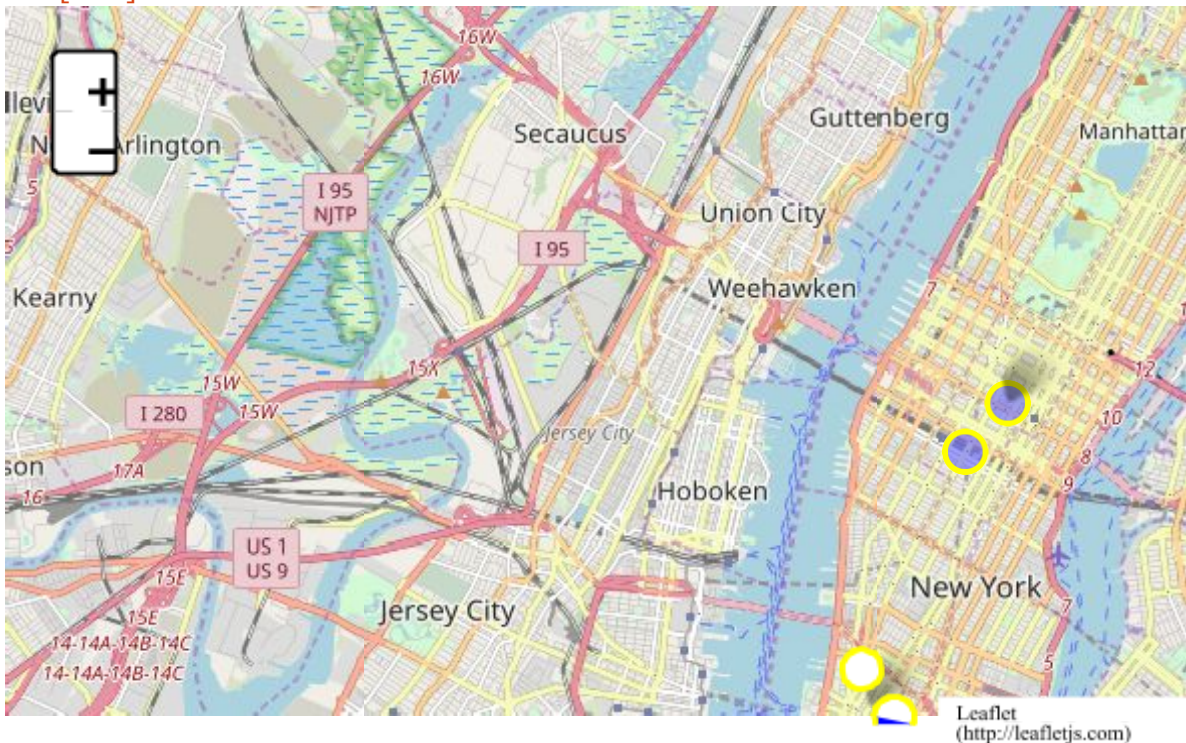
```
ny_neighborhood_stats['Label']=ny_neighborhood_stats['Neighborhood']+', '+ny_neighb
```

```
In [479]:
```

In [480]:

```
# add pop-up text to each marker on the map
for lat, lng, label in ny_neighborhood_stats[['Latitude','Longitude','Label']].valu
    folium.Marker([lat, lng], popup=label).add_to(ny_map)
# add incidents to map
ny_map.add_child(incidents)
```

Out[480]:



Now that we have visualized the Neighborhoods.
Lets Visualize Boroughs based on average Rating

In [482]:

```
ny_map = folium.Map(location=geo_location('New York'), zoom_start=12) ny_geo = r'Borough Boundaries.geojson'

ny_map.choropleth(
    geo_data=ny_geo,
    data=ny_borough_stats,
    columns=['Borough', 'Average Rating'],
    key_on='feature.properties.boro_name',
    fill_color='YlOrRd',
    fill_opacity=0.7, line_opacity=0.2,
    legend_name='Average Rating'
)

# display map
# as this is huge map data , we will save it to a file
ny_map.save('borough_rating.html')
```


The saved image can be downloaded at :

https://github.com/SaileshShocker/Coursera_Capstone/blob/master/borough_rating.html

Conclusion

- Astoria(Queens), Blissville(Queens), Civic Center(Manhattan) are some of the best neighborhoods for indian cuisine.
- Manhattan have potential Indian Resturant Market/
- Staten Island ranks last in average rating of Indian Resturants.
- Manhattan is the best place to stay if you prefer Indian Cuisine.

Limitations

- The ranking is purely on basis of rating of resturants
- The accuracy of data depends purely depends on the data provided by FourSquare

In []: