

Cognifyz Data Analysys LEVEL_2 (Shantanu Garain)

Importing Libraries

```
In [1]: import pandas as pd
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans
import folium
import warnings
warnings.filterwarnings('ignore')
```

```
In [2]: df = pd.read_csv('Dataset.csv')
```

LEVEL_2 Task_1 (Restaurent Ratings)

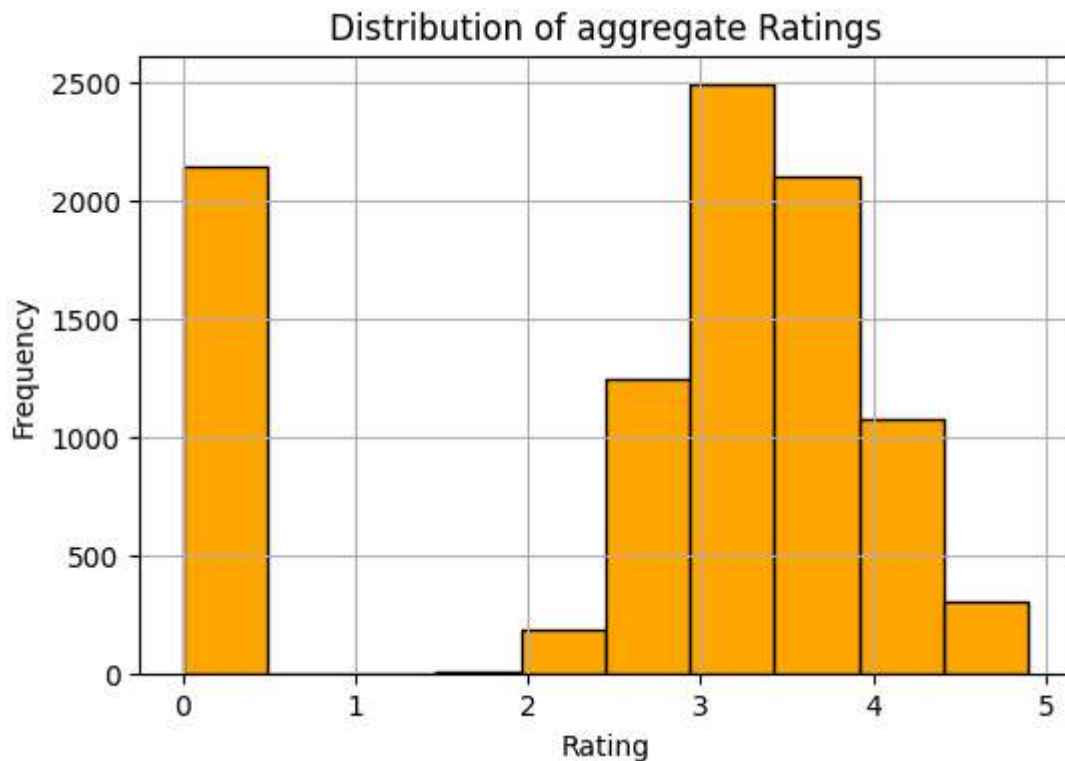
1. Analyze the distribution of aggregate ratings and determine the most common rating range

```
In [3]: def rating_distribution(dataset_path):
plt.figure(figsize=(6,4))
plt.hist(df['Aggregate rating'], bins = 10, edgecolor = 'black', color = 'orange')
plt.xlabel('Rating')
plt.ylabel('Frequency')
plt.title('Distribution of aggregate Ratings')
plt.grid(True)
plt.show()

most_common_range = df['Aggregate rating'].mode()[0]
return most_common_range

dataset_path = 'Restaurants.csv'
most_common_rating_range = rating_distribution(dataset_path)

print(f"The most common rating range is: {most_common_rating_range}")
```



The most common rating range is: 0.0

2. Calculate the average number of votes received by restaurants.

```
In [4]: avg_votes = df['Votes'].mean()  
print(f"The average number of votes is {avg_votes} received by restaurents")
```

The average number of votes is 156.909747670401 received by restaurents

LEVEL_2 Task_2 (Cuisine Combination)

1. Identify the most common combinations of cuisines in the dataset.

```
In [5]: common_combination = df['Cuisines'].str.split(',').explode()  
count = common_combination.value_counts()  
print(count)
```

```

Cuisines
North Indian      2992
Chinese           1880
Fast Food         1314
North Indian      968
Chinese           855
...
Ramen             1
Indonesian        1
Bubble Tea        1
Kiwi              1
World Cuisine     1
Name: count, Length: 249, dtype: int64

```

2. Determine if certain cuisine combinations tend to have higher ratings.

```

In [6]: rating = df.groupby('Cuisines')['Aggregate rating'].mean()
rating

```

```

Out[6]: Cuisines
Afghani                      0.725
Afghani, Mughlai, Chinese     0.000
Afghani, North Indian         0.000
Afghani, North Indian, Pakistani, Arabian  0.000
African                      4.700
...
Western, Asian, Cafe          4.200
Western, Fusion, Fast Food    3.200
World Cuisine                 4.900
World Cuisine, Mexican, Italian  4.400
World Cuisine, Patisserie, Cafe  4.200
Name: Aggregate rating, Length: 1825, dtype: float64

```

```

In [7]: # Finding the top ratings
top_rating = rating.sort_values(ascending=False)
top_rating

```

```

Out[7]: Cuisines
Italian, Deli                4.9
Hawaiian, Seafood            4.9
American, Sandwich, Tea      4.9
Continental, Indian           4.9
European, Asian, Indian       4.9
...
Fast Food, Desserts, Beverages  0.0
Fast Food, Chinese, Burger     0.0
Cafe, French                  0.0
Seafood, Mughlai, North Indian  0.0
Fast Food, Mughlai, Armenian    0.0
Name: Aggregate rating, Length: 1825, dtype: float64

```

```

In [8]: print(f"cuisine combinations tend to have higher ratings is: {top_rating}")

```

```
cuisine combinations tend to have higher ratings is: Cuisines
Italian, Deli 4.9
Hawaiian, Seafood 4.9
American, Sandwich, Tea 4.9
Continental, Indian 4.9
European, Asian, Indian 4.9
...
Fast Food, Desserts, Beverages 0.0
Fast Food, Chinese, Burger 0.0
Cafe, French 0.0
Seafood, Mughlai, North Indian 0.0
Fast Food, Mughlai, Armenian 0.0
Name: Aggregate rating, Length: 1825, dtype: float64
```

LEVEL_2 Task_3 (Geographic Analysis)

1. Plotting the locations of restaurants on a map using longitude and latitude coordinates.

```
In [9]: df.columns
```

```
Out[9]: Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address',
            'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines',
            'Average Cost for two', 'Currency', 'Has Table booking',
            'Has Online delivery', 'Is delivering now', 'Switch to order menu',
            'Price range', 'Aggregate rating', 'Rating color', 'Rating text',
            'Votes'],
            dtype='object')
```

```
In [10]: res_name = df['Restaurant Name']
lat = df['Latitude']
long = df['Longitude']
```

```
In [11]: X = df[['Latitude', 'Longitude']]
num_cluster = 5
```

```
In [12]: kmeans = KMeans(n_clusters=num_cluster, n_init = 10, random_state=42)
df['cluster'] = kmeans.fit_predict(X)
```

```
In [13]: map_centre = [lat.mean(), long.mean()]
rest_map = folium.Map(location = map_centre, zoom_starts = 12)
```

```
In [14]: cl_color = ['purple', 'red', 'blue', 'Green', 'black']
```

```
In [15]: for index, row in df.iterrows():
            res_name = row['Restaurant Name']
            lat = row['Latitude']
            long = row['Longitude']
            cusines = row['Cuisines']
            rating = row['Aggregate rating']
            cluster = row['cluster']
```

```
In [16]: popup_text = f'Restaurent: {res_name}\nCusine: {cusines}\nRating: {rating}'
```

```
In [17]: marker = folium.Marker([lat, long], popup = popup_text)
```


```
In [18]: marker.add_to(rest_map)
```

```
Out[18]: <folium.map.Marker at 0x213205e1fd0>
```

```
In [19]: display(rest_map)
```

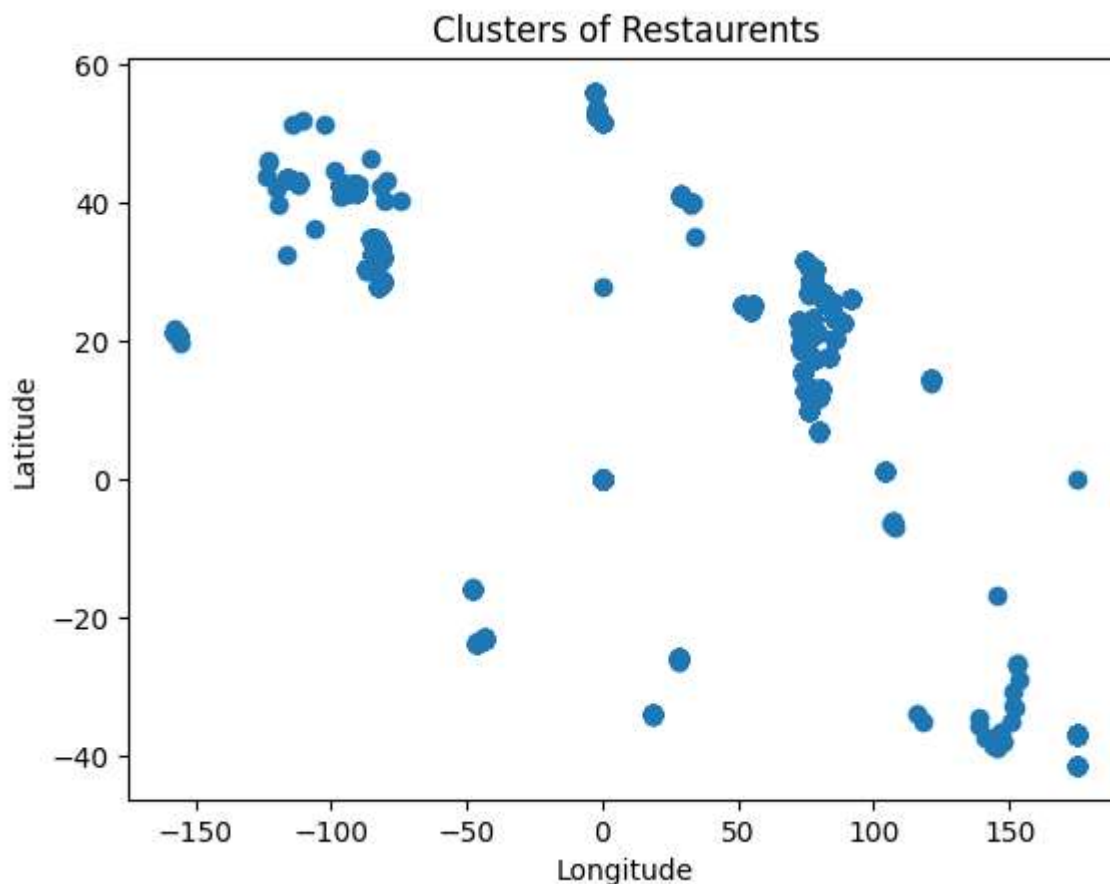
Make this Notebook Trusted to load map: File -> Trust Notebook



