Sejal Sanas

Perform a geographical analysis of the restaurants in the dataset.

```
In [3]: # Import Libraries
      import pandas as pd
      import matplotlib.pyplot as plt
      import folium
In [2]: pip install folium
      Collecting folium
        Obtaining dependency information for folium from https://files.pythonhosted.org/packages/87/1f/b053
      4aa3c537bc4477f371de1966b92c041816e5f943dcf73298c1f1b8a5/folium-0.19.3-py2.py3-none-any.whl.metadata
      (https://files.pythonhosted.org/packages/87/1f/b0534aa3c537bc4477f371de1966b92c041816e5f943dcf73298c1
      f1b8a5/folium-0.19.3-py2.py3-none-any.whl.metadata)
        Downloading folium-0.19.3-py2.py3-none-any.whl.metadata (3.8 kB)
      Collecting branca>=0.6.0 (from folium)
        Obtaining dependency information for branca>=0.6.0 from https://files.pythonhosted.org/packages/f
      8/9d/91cddd38bd00170aad1a4b198c47b4ed716be45c234e09b835af41f4e717/branca-0.8.1-py3-none-any.whl.metad
      ata (https://files.pythonhosted.org/packages/f8/9d/91cddd38bd00170aad1a4b198c47b4ed716be45c234e09b835
      af41f4e717/branca-0.8.1-py3-none-any.whl.metadata)
        Downloading branca-0.8.1-py3-none-any.whl.metadata (1.5 kB)
      Requirement already satisfied: jinja2>=2.9 in c:\users\shejal sanas\anaconda3\lib\site-packages (from
      Requirement already satisfied: numpy in c:\users\shejal sanas\anaconda3\lib\site-packages (from foliu
      Requirement already satisfied: requests in c:\users\shejal sanas\anaconda3\lib\site-packages (from fo
      lium) (2.31.0)
      Requirement already satisfied: xyzservices in c:\users\shejal sanas\anaconda3\lib\site-packages (from
      folium) (2022.9.0)
      Requirement already satisfied: MarkupSafe>=2.0 in c:\users\shejal sanas\anaconda3\lib\site-packages
      (from jinja2>=2.9->folium) (2.1.1)
      Requirement already satisfied: charset-normalizer<4,>=2 in c:\users\shejal sanas\anaconda3\lib\site-p
      ackages (from requests->folium) (2.0.4)
      Requirement already satisfied: idna<4,>=2.5 in c:\users\shejal sanas\anaconda3\lib\site-packages (fro
      m requests->folium) (3.4)
      Requirement already satisfied: urllib3<3,>=1.21.1 in c:\users\shejal sanas\anaconda3\lib\site-package
      s (from requests->folium) (1.26.16)
      Requirement already satisfied: certifi>=2017.4.17 in c:\users\shejal sanas\anaconda3\lib\site-package
      s (from requests->folium) (2023.7.22)
      Downloading folium-0.19.3-py2.py3-none-any.whl (110 kB)
         ----- 0.0/110.5 kB ? eta -:--:--
         ------ 20.5/110.5 kB ? eta -:--:--
           ---- ------------------------- 20.5/110.5 kB ? eta -:--:--
         ------ 20.5/110.5 kB ? eta -:--:--
         ------ 41.0/110.5 kB 217.9 kB/s eta 0:00:01
         ------ 41.0/110.5 kB 217.9 kB/s eta 0:00:01
         ------ 61.4/110.5 kB 218.8 kB/s eta 0:00:01
         ------ 61.4/110.5 kB 218.8 kB/s eta 0:00:01
         ----- 92.2/110.5 kB 275.8 kB/s eta 0:00:01
                  ----- 92.2/110.5 kB 275.8 kB/s eta 0:00:01
          ----- 92.2/110.5 kB 275.8 kB/s eta 0:00:01
         ----- 92.2/110.5 kB 275.8 kB/s eta 0:00:01
         ----- 92.2/110.5 kB 275.8 kB/s eta 0:00:01
         ----- 92.2/110.5 kB 275.8 kB/s eta 0:00:01
         ----- 92.2/110.5 kB 275.8 kB/s eta 0:00:01
         ----- 92.2/110.5 kB 275.8 kB/s eta 0:00:01
         ----- 92.2/110.5 kB 275.8 kB/s eta 0:00:01
         ----- 92.2/110.5 kB 275.8 kB/s eta 0:00:01
         ----- 92.2/110.5 kB 275.8 kB/s eta 0:00:01
         ----- 92.2/110.5 kB 275.8 kB/s eta 0:00:01
         ----- -- 102.4/110.5 kB 88.0 kB/s eta 0:00:01
         ------ 110.5/110.5 kB 93.0 kB/s eta 0:00:00
```

Downloading branca-0.8.1-py3-none-any.whl (26 kB) Installing collected packages: branca, folium Successfully installed branca-0.8.1 folium-0.19.3

Note: you may need to restart the kernel to use updated packages.

