

Perform a geographical analysis of the restaurants in the dataset.

```
In [2]: pip install folium
```

```
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	Izakaya 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()
```

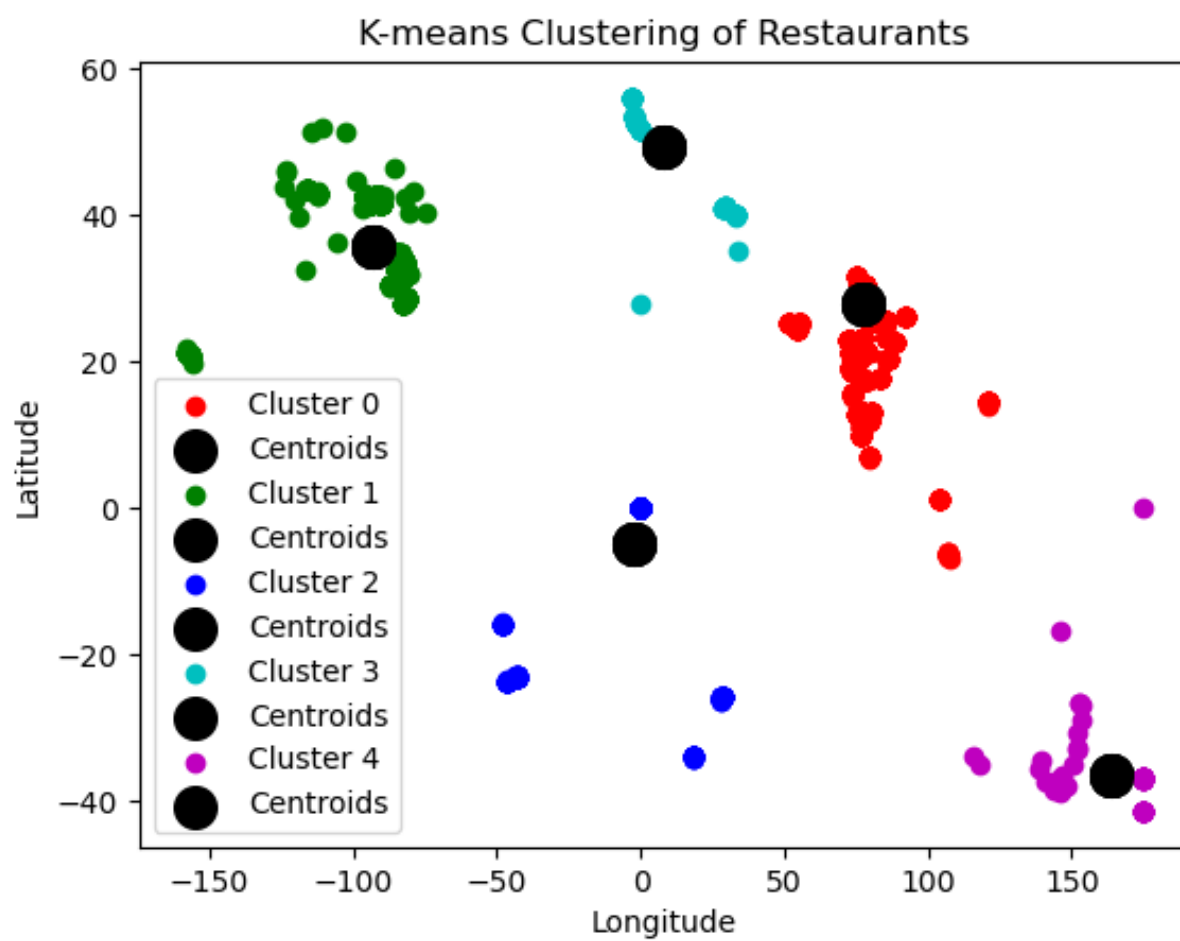
Restaurant Concentration by City/Local Area

City/Local Area	Number of Restaurants (Approximate)
Aberdeen	~10
Aldershot	~10
Alford	~10
Amble	~10
Ankney	~10
Aurora	~10
Barnes	~10
Birmingham	~10
Blyth	~10
Cedar Rapids	~10
Chesham	~10
Colchester	~10
Dorset	~10
Early	~300
Epsom	~10
Gillingham	~10
Hepburn	~1100
Hull	~10
Johannesburg	~10
Lakeside	~10
Mandarin	~10
Middleton	~10
Milton Keynes	~10
Norwich	~10
Oak Ridge	~5500
Peterborough	~10
Philly	~10
Rochester	~10
Sutton	~10
Telford	~10
Trenton	~10
Vancouver	~10
Walsingham	~10
Westminster	~10

```
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)
```

```
In [10]: colors = ['r', 'g', 'b', 'c', 'm', 'y', 'k']

for cluster_num in range(k):
    cluster_data = df[df['Cluster'] == cluster_num]
    plt.scatter(cluster_data['Longitude'], cluster_data['Latitude'],
                c=colors[cluster_num], label=f'Cluster {cluster_num}')
    plt.scatter(kmeans.cluster_centers[:, 1], kmeans.cluster_centers[:, 0],
                s=200, c='black', label='Centroids')
    plt.xlabel('Longitude')
    plt.ylabel('Latitude')
    plt.title('K-means Clustering of Restaurants')
    plt.legend()
plt.show()
```



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
Wellington City 4.250000
Winchester Bay 3.200000
Yorkton        3.300000
📍📍istanbul    4.292857
Name: Aggregate rating, Length: 141, dtype: float64
```

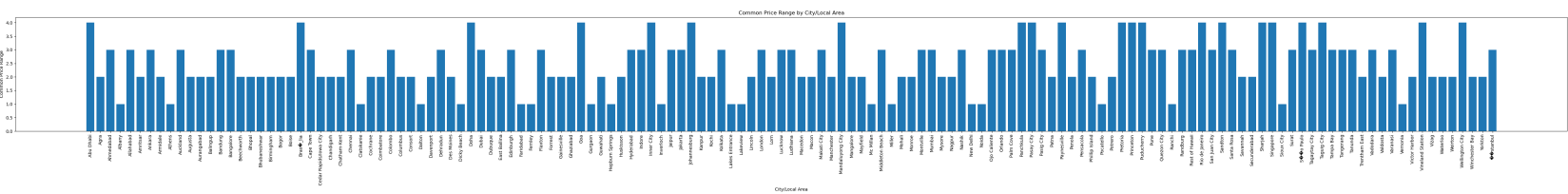

Price Range Analysis by City/Locality

```
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
Ahmedabad      3
Albany          1
Allahabad      3
..
Weirton         2
Wellington City 4
Winchester Bay  2
Yorkton         2
🇹🇷🇹🇷istanbul    3
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
Abu Dhabi      17
Agra            15
Ahmedabad      21
Albany          17
Allahabad      18
..
Weirton         1
Wellington City 17
Winchester Bay  1
Yorkton         1
🇹🇷🇹🇷istanbul    11
Name: Cuisines, Length: 141, dtype: int64
```


[illegible]

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