The Battle of Neighborhoods (Week 1)

Part 1: Introduction and Data Sections

```
In []: import numpy as np # library to handle data in a vectorized manner
        import time
        import pandas as pd # library for data analsysis
        pd.set option('display.max columns', None)
        pd.set option('display.max rows', None)
        import json # library to handle JSON files
        import requests # library to handle requests
        from pandas.io.json import json normalize # tranform JSON file into a pandas dataframe
```

```
In []: from geopy.geocoders import Nominatim # convert an address into latitude and longitude values
        !conda install -c conda-forge folium=0.5.0 --yes # uncomment this line if you haven't completed the
        Foursquare API lab
        import folium # map rendering library
        print('Libraries imported.')
```

1. Introduction Section:

Discussion of the business problem and the audience who would be interested in this project.

Description of the Problem and Background

Scenario:

I am a student living Singapore. I want to use this opportunity to practice my learnings in Coursera in order to answer relevant questions arisen. The key question is: How can I find a convenient and enjoyable place similar to mine now in Singapore? Certainly, I can use available real estate apps and Google but the idea is to use and apply myself the learned tools during the course. In order to make a comparison and evaluation of the rental options in Manhattan NY, I must set some basis, therefore the apartment in Manhattan must meet the following demands:

```
apartment must be 2 or 3 bedrooms
desired location is near a metro station in the Manhattan area and within 1.0 mile (1.6 km) r
adius
price of rent not exceed $7,000 per month
top ammenities in the selected neighborhood shall be similar to current residence
desirable to have venues such as coffee shops, restaurants Asian Thai, wine stores, gym and f
ood shops
as a reference, I have included a map of venues near current residence in Singapore.
```

Business Problem:

The challenge is to find a suitable apartment for rent in Manhattan NY that complies with the demands on location, price and venues. The data required to resolve this challenge is described in the following section 2, below. Interested Audience

I believe this is a relevant challenge with valid questions for anyone moving to other large city in US, EU or Asia. The same methodology can be applied in accordance to demands as applicable. This case is also applicable for anyone interested in exploring starting or locating a new business in any city. Lastly, it can also serve as a good practical exercise to develop Data Science skills.

2. Data Section:¶

Description of the data and its sources that will be used to solve the problem

Description of the Data:¶

The following data is required to answer the issues of the problem:

```
List of Boroughs and neighborhoods of Manhattan with their geodata (latitud and longitud)
List of Subway metro stations in Manhattan with their address location
List of apartments for rent in Manhattan area with their addresses and price
Preferably, a list of apartment for rent with additional information, such as price, address,
area, # of beds, etc
Venues for each Manhattan neighborhood ( than can be clustered)
Venues for subway metro stations, as needed
```

The data will be used as follows:

How the data will be used to solve the problem

Use Foursquare and geopy data to map top 10 venues for all Manhattan neighborhoods and cluste

```
red in groups ( as per Course LAB)
Use foursquare and geopy data to map the location of subway metro stations , separately and o
```

n top of the above clustered map in order to be able to identify the venues and ammenities ne ar each metro station, or explore each subway location separately Use Foursquare and geopy data to map the location of rental places, in some form, linked to t he subway locations.

create a map that depicts, for instance, the average rental price per square ft, around a rad ious of 1.0 mile (1.6 km) around each subway station - or a similar metrics. I will be able t o quickly point to the popups to know the relative price per subway area. Addresses from rental locations will be converted to geodata (lat, long) using Geopy-distance and Nominatim.

Data will be searched in open data sources if available, from real estate sites if open to re ading, libraries or other government agencies such as Metro New York MTA, etc.

The processing of these DATA will allow to answer the key questions to make a decision: what is the cost of rent (per square ft) around a mile radius from each subway metro station?

```
what is the area of Manhattan with best rental pricing that meets criteria established?
What is the distance from work place ( Park Ave and 53 rd St) and the tentative future home?
What are the venues of the two best places to live? How the prices compare?
How venues distribute among Manhattan neighborhoods and around metro stations?
Are there tradeoffs between size and price and location?
Any other interesting statistical data findings of the real estate and overall data.
```

