

## 1. Introduction: Business Problem

My sister – in - law wanted to open a restaurant or a cafe in District 2, Ho Chi Minh, but she didn't know where to open with little competition. This data analysis article will clarify and may help him with some useful information for her decision.

In this project we will try to find an optimal location for a restaurant or cafe. Specifically, this report will be targeted to stakeholders interested in opening an **Restaurant or Cafe in District 2, Ha Noi, Viet nam.**

We will use our data science powers to generate a few most promising neighborhoods based on this criteria. Advantages of each area will then be clearly expressed so that best possible final location can be chosen by stakeholders.

## 2. Data

**Based on definition of our problem, factors that will influence our decision are:**

- Detail information of neighborhoods in District 2, list of districts, wards of District 2, Ho Chi Minh from the following URL:  
<http://www.pso.hochiminhcity.gov.vn/web/guest/danhmucthongke-danhmuctinhthanhpho>  
<http://www.pso.hochiminhcity.gov.vn/web/guest/danhmucthongke-danhmucphuongxa> or file data xls from the following:  
[https://github.com/chaudb39/Capstone\\_Cousera/blob/e4b872054271da617fcb10566faa3ea8966df29a/HCM\\_DISTRICT2.xlsx](https://github.com/chaudb39/Capstone_Cousera/blob/e4b872054271da617fcb10566faa3ea8966df29a/HCM_DISTRICT2.xlsx)

- Number of existing restaurants in the neighborhood (any type of restaurant)

## Google map API

This project would use Google Map API Geocoder to get the Latitude and Longitude of each area

## Foursquare API

This project would use Four-square API as its prime data gathering source. This API provides the ability to perform location search, location sharing and details about a business.

## Step by step following

### Install packages

```
[3]: !pip install lxml
      !pip install bs4
      !pip install Nominatim
      !pip install geopy
      !pip install geocoder
      !pip install xlrd
```

### 2.1. Load necessary library

```
[4]: import numpy as np # library to handle data in a vectorized manner
import pandas as pd # library for data analysis
pd.set_option("display.max_columns", None)
pd.set_option("display.max_rows", None)
import json # library to handle JSON files
from geopy.geocoders import Nominatim # convert an address into latitude and longitude values
import geocoder # to get coordinates
import requests # library to handle requests
from bs4 import BeautifulSoup # library to parse HTML and XML documents
from pandas.io.json import json_normalize # tranform JSON file into a pandas dataframe
# Matplotlib and associated plotting modules
import matplotlib.cm as cm
import matplotlib.colors as colors
# import k-means from clustering stage
from sklearn.cluster import KMeans
import lxml
import folium # map rendering library
import pandas as pd
import lxml
import xlrd
print("Libraries imported.")
```

Libraries imported.

## 2.2. Get Data District 2

<http://www.pso.hochiminhcity.gov.vn/web/guest/danhmucthongke-danhmuctinhthanhpho>

<http://www.pso.hochiminhcity.gov.vn/web/guest/danhmucthongke-danhmucphuongxa> or file data xls from the following:

[https://github.com/chaudb39/Capstone\\_Cousera/blob/e4b872054271da617fcb10566faa3ea8966df29a/HCM\\_DISTRICT2.xlsx](https://github.com/chaudb39/Capstone_Cousera/blob/e4b872054271da617fcb10566faa3ea8966df29a/HCM_DISTRICT2.xlsx)

## 2.3. Load file excel districts, wards of VietNam

```
[6]: df = pd.read_excel('HCM_DISTRICT2.xlsx')
```

```
[7]: df.head()
```

```
[7]:
```