In [4]: # Reading the DataFrame
df = pd.read_csv(r'C:/Users/Shejal Sanas/Downloads/Dataset.csv')

In [5]: df.head(5)

Out[5]:

| | Restaurant ID | Restaurant Name | Country Code | City | Address | Locality | Locality Verbose | Longitude | Latitude | Cuisines | (| |
|---|------------------|------------------------------|-----------------|---------------------|---|--|---|------------|-----------|---|---|--|
| 0 | 6317637 | Le Petit Souffle | 162 | Makati City | Third Floor, Century City Mall, Kalayaan Avenu | Century City Mall, Poblacion, Makati City | Century City Mall, Poblacion, Makati City, Mak | 121.027535 | 14.565443 | French, Japanese, Desserts | | |
| 1 | 6304287 | lzakaya Kikufuji | 162 | Makati City | Little Tokyo, 2277 Chino Roces Avenue, Legaspi | Little Tokyo, Legaspi Village, Makati City | Little Tokyo, Legaspi Village, Makati City, Ma | 121.014101 | 14.553708 | Japanese | | |
| 2 | 6300002 | Heat - Edsa Shangri-La | 162 | Mandaluyong City | Edsa Shangri- La, 1 Garden Way, Ortigas, Mandal | Edsa Shangri-La, Ortigas, Mandaluyong City | Edsa Shangri-La, Ortigas, Mandaluyong City, Ma | 121.056831 | 14.581404 | Seafood, Asian, Filipino, Indian | | |
| 3 | 6318506 | Ooma | 162 | Mandaluyong City | Third Floor, Mega Fashion Hall, SM Megamall, O | SM Megamall, Ortigas, Mandaluyong City | SM Megamall, Ortigas, Mandaluyong City, Mandal | 121.056475 | 14.585318 | Japanese, Sushi | | |
| 4 | 6314302 | Sambo Kojin | 162 | Mandaluyong City | Third Floor, Mega Atrium, SM Megamall, Ortigas | SM Megamall, Ortigas, Mandaluyong City | SM Megamall, Ortigas, Mandaluyong City, Mandal | 121.057508 | 14.584450 | Japanese, Korean | | |

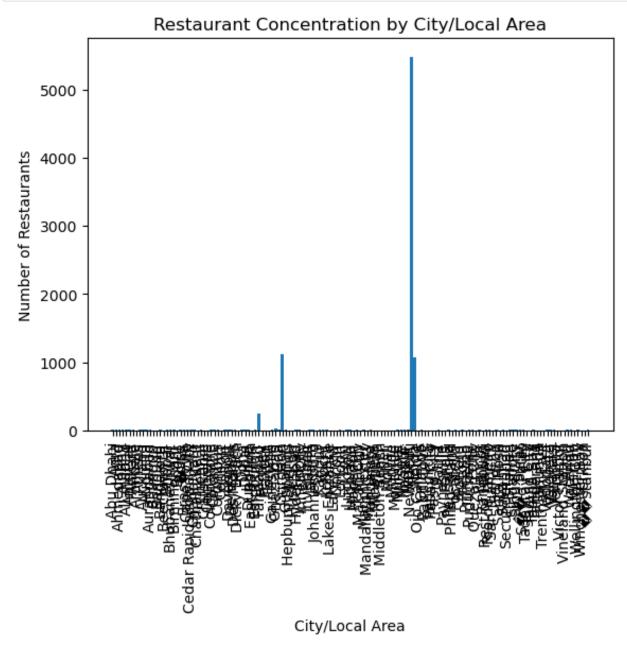
5 rows × 21 columns

```
In [6]: # Using Group by function, grouping the City Column
grouped_by_city = df.groupby('City')
```

```
In [7]: # Count of Restaurant
restaurant_count = grouped_by_city['Restaurant Name'].count()
```

```
In [8]: import matplotlib.pyplot as plt

plt.bar(restaurant_count.index, restaurant_count.values)
plt.xlabel('City/Local Area')
plt.ylabel('Number of Restaurants')
plt.title('Restaurant Concentration by City/Local Area')
plt.xticks(rotation=90)
plt.show()
```

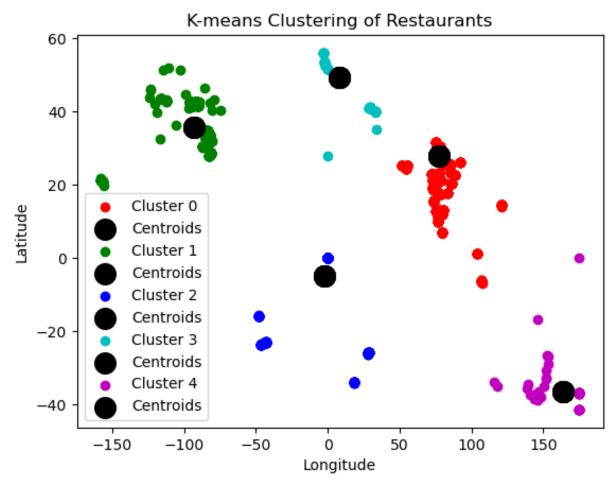


```
In [9]: # Forming into Clusters
from sklearn.cluster import KMeans

X = df[['Latitude', 'Longitude']]
k = 5
kmeans = KMeans(n_clusters=k, random_state=0)
cluster_labels = kmeans.fit_predict(X)
df['Cluster'] = cluster_labels
```

C:\Users\Shejal Sanas\anaconda3\Lib\site-packages\sklearn\cluster_kmeans.py:1412: FutureWarning: The default value of `n_init` will change from 10 to 'auto' in 1.4. Set the value of `n_init` explicitly to suppress the warning

super()._check_params_vs_input(X, default_n_init=10)

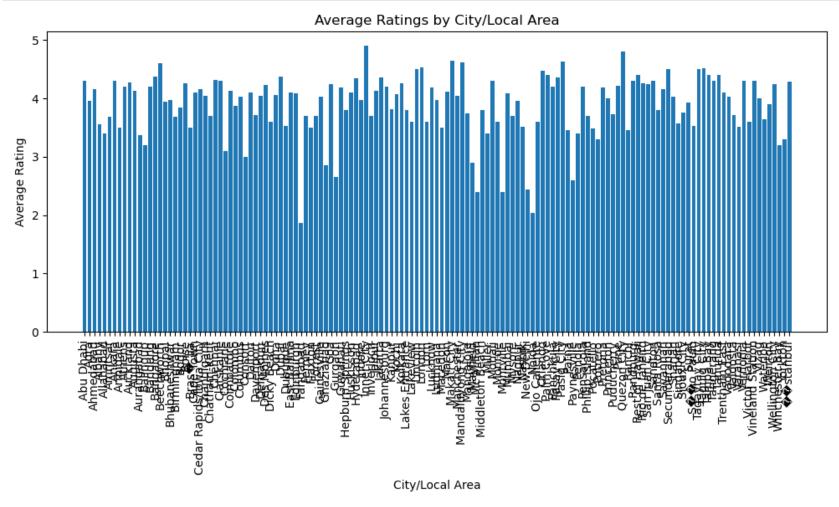


Average Ratings by City/Locality

```
In [11]:
         average_ratings_by_city = grouped_by_city['Aggregate rating'].mean()
         print(average_ratings_by_city)
         City
         Abu Dhabi
                            4.300000
         Agra
                            3.965000
         Ahmedabad
                            4.161905
         Albany
                            3.555000
         Allahabad
                            3.395000
         Weirton
                            3.900000
                            4.250000
         Wellington City
         Winchester Bay
                            3.200000
         Yorkton
                            3.300000
         ♦♦stanbul
                              4.292857
         Name: Aggregate rating, Length: 141, dtype: float64
```

```
In [12]:
    plt.figure(figsize=(10, 6))

    plt.bar(average_ratings_by_city.index, average_ratings_by_city.values)
    plt.xlabel('City/Local Area')
    plt.ylabel('Average Rating')
    plt.title('Average Ratings by City/Local Area')
    plt.xticks(rotation=90) # Rotate x-axis labels for better readability
    plt.tight_layout()
    plt.show()
```



Popular Cuisines by City/Locality

```
In [13]: popular_cuisines_by_city = grouped_by_city['Cuisines'].agg(lambda x: x.mode().tolist())
         print(popular_cuisines_by_city)
         City
         Abu Dhabi
                                            [American, Indian, Italian, Pizza]
         Agra
                                                       [North Indian, Mughlai]
                             [Cafe, American, Continental, Armenian, Fast F...
         Ahmedabad
         Albany
                                                      [Japanese, Steak, Sushi]
         Allahabad
                                                       [North Indian, Chinese]
         Weirton
                                                     [Burger, Greek, Sandwich]
         Wellington City
                                                                         [Cafe]
         Winchester Bay
                                                      [Burger, Seafood, Steak]
         Yorkton
                                                                        [Asian]
         ♦♦stanbul
                                                                          [Cafe]
         Name: Cuisines, Length: 141, dtype: object
In [14]: plt.figure(figsize=(10, 6))
         for city, cuisines in popular_cuisines_by_city.items():
             plt.bar(city, ', '.join(cuisines))
             plt.xlabel('City/Local Area')
             plt.ylabel('Popular Cuisines')
             plt.title('Popular Cuisines by City/Local Area')
             plt.xticks(rotation=90) # Rotate x-axis labels for better readability
             plt.tight_layout()
         plt.show()
         C:\Users\Shejal Sanas\AppData\Local\Temp\ipykernel_27988\3718216243.py:9: UserWarning: Tight layout n
         ot applied. The left and right margins cannot be made large enough to accommodate all axes decoration
           plt.tight_layout()
```

Price Range Analysis by City/Locality

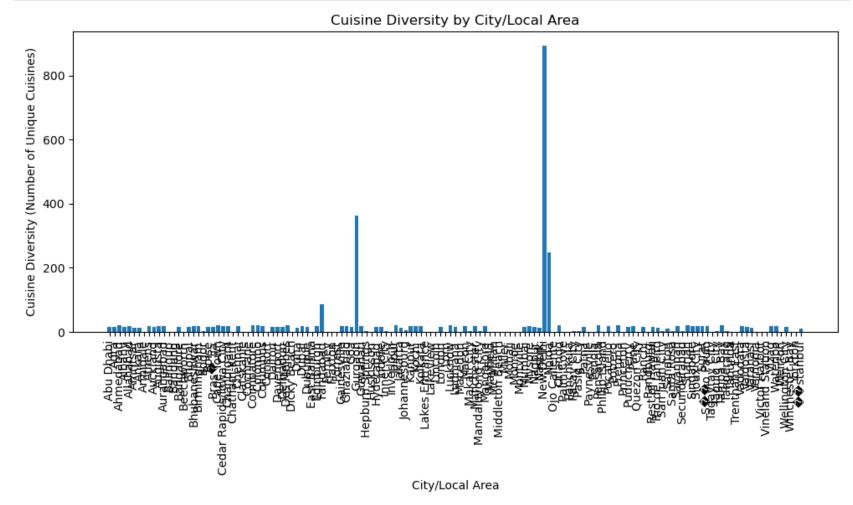
```
In [15]: common_price_range_by_city = grouped_by_city['Price range'].agg(lambda x: x.mode().iloc[0])
         print(common_price_range_by_city)
         City
         Abu Dhabi
                             4
         Agra
                            2
                            3
         Ahmedabad
                            1
         Albany
         Allahabad
                            3
         Weirton
                            2
                            4
         Wellington City
                            2
         Winchester Bay
         Yorkton
                             2
         ♦♦stanbul
         Name: Price range, Length: 141, dtype: int64
In [16]: plt.figure(figsize=(50, 6))
         plt.bar(common_price_range_by_city.index, common_price_range_by_city.values)
         plt.xlabel('City/Local Area')
         plt.ylabel('Common Price Range')
         plt.title('Common Price Range by City/Local Area')
         plt.xticks(rotation=90)
         plt.tight_layout()
         plt.show()
```

Cuisine Diversity

```
In [17]:
         grouped_by_city = df.groupby('City')
         cuisine_diversity = grouped_by_city['Cuisines'].apply(lambda x: len(set(x)))
         print(cuisine_diversity)
         City
                             17
         Abu Dhabi
                             15
         Agra
         Ahmedabad
                             21
                             17
         Albany
         Allahabad
                             18
                             . .
         Weirton
                             1
         Wellington City
                             17
         Winchester Bay
                             1
         Yorkton
                              1
         ♦♦stanbul
                              11
         Name: Cuisines, Length: 141, dtype: int64
```

```
In [18]: plt.figure(figsize=(10, 6))

plt.bar(cuisine_diversity.index, cuisine_diversity.values)
plt.xlabel('City/Local Area')
plt.ylabel('Cuisine Diversity (Number of Unique Cuisines)')
plt.title('Cuisine Diversity by City/Local Area')
plt.xticks(rotation=90)
plt.tight_layout()
plt.show()
```



Conclusion: Performed a geographical analysis of the restaurants based on Average Ratings, Cuisine Ratings and Cuisines Diversity by City/Locality.