Reference of venues around current residence in Singapore for comparison to Manhattan place In []: # Shenton Way, District 01, Singapore

```
address = 'Mccallum Street, Singapore'
        geolocator = Nominatim()
        location = geolocator.geocode(address)
        latitude = location.latitude
        longitude = location.longitude
        print('The geograpical coordinate of Singapore home are {}, {}.'.format(latitude, longitude))
In [ ]: neighborhood latitude=1.2792655
        neighborhood longitude=103.8480938
```

```
In [16]: CLIENT ID = 'DVCTZDPDYXTS0BRJFPLMHM323APGXNWZI5PLRQ1VC0CFLF1T' # your Foursquare ID
         CLIENT SECRET = '5NWAGXRLXIXAVOL3DNYY1EPIHNMAAAIZFDFELYSYXL5LFWL1' # your Foursquare Secret
         VERSION = '20180605' # Foursquare API version
         print('Your credentails:')
         print('CLIENT ID: ' + CLIENT ID)
         print('CLIENT SECRET:' + CLIENT SECRET)
         Your credentails:
         CLIENT ID: DVCTZDPDYXTS0BRJFPLMHM323APGXNWZI5PLRQ1VC0CFLF1T
         CLIENT SECRET: 5NWAGXRLXIXAV0L3DNYY1EPIHNMAAAIZFDFELYSYXL5LFWL1
```

In [17]: LIMIT = 100 # limit of number of venues returned by Foursquare API

5,103.8480938&radius=500&limit=100'

```
radius = 500 # define radius
         # create URL
         url = 'https://api.foursquare.com/v2/venues/explore?&client id={}&client secret={}&v={}&l={},{}&rad
         ius={}&limit={}'.format(
             CLIENT ID,
             CLIENT SECRET,
             VERSION,
             neighborhood latitude,
             neighborhood longitude,
             radius,
             LIMIT)
         url # display URL
Out[17]: 'https://api.foursquare.com/v2/venues/explore?&client id=DVCTZDPDYXTS0BRJFPLMHM323APGXNWZI5PLRQ1
         VC0CFLF1T&client secret=5NWAGXRLXIXAV0L3DNYY1EPIHNMAAAIZFDFELYSYXL5LFWL1&v=20180605&11=1.279265
```

In [18]: results = requests.get(url).json() #results

```
In [19]: # function that extracts the category of the venue
         def get_category_type(row):
             try:
                 categories list = row['categories']
                 categories_list = row['venue.categories']
             if len(categories list) == 0:
                 return None
             else:
                 return categories_list[0]['name']
In [20]: venues = results['response']['groups'][0]['items']
         SGnearby venues = json normalize(venues) # flatten JSON
         # filter columns
```

```
filtered_columns = ['venue.name', 'venue.categories', 'venue.location.lat', 'venue.location.lng']
          SGnearby_venues =SGnearby_venues.loc[:, filtered_columns]
          # filter the category for each row
          SGnearby venues['venue.categories'] = SGnearby_venues.apply(get_category_type, axis=1)
          # clean columns
          SGnearby_venues.columns = [col.split(".")[-1] for col in SGnearby_venues.columns]
          SGnearby venues.head(10)
Out[20]:
                                         categories
                                                                 Ing
                            name
          0 Napoleon Food & Wine Bar
                                          Wine Bar 1.279925 103.847333
```

```
Native
                                             Cocktail Bar 1.280135 103.846844
           2
                          Pepper Bowl
                                         Asian Restaurant 1.279371 103.846710
           3
                       Park Bench Deli
                                            Deli / Bodega 1.279872 103.847287
                           Muchachos
                                            Burrito Place 1.279072 103.847026
           5
                       Mellower Coffee
                                                   Café 1.277814 103.848188
           6
                                            Beer Garden 1.281254 103.848513
                           Freehouse
           7
                    Sofitel So Singapore
                                                  Hotel 1.280017 103.849813
                     Dumpling Darlings Dumpling Restaurant 1.280483 103.846942
           9
                             PS.Cafe
                                                   Café 1.280468 103.846264
In [21]: # create map of Singapore place using latitude and longitude values
           map sg = folium.Map(location=[latitude, longitude], zoom start=20)
           # add markers to map
```

```
for lat, lng, label in zip(SGnearby venues['lat'], SGnearby venues['lng'], SGnearby venues['name']):
             label = folium.Popup(label, parse html=True)
             folium.RegularPolygonMarker(
                 [lat, lng],
                 number of sides=4,
                 radius=10,
                 popup=label,
                 color='blue',
                 fill color='#0f0f0f',
                 fill opacity=0.7,
             ).add_to(map_sg)
         map_sg
Out[21]:
            +
```