Capstone Project - Whereto? Phase I

IBM Datascience Professional Certificate Course

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Introduction: Business Problem

The purpose of this project is to find the ideal location for a fusion restaurant in Manhattan NY. The project will find a location which is not too crowded with restaurants of that type. This project will identify a few suitable neighborhoods to setup shop. This project is executed for investors who are looking to invest money for this type of venture.

Data

Following items will factor into the decision making:

- · number of existing restaurants in the neighborhood
- · number of fusion restuarants and their proximity

Below data sources are used for this:

- neighborhood clustering and segmentation data for Manhattan (pre-exising local file mnhttn_nghbrhds.tsv)
- Foursquare API will be used to determine type and number of restaurants in each neighborhood

Neighborhoods

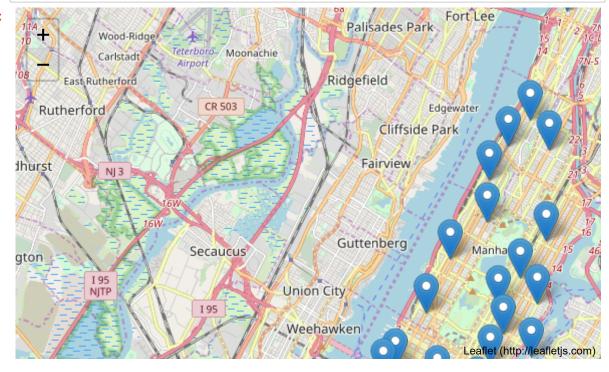
List of Manhattan Neighborhoods

Out[2]:

200

	Unnamed: 0	Borough	Neighborhood	Latitude	Longitude
0	0	Manhattan	Marble Hill	40.876551	-73.910660
1	1	Manhattan	Chinatown	40.715618	-73.994279
2	2	Manhattan	Washington Heights	40.851903	-73.936900
3	3	Manhattan	Inwood	40.867684	-73.921210
4	4	Manhattan	Hamilton Heights	40.823604	-73.949688

Out[49]:



Foursquare

Using foursquare API to get locations and count of restaurants. Since Foursquare doesn't have a fusion category so a google maps search result data was also used to get an idea of the fusion restaurants <u>URL</u>

 $\underline{(https://www.google.com/maps/search/fusion+restaurant/@40.7801874,-74.0368307,12z/data=!4m873.9712488!2d40.7830603)}$

API call output data is **not uploaded** to Github (response json was processed and results are posted on this notebook)

Some sample URLs (the real url was limited to a specific lat,long & radius with api keys included)

- Search Resturants URL https://api.foursquare.com/v2/venues/search?
 (https://api.foursquare.com/v2/venues/search?
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 (https://api.foursquare.com/v2/venues/search?
 (https://api.foursquare.com/v2/venues/search?
- Search Fusion Restaurants URL https://api.foursquare.com/v2/venues/search?
 II=40.7484,-73.9857&categoryId=4d4b7105d754a06374d81259&query=fusion
 (https://api.foursquare.com/v2/venues/search?
 II=40.7484,-73.9857&categoryId=4d4b7105d754a06374d81259&query=fusion)
- Explore Resturants URL https://api.foursquare.com/v2/venues/explore?
 (https://api.foursquare.com/v2/venues/explore?
- Explore Fusion Restaurants URL https://api.foursquare.com/v2/venues/explore?
 II=40.7484,-73.9857&categoryId=4d4b7105d754a06374d81259&query=fusion (https://api.foursquare.com/v2/venues/explore?
 II=40.7484,-73.9857&categoryId=4d4b7105d754a06374d81259&query=fusion)

```
In [18]:
             import json
             import ast
             from pandas.io.json import json normalize
             venue map = []
             fusion_venue_map = []
             uq ids = set()
             uq fids = set()
             for index, row in mtn_data.iterrows():
                 with open('foursquare/rest/'+str(index)+'.txt', encoding='utf-8-sig') as
                     venue_search_result = json.load(fh)
                     for venue in venue_search_result['response']['venues']:
                         category = 'U'
                         uq ids.add(venue.get("id"))
                         venue_map.append({"neigh":index,"id":venue.get("id"),"name":venue
                                            "lat":venue['location']['lat'],"lng":venue['loc
                 with open('foursquare/fuse/'+str(index)+'.txt', encoding='utf-8-sig') as
                     fusion venue search result = json.load(fh)
                     for venue in fusion venue search result['response']['venues']:
                         category = 'U'
                         uq fids.add(venue.get("id"))
                         fusion venue map.append({"neigh":index,"id":venue.get("id"),"name
                                            "lat":venue['location']['lat'],"lng":venue['loc
             all eateries = pd.DataFrame(venue map)
             all fusion eateries = pd.DataFrame(fusion venue map)
 In [4]:
             import numpy as np
             print('Total number of restaurants:',len(uq_ids))
             print('Total number of fusion restaurants:', len(uq fids))
             print('Percentage of fusion restaurants: {:.2f}%'.format(len(uq fids) / len(u
```

```
Show all collected data on the map with fusion restaurants in a different color
```

Total number of restaurants: 929
Total number of fusion restaurants: 6
Percentage of fusion restaurants: 0.65%

```
In [52]: M import folium
map_mtn = folium.Map(location=[40.7900869, -73.9598295], zoom_start=12)

#for index, row in all_eateries.iterrows():
    #folium.Marker(location=[ row.lat, row.lng ]).add_to( map_mtn )
for index, row in all_fusion_eateries.iterrows():
    folium.Marker(location=[ row.lat, row.lng ],icon=folium.Icon(color='#ff0000)
map_mtn
```

Out[52]:



Methodology

With data in hand we will look for desirable places with low restaurant density and with no fusion themed restaurants nearby. A heatmap will be generated to get a good picture of restaurant density. The promising ares will be further filtered using filter criteria.

Analysis

A count of restaurants in every neighbourhood

```
In [48]:
             rest counts by neigh = all eateries.groupby('neigh').size()
             for index, row in mtn_data.iterrows():
                  print(str(index)+" "+row['Neighborhood']+": "+str(rest counts by neigh[ir
             0 Marble Hill: 7
             1 Chinatown: 30
             2 Washington Heights: 16
             3 Inwood: 22
             4 Hamilton Heights: 30
             5 Manhattanville: 21
             6 Central Harlem: 21
             7 East Harlem: 30
             8 Upper East Side: 17
             9 Yorkville: 27
             10 Lenox Hill: 24
             11 Roosevelt Island: 8
             12 Upper West Side: 27
             13 Lincoln Square: 30
             14 Clinton: 17
             15 Midtown: 30
             16 Murray Hill: 30
             17 Chelsea: 25
             18 Greenwich Village: 30
             19 East Village: 30
             20 Lower East Side: 21
             21 Tribeca: 18
             22 Little Italy: 30
             23 Soho: 12
             24 West Village: 30
             25 Manhattan Valley: 18
             26 Morningside Heights: 30
             27 Gramercy: 15
             28 Battery Park City: 30
             29 Financial District: 30
             30 Carnegie Hill: 15
             31 Noho: 30
             32 Civic Center: 30
             33 Midtown South: 30
             34 Sutton Place: 30
             35 Turtle Bay: 22
             36 Tudor City: 9
             37 Stuyvesant Town: 4
             38 Flatiron: 30
             39 Hudson Yards: 23
```

List of neighborhoods with fusion eateries

At the start of the project the idea was to look for places where there arent many resturants but since it looks like there aren't many fusion restaurants we should look for places with quite a few restuarants indicating areas with good chunk of customers but away from existing fusion themed restaurants.

Results and Discussion

The analysis has shown that eventhough there are many restaurants in Manhattan there aren't many fusion themed restaurants. Based on the count and cluster there are many promising neighborhoods like Midtown South. Detailed analyis and additional reporting is outside the scope of this project and will be taken up in phase 2 and beyond.

Conclusion

Purpose of this project was to determine the viability of a fusion restaurant in Manhattan NY. The analysis has shown a few promising neighborhoods. The final decision will be made by investors based on additional factors like rental prices, crime, living vs work areas, walkability and other ROI factors. This concludes Phase I one of the 'Fusion Reactor' project.