	Tỉnh Thành Phố	Mã TP	Quận Huyện	Mã QH	Phường Xã	Mã PX	Cấp	Tên Tiếng Anh
0	Thành phố Hồ Chí Minh	79	Quận 2	769	Phường Thảo Điền	27088	Phường	NaN
1	Thành phố Hồ Chí Minh	79	Quận 2	769	Phường An Phú	27091	Phường	NaN
2	Thành phố Hồ Chí Minh	79	Quận 2	769	Phường Bình An	27094	Phường	NaN
3	Thành phố Hồ Chí Minh	79	Quận 2	769	Phường Bình Trưng Đông	27097	Phường	NaN
4	Thành phố Hồ Chí Minh	79	Quận 2	769	Phường Bình Trưng Đông	27100	Phường	NaN

```

+ ✂ 📄 📌 ▶ ■ ↺ ▶ Code ▼ ⌚ git Run as Pipeline
[8]: df['area'] = df['Phường Xã'] + ', ' + df['Quận Huyện'] + ', Hồ Chí Minh'

df_district2 = df[['Phường Xã', 'Quận Huyện', 'area']]
df_district2.columns = ['ward', 'district', 'area']

[9]: df_district2.head(11)

[9]:

```

	ward	district	area
0	Phường Thảo Điền	Quận 2	Phường Thảo Điền, Quận 2, Hồ Chí Minh
1	Phường An Phú	Quận 2	Phường An Phú, Quận 2, Hồ Chí Minh
2	Phường Bình An	Quận 2	Phường Bình An, Quận 2, Hồ Chí Minh
3	Phường Bình Trưng Đông	Quận 2	Phường Bình Trưng Đông, Quận 2, Hồ Chí Minh
4	Phường Bình Trưng Đông	Quận 2	Phường Bình Trưng Đông, Quận 2, Hồ Chí Minh
5	Phường Bình Khánh	Quận 2	Phường Bình Khánh, Quận 2, Hồ Chí Minh
6	Phường An Khánh	Quận 2	Phường An Khánh, Quận 2, Hồ Chí Minh
7	Phường Cát Lái	Quận 2	Phường Cát Lái, Quận 2, Hồ Chí Minh
8	Phường Thạnh Mỹ Lợi	Quận 2	Phường Thạnh Mỹ Lợi, Quận 2, Hồ Chí Minh
9	Phường An Lợi Đông	Quận 2	Phường An Lợi Đông, Quận 2, Hồ Chí Minh
10	Phường Thủ Thiêm	Quận 2	Phường Thủ Thiêm, Quận 2, Hồ Chí Minh

## 2.4. Add latitude, longitude by call Google Geocode API

```

[12]: # define a function to get coordinates
def get_latlng(neighborhood):
    # initialize your variable to None
    lat_lng_coors = None
    # loop until you get the coordinates
    while(lat_lng_coors is None):
        g = geocoder.arcgis('{}, Malaysia'.format(neighborhood))
        lat_lng_coors = g.latlng
    return lat_lng_coors

[14]: coors = [get_latlng(neighborhood) for neighborhood in df_district2["area"].tolist()]

[15]: # create temporary dataframe to populate the coordinates into Latitude and Longitude
df_district2_coors = pd.DataFrame(coors, columns=['Latitude', 'Longitude'])

[17]: df_district2_coors.head(11)

[17]:

```

	Latitude	Longitude
0	10.81029	106.72968
1	10.80156	106.75369
2	10.79289	106.73087
3	10.78511	106.77805

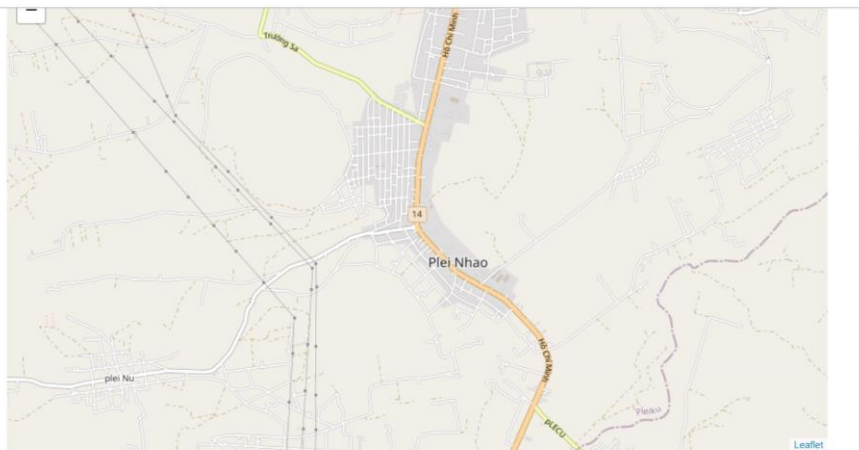
## 2.5. Create a map of District 2 – Ho Chi Minh City with neighborhoods superimposed on top

```
[25]: address='Quận 2, Hồ Chí Minh, Việt Nam'
geolocator = Nominatim(user_agent="HoChiMinh")
location = geolocator.geocode(address)
lat_HCM=location.latitude
long_HCM=location.longitude
print('The geographical coordinate of District 2, Ho Chi Minh are {},{}'.format(lat_HCM,long_HCM))

The geographical coordinate of District 2, Ho Chi Minh are 13.9173618,108.0051396.

[26]: map_HCM = folium.Map(location=[lat_HCM, long_HCM], zoom_start=14)
# add markers to map
for lat, lng, Neighbourhood in zip(df_district2_new['Latitude'], df_district2_new['Longitude'], df_district2_new['ward']):
    label = '{}'.format(Neighbourhood)
    marker = folium.Marker(location=[lat, lng], popup=folium.Popup(label, parse_html=True))
    marker.add_to(map_HCM)

map_HCM
```



## 2.6. Use the Foursquare API to explore the neighborhoods

```
[20]: CLIENT_ID = '4IWPTPZBLHK1ISW0CQEP0IBTOX3W40SIRB1DHH3VMM3PYXWU' # your Foursquare ID
CLIENT_SECRET = 'QTRBGWSK1RJJBEGSP0YGEE1IIS3EOZKKQY40FWTRZDLNB4' # your Foursquare Secret
VERSION = '20180604'
LIMIT = 100
print('Your credentails:')
print('CLIENT_ID: ' + CLIENT_ID)
print('CLIENT_SECRET: ' + CLIENT_SECRET)

Your credentails:
CLIENT_ID: 4IWPTPZBLHK1ISW0CQEP0IBTOX3W40SIRB1DHH3VMM3PYXWU
CLIENT_SECRET: QTRBGWSK1RJJBEGSP0YGEE1IIS3EOZKKQY40FWTRZDLNB4

[21]: def getNearbyVenues(names, latitudes, longitudes, radius=500):
    venues_list=[]
    for name, lat, lng in zip(names, latitudes, longitudes):
        print(name)

    # create the API request URL
    url = 'https://api.foursquare.com/v2/venues/explore?&client_id={}&client_secret={}&v={}&ll={}&radius={}&limit={}'.format(
        CLIENT_ID,
        CLIENT_SECRET,
        VERSION,
        lat,
        lng,
        radius,
        LIMIT)
```

```

CLIENT_SECRET,--
VERSION,--
lat,--
lng,--
radius,--
LIMIT)

-----
# make the GET request
results = requests.get(url).json()["response"][0]['items']

-----
# return only relevant information for each nearby venue
venues_list.append([
    name,--
    lat,--
    lng,--
    v['venue']['name'],--
    v['venue']['location']['lat'],--
    v['venue']['location']['lng'],--
    v['venue']['categories'][0]['name'] for v in results])

nearby_venues = pd.DataFrame([item for venue_list in venues_list for item in venue_list])
nearby_venues.columns = ['Neighbourhood',--
                        'Neighbourhood Latitude',--
                        'Neighbourhood Longitude',--
                        'Venue',--
                        'Venue Latitude',--
                        'Venue Longitude',--
                        'Venue Category']

-----
return(nearby_venues)

```

## Check how many venues were returned for each neighborhood

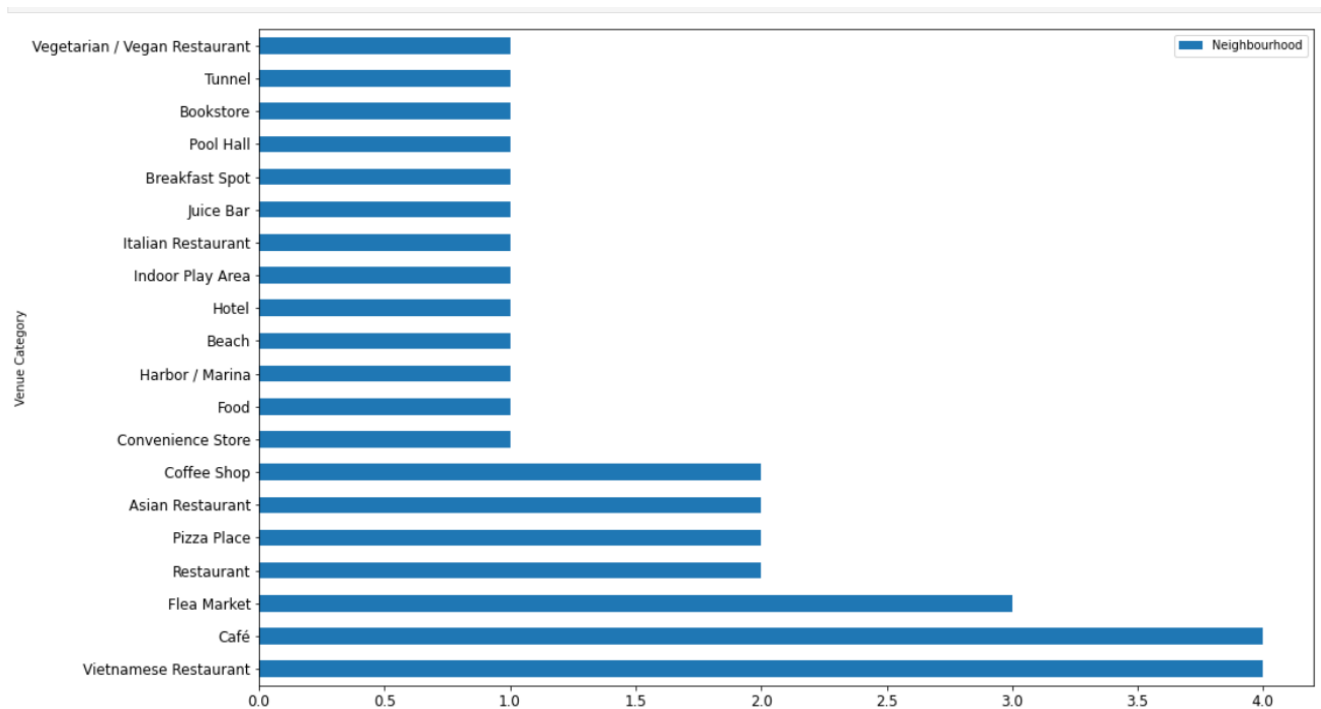
```

[40]: HCM_District2_venues = HCM_venues.groupby('Venue Category').count()

[41]: HCM_District2_venues = HCM_District2_venues.reindex(columns=['Neighbourhood'])
HCM_District2_venues = HCM_District2_venues.sort_values(by=['Neighbourhood'], ascending=False).head(20)
HCM_District2_venues.to_csv('HCM_District2_venues.csv')

```

## Draw char top Venue Category common



Next, let's group rows by neighborhood and by taking the mean of the frequency of occurrence of each category

```
[47]: HCM_grouped=hcm_onehot.groupby('Neighbourhood').mean().reset_index()
      HCM_grouped
```

[47]:

	Neighbourhood	Asian Restaurant	Beach	Bookstore	Breakfast Spot	Café	Coffee Shop	Convenience Store	Flea Market	Food	Harbor / Marina	Health & Beauty Service	Hotel	Indoor Play Area	Italian Restaurant
0	Phường An Lợi Đông, Quận 2, Hồ Chí Minh	0.000000	0.25	0.00	0.000000	0.250000	0.000000	0.0	0.00	0.00	0.0	0.00	0.0	0.0	0.0
1	Phường An Phú, Quận 2, Hồ Chí Minh	0.000000	0.00	0.00	0.000000	0.000000	0.200000	0.2	0.00	0.00	0.0	0.00	0.2	0.2	0.0
2	Phường Bình An, Quận 2, Hồ Chí Minh	0.166667	0.00	0.00	0.000000	0.000000	0.000000	0.0	0.00	0.00	0.0	0.00	0.0	0.0	0.0
3	Phường Bình Khánh, Quận 2, Hồ Chí Minh	0.000000	0.00	0.25	0.000000	0.000000	0.000000	0.0	0.25	0.25	0.0	0.00	0.0	0.0	0.0
4	Phường Bình Trưng Đông, Quận 2, Hồ Chí Minh	0.000000	0.00	0.00	0.000000	0.000000	0.000000	0.0	1.00	0.00	0.0	0.00	0.0	0.0	0.0
5	Phường Cát Lái, Quận 2, Hồ Chí Minh	0.000000	0.00	0.00	0.000000	0.500000	0.000000	0.0	0.00	0.00	0.5	0.00	0.0	0.0	0.0

Create the new dataframe and display the top 10 venues for each neighborhood

```
[50]: num_top_venues = 10
indicators = ['st', 'nd', 'rd']

# create columns according to number of top venues
columns = ['Neighbourhood']
for ind in np.arange(num_top_venues):
    try:
        columns.append('{}{} Most Common Venue'.format(ind+1, indicators[ind]))
    except:
        columns.append('{}th Most Common Venue'.format(ind+1))

# create a new dataframe
neighbourhoods_venues_sorted = pd.DataFrame(columns=columns)
neighbourhoods_venues_sorted['Neighbourhood'] = HCM_grouped['Neighbourhood']

for ind in np.arange(HCM_grouped.shape[0]):
    neighbourhoods_venues_sorted.iloc[ind, 1:] = return_most_common_venues(HCM_grouped.iloc[ind, :], num_top_venues)

neighbourhoods_venues_sorted.head()
```

	Neighbourhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
0	Phường An Lợi Đông, Quận 2, Hồ Chí Minh	Beach	Pool Hall	Juice Bar	Café	Vietnamese Restaurant	Food	Bookstore	Breakfast Spot	Coffee Shop	Convenience Store
1	Phường An Phú, Quận 2, Hồ Chí Minh	Convenience	Indoor	Vegetarian / Vietnamese	Vegetarian / Vietnamese	Vegetarian / Vietnamese	Vegetarian / Vietnamese	Vegetarian / Vietnamese	Vegetarian / Vietnamese	Vegetarian / Vietnamese	Breakfast

### 3. Methodology

After data acquisition and cleaning, this project applies **K-mean clustering unsupervised machine learning algorithm** to cluster the venues based on a list of locations for different types of food and beverage service points such as bars, cafes, Chinese restaurants, Vietnamese restaurants, Seafood restaurants, etc. This would give a better understanding of the similarities and dissimilarities between the chosen neighborhoods to retrieve more insights.

Analyze Each Neighborhood, group rows by neighborhood and by taking the mean of the frequency of occurrence of each category. Next, create the new data frame and display the top 10 venues for each neighborhood.

Then use the Kmean algorithm from the sklearn library to divide it into 5 groups with similar properties. Next, assign labels from Kmean result to each neighborhood using the Pandas merge function

### 4. Analysis



Let’s create a new dataframe that includes the cluster as well as the top 10 venues for each neighborhood.

```
[52]: # add clustering labels
neighbourhoods_venues_sorted.insert(0, 'Cluster_Labels', kmeans.labels_)
neighbourhoods_venues_sorted.head()
```

```
[52]:
```

	Cluster_Labels	Neighbourhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
0	1	Phường An Lợi Đông, Quận 2, Hồ Chí Minh	Beach	Pool Hall	Juice Bar	Café	Vietnamese Restaurant	Food	Bookstore	Breakfast Spot	Coffee Shop	Convenience Store
1	3	Phường An Phú, Quận 2, Hồ Chí Minh	Coffee Shop	Convenience Store	Indoor Play Area	Hotel	Vegetarian / Vegan Restaurant	Vietnamese Restaurant	Food	Beach	Bookstore	Breakfast Spot
2	4	Phường Bình An, Quận 2, Hồ Chí Minh	Vietnamese Restaurant	Restaurant	Pizza Place	Asian Restaurant	Pool Hall	Flea Market	Beach	Bookstore	Breakfast Spot	Café
3	4	Phường Bình Khánh, Quận 2, Hồ Chí Minh	Vietnamese Restaurant	Bookstore	Flea Market	Food	Harbor / Marina	Beach	Breakfast Spot	Café	Coffee Shop	Convenience Store
4	2	Phường Bình Trưng Đông, Quận 2, Hồ Chí Minh	Flea Market	Vietnamese Restaurant	Harbor / Marina	Beach	Bookstore	Breakfast Spot	Café	Coffee Shop	Convenience Store	Food

```
[53]: HCM_merged = df_district2_new

# merge toronto_grouped with toronto_data to add Latitude/Longitude for each neighborhood
HCM_merged = HCM_merged.join(neighbourhoods_venues_sorted.set_index('Neighbourhood'), on='area')

HCM_merged.head() # check the last columns!
```

```
[53]:
```

	ward	district	area	Latitude	Longitude	Cluster_Labels	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue
0	Phường Thảo Điền	Quận 2	Phường Thảo Điền, Quận 2, Hồ Chí Minh	10.81029	106.72968	1.0	Health & Beauty Service	Pizza Place	Café	Italian Restaurant	Food	Beach	Bookstore	Breakfast Spot
1	Phường An Phú	Quận 2	Phường An Phú, Quận 2, Hồ Chí Minh	10.80156	106.75369	3.0	Coffee Shop	Convenience Store	Indoor Play Area	Hotel	Vegetarian / Vegan Restaurant	Vietnamese Restaurant	Food	Breakfast Spot
2	Phường Bình An	Quận 2	Phường Bình An, Quận 2, Hồ Chí Minh	10.79289	106.73087	4.0	Vietnamese Restaurant	Restaurant	Pizza Place	Asian Restaurant	Pool Hall	Flea Market	Beach	Bookstore

Create map cluster

```
[93]: HCM_merged['Cluster_Labels'] = HCM_merged.Cluster_Labels.astype(int)
      HCM_merged.head(11)
```

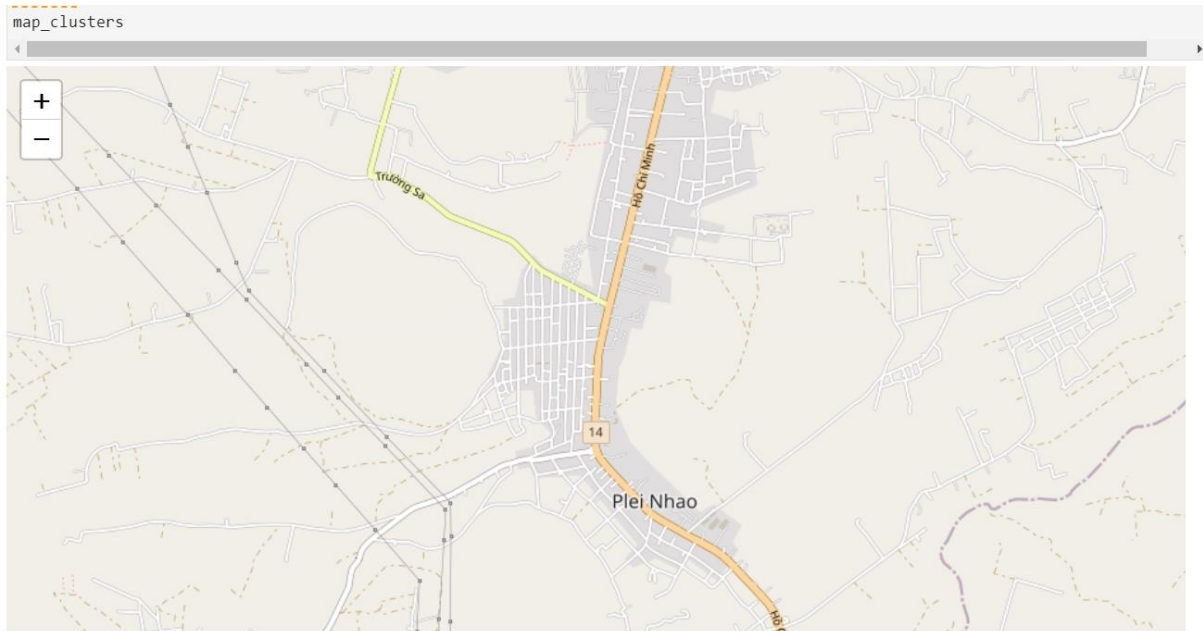
```
[93]:
```

	ward	district	area	Latitude	Longitude	Cluster_Labels	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue
0	Phường Thảo Điền	Quận 2	Phường Thảo Điền, Quận 2, Hồ Chí Minh	10.81029	106.72968	1	Health & Beauty Service	Pizza Place	Café	Italian Restaurant	Food	Beach	Bookstore
1	Phường An Phú	Quận 2	Phường An Phú, Quận 2, Hồ Chí Minh	10.80156	106.75369	3	Coffee Shop	Convenience Store	Indoor Play Area	Hotel	Vegetarian / Vegan Restaurant	Vietnamese Restaurant	Food
2	Phường Bình An	Quận 2	Phường Bình An, Quận 2, Hồ Chí Minh	10.79289	106.73087	4	Vietnamese Restaurant	Restaurant	Pizza Place	Asian Restaurant	Pool Hall	Flea Market	Beach
	Phường Bình Trưng						Bar /				Breakfast		

```
[94]: # create map
      map_clusters = folium.Map(location=[lat_HCM, long_HCM], zoom_start=14)

      # set color scheme for the clusters
      x = np.arange(kclusters)
      ys = [i + x + (i*x)**2 for i in range(kclusters)]
      colors_array = cm.rainbow(np.linspace(0, 1, len(ys)))
      rainbow = [colors.rgb2hex(i) for i in colors_array]

      # add markers to the map
      markers_colors = []
      for lat, lon, poi, cluster in zip(HCM_merged['Latitude'], HCM_merged['Longitude'], HCM_merged['area'], HCM_merged['Cluster_Labels']):
          label = folium.Popup(str(poi) + ' Cluster ' + str(cluster), parse_html=True)
          folium.CircleMarker(
              [lat, lon],
              radius=5,
              popup=label,
              color=rainbow[cluster-1],
              fill=True,
              fill_color=rainbow[cluster-1],
              fill_opacity=0.7).add_to(map_clusters)
```



## **5. Conclusion**

Finally, I have got a small glimpse of how real-life data-science projects look like. I used various types of APIs to collect data, used the Pandas library to eliminate redundant data, used it, and used Python libraries to draw graphs, using unsupervised machine learning algorithms to group data into similar characteristics. From that it is possible to discover the information that is hidden in it, making it easier to make decisions such as where to open a restaurant or a cafe is appropriate and less competitive

## **6. Final Notes**

This is my assignment: a part of the IBM Data Science Course on Coursera.

The full project Jupiter Notebook from data scraping to preprocessing to results

here: [https://github.com/chaudb39/Capstone\\_Cousera/blob/e4b872054271da617fcb10566faa3ea8966df29a/CourseraCapstone\(Week2\).ipynb](https://github.com/chaudb39/Capstone_Cousera/blob/e4b872054271da617fcb10566faa3ea8966df29a/CourseraCapstone(Week2).ipynb)

Dinh Bao Chau