 Leaflet (<https://leafletjs.com>) | Data by © OpenStreetMap (<http://openstreetmap.org>), under ODbL (<http://www.openstreetmap.org/copyright>).

2. Identify any patterns or clusters of restaurants in specific areas.

```
In [20]: plt.scatter(df['Longitude'],df['Latitude'])
plt.xlabel('Longitude')
plt.ylabel('Latitude')
plt.title('Clusters of Restaurents')
plt.show()
```



LEVEL_2 Task_4 (Restaurant Chains)

1. Identify if there are any restaurant chains present in the dataset.

```
In [21]: chain = df['Restaurant Name'].value_counts()
rest_chain = chain[chain>1]
rest_chain
```

```
Out[21]: Restaurant Name
Cafe Coffee Day      83
Domino's Pizza       79
Subway               63
Green Chick Chop     51
McDonald's          48
..
Town Hall            2
Halki Aanch          2
Snack Junction        2
Delhi Biryani Hut     2
Beliram Degchiwala    2
Name: count, Length: 734, dtype: int64
```

2. Analyze the ratings and popularity of different restaurant chains.

```
In [22]: chain_stat = df.groupby('Restaurant Name').agg({'Aggregate rating': 'mean', 'Restaurant
chain_stat.columns = ['Restaurant Name', 'Average Rating', 'Popularity']
```

```
In [23]: # Analyze popularity of different restaurant with avg rating
chain_stat.sort_values('Average Rating', ascending=False, inplace=True)
chain_stat
```

```
Out[23]:
```

	Restaurant Name	Average Rating	Popularity
--	-----------------	----------------	------------

5322	Restaurant Mosaic @ The Orient	4.9	1
4177	Ministry of Crab	4.9	1
4135	Miann	4.9	1
5757	Shorts Burger and Shine	4.9	1
4165	Milse	4.9	1
...
1105	Cafe Corner	0.0	1
4905	Pheva Tandooris	0.0	1
4913	Pick & Carry	0.0	1
1096	Cafe Brownie	0.0	1
2292	Famous Parantha and Poori Sabzi	0.0	1

7446 rows × 3 columns

```
In [24]: chain_stat.sort_values('Popularity', ascending=True)
chain_stat
```

Out[24]:

	Restaurant Name	Average Rating	Popularity
5322	Restaurant Mosaic @ The Orient	4.9	1
4177	Ministry of Crab	4.9	1
4135	Miann	4.9	1
5757	Shorts Burger and Shine	4.9	1
4165	Milse	4.9	1
...
1105	Cafe Corner	0.0	1
4905	Pheva Tandooris	0.0	1
4913	Pick & Carry	0.0	1
1096	Cafe Brownie	0.0	1
2292	Famous Parantha and Poori Sabzi	0.0	1

7446 rows × 3 columns

Let's visualize it

```
In [25]: plt.figure(figsize=(8,6))
plt.subplot(1,2,1)
plt.bar(chain_stat['Restaurant Name'][:10], chain_stat['Average Rating'][:10],color='r')
plt.xlabel('Restaurant Name')
plt.ylabel('Average Rating')
plt.title('Ratings of Restaurant Chain')
plt.subplot(1,2,2)
plt.bar(chain_stat['Restaurant Name'][:10],chain_stat['Popularity'][:10],color='y')
plt.xlabel('Restaurant Name')
plt.ylabel('Popularity')
plt.title('Popularity of Restaurant Chain')
plt.show()
```




In []: