

CITY ANALYSIS

IMPORTING LIBRARIES

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
```

DATA COLLECTION

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

```
In [3]: df.head()
```

Out[3]:

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Locality Verbose	Longitude	Latitude	Cuisines	...	C
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	...	B
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	...	B
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	...	B
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	...	B
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	...	B

5 rows × 21 columns

```
In [4]: df.shape
```

```
Out[4]: (9551, 21)
```

```
In [5]: df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9551 entries, 0 to 9550
Data columns (total 21 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   i»¿Restaurant ID                      9551 non-null   int64
1   Restaurant Name                       9551 non-null   object
2   Country Code                          9551 non-null   int64
3   City                                  9551 non-null   object
4   Address                               9551 non-null   object
5   Locality                              9551 non-null   object
6   Locality Verbose                      9551 non-null   object
7   Longitude                             9551 non-null   float64
8   Latitude                             9551 non-null   float64
9   Cuisines                              9542 non-null   object
10  Average Cost for two                  9551 non-null   int64
11  Currency                             9551 non-null   object
12  Has Table booking                     9551 non-null   object
13  Has Online delivery                   9551 non-null   object
14  Is delivering now                     9551 non-null   object
15  Switch to order menu                  9551 non-null   object
16  Price range                           9551 non-null   int64
17  Aggregate rating                      9551 non-null   float64
18  Rating color                          9551 non-null   object
19  Rating text                           9551 non-null   object
20  Votes                                9551 non-null   int64
dtypes: float64(3), int64(5), object(13)
memory usage: 1.5+ MB

```

TO CHECK NULL VALUES

```
In [6]: pd.isnull(df).sum()
```

```

Out[6]: i»¿Restaurant ID      0
        Restaurant Name      0
        Country Code         0
        City                  0
        Address               0
        Locality              0
        Locality Verbose      0
        Longitude             0
        Latitude              0
        Cuisines              9
        Average Cost for two  0
        Currency              0
        Has Table booking     0
        Has Online delivery   0
        Is delivering now     0
        Switch to order menu  0
        Price range           0
        Aggregate rating      0
        Rating color          0
        Rating text           0
        Votes                 0
dtype: int64

```

TO DELETE NULL VALUES

```
In [7]: df.dropna(inplace=True)
```

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

```
Out[9]: (9542, 21)
```

```
In [10]: df.describe()
```

Out[10]:

	ï»¿Restaurant ID	Country Code	Longitude	Latitude	Average Cost for two	Price range	Aggregate rating	Votes
count	9.542000e+03	9542.000000	9542.000000	9542.000000	9542.000000	9542.000000	9542.000000	9542.000000
mean	9.043301e+06	18.179208	64.274997	25.848532	1200.326137	1.804968	2.665238	156.772060
std	8.791967e+06	56.451600	41.197602	11.010094	16128.743876	0.905563	1.516588	430.203324
min	5.300000e+01	1.000000	-157.948486	-41.330428	0.000000	1.000000	0.000000	0.000000
25%	3.019312e+05	1.000000	77.081565	28.478658	250.000000	1.000000	2.500000	5.000000
50%	6.002726e+06	1.000000	77.192031	28.570444	400.000000	2.000000	3.200000	31.000000
75%	1.835260e+07	1.000000	77.282043	28.642711	700.000000	2.000000	3.700000	130.000000
max	1.850065e+07	216.000000	174.832089	55.976980	800000.000000	4.000000	4.900000	10934.000000

IDENTIFY THE CITY WITH THE HIGHEST NUMBER OF RESTAURANTS IN THE DATASET.

```
In [11]: df.columns

Out[11]: Index(['ï»¿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 [15]: # Group the data by city and count the number of restaurants in each city
city_restaurants_count = df.groupby('City')['Restaurant Name'].count()

In [16]: # Find the city with the highest number of restaurants
city_with_most_restaurants = city_restaurants_count.idxmax()
num_restaurants = city_restaurants_count.max()

In [17]: print(f"The city with the highest number of restaurants is {city_with_most_restaurants} with {num_restaurants}")

The city with the highest number of restaurants is New Delhi with 5473 restaurants.
```

CALCULATE THE AVERAGE RATING FOR RESTAURANTS IN EACH CITY.

```
In [18]: df.columns

Out[18]: Index(['ï»¿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 [46]: df.dtypes

Out[46]: ï»¿Restaurant ID      int64
Restaurant Name      object
Country Code        int64
dtype: object

In [47]: df=pd.read_csv('Dataset .csv',encoding='unicode_escape')

In [49]: df.dtypes
```

```
Out[49]: i»Restaurant ID          int64
Restaurant Name          object
Country Code             int64
City                     object
Address                  object
Locality                 object
Locality Verbose         object
Longitude                float64
Latitude                 float64
Cuisines                 object
Average Cost for two     int64
Currency                 object
Has Table booking        object
Has Online delivery      object
Is delivering now        object
Switch to order menu     object
Price range              int64
Aggregate rating         float64
Rating color             object
Rating text              object
Votes                   int64
dtype: object
```

```
In [50]: # Group the data by city and calculate the average rating for restaurants in each city
city_avg_rating = df.groupby('City')['Aggregate rating'].mean()

print("Average ratings for restaurants in each city:")
print(city_avg_rating)
```

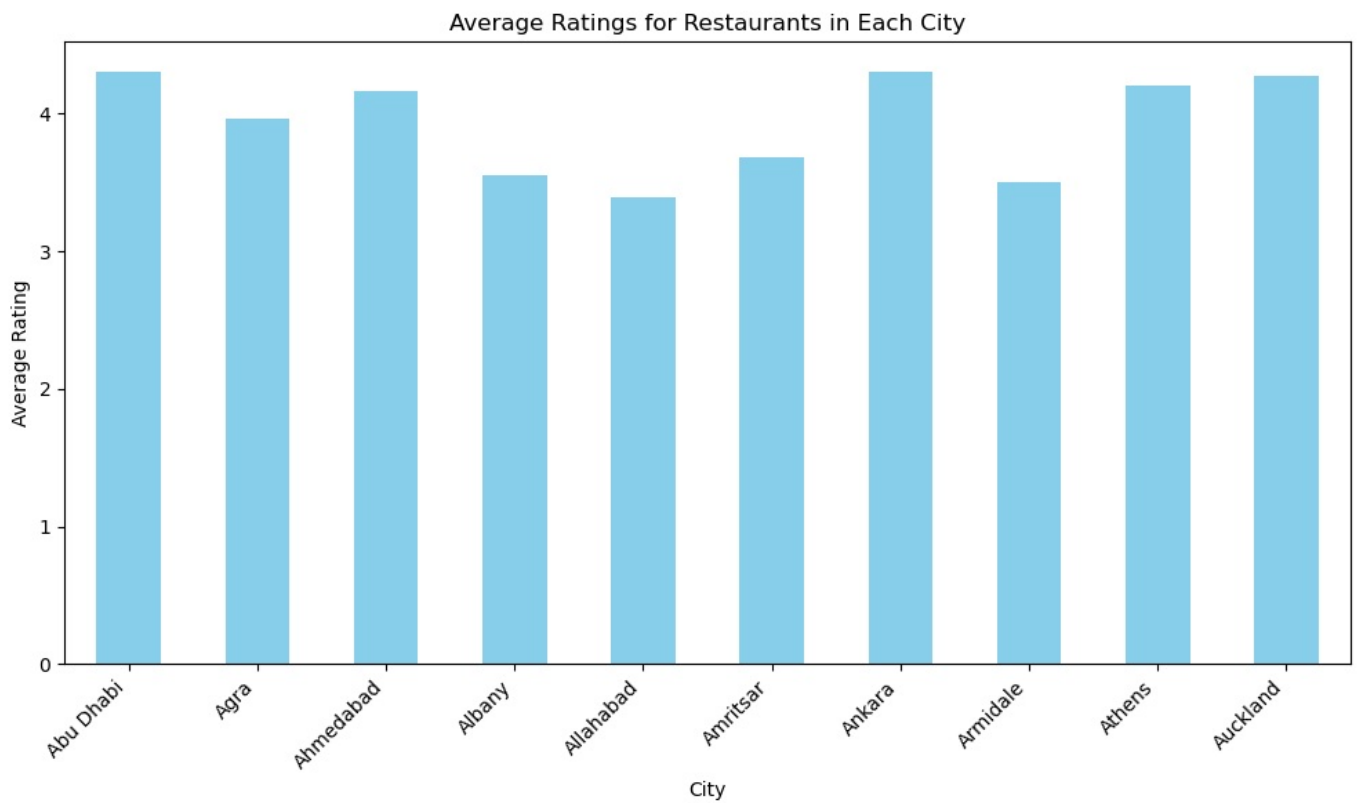
Average ratings for restaurants in each city:

```
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
İstanbul           4.292857
```

Name: Aggregate rating, Length: 141, dtype: float64

```
In [53]: city_avg_rating = df.groupby('City')['Aggregate rating'].mean().head(10)

# Plot the data
plt.figure(figsize=(10, 6))
city_avg_rating.plot(kind='bar', color='skyblue')
plt.title('Average Ratings for Restaurants in Each City')
plt.xlabel('City')
plt.ylabel('Average Rating')
plt.xticks(rotation=45, ha='right')
plt.tight_layout()
plt.show()
```



DETERMINE THE CITY WITH THE HIGHEST AVERAGE RATING.

```
In [54]: df.columns
```

```
Out[54]: 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 [55]: # Group the data by city and calculate the average rating for restaurants in each city
city_avg_rating = df.groupby('City')['Aggregate rating'].mean()

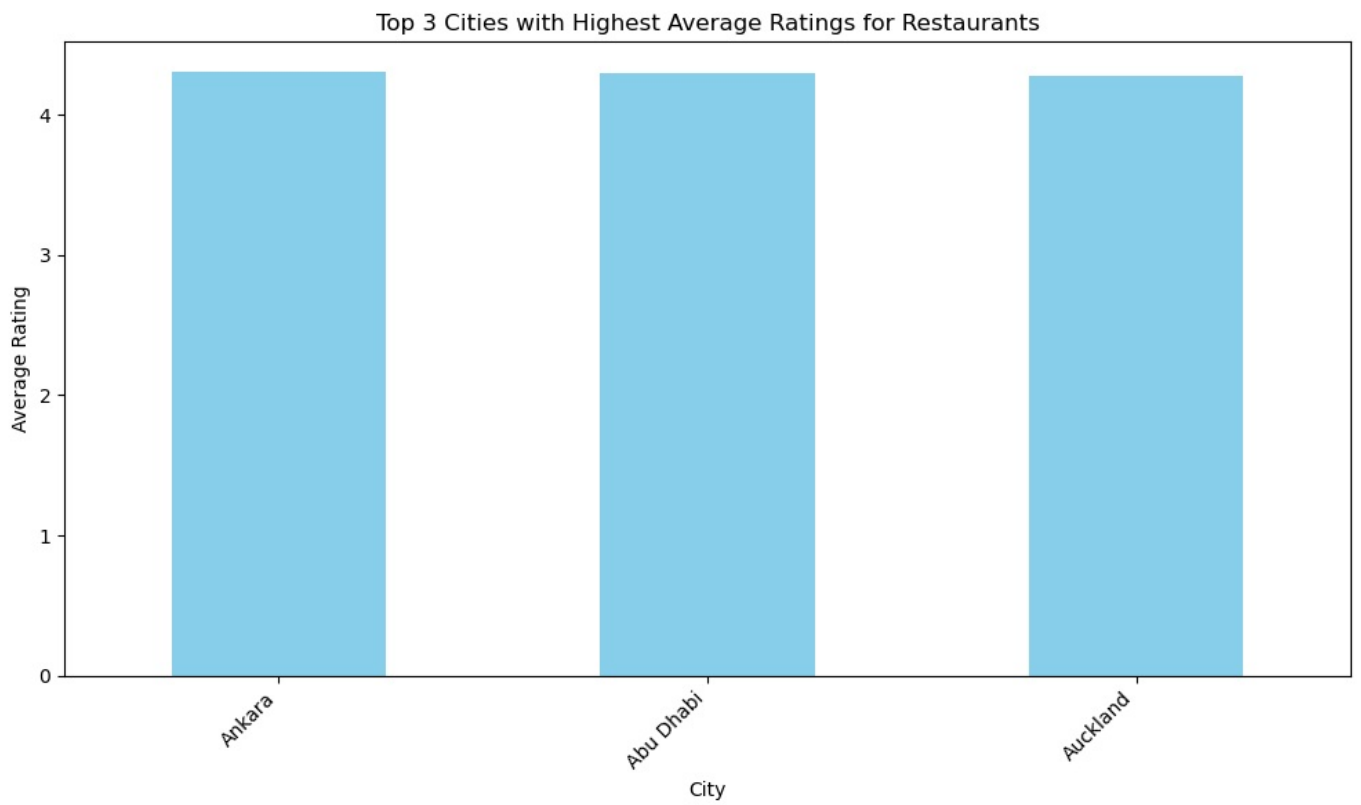
# Determine the city with the highest average rating
highest_avg_rating_city = city_avg_rating.idxmax()
highest_avg_rating = city_avg_rating.max()

print(f"The city with the highest average rating is {highest_avg_rating_city} with an average rating of {highest_avg_rating}")
```

The city with the highest average rating is Inner City with an average rating of 4.90.

```
In [58]: # Sort the cities by average rating in descending order and select the top 3
top_3_cities = city_avg_rating.sort_values(ascending=False).head(3)

# Plot the data for top 3 cities
plt.figure(figsize=(10, 6))
top_3_cities.plot(kind='bar', color='skyblue')
plt.title('Top 3 Cities with Highest Average Ratings for Restaurants')
plt.xlabel('City')
plt.ylabel('Average Rating')
plt.xticks(rotation=45, ha='right')
plt.tight_layout()
plt.show()
```



THANKYOU!

CONNECT WITH ME:

LinkedIn: <https://www.linkedin.com/in/harshita-sharma-b68154220/>

GitHub: <https://github.com/DATAPREDICTS>

Instagram: https://www.instagram.com/datapredicts?utm_source=qr&igsh=czVzc2k5c3oxOWQ4

YouTube: <https://youtube.com/@Datapredicts?si=eDKAqVciVxg23zab>