

Task: Descriptive Analysis

Importing Libraries

```
In [1]: import pandas as pd
import statistics as ST
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [3]: df=pd.read_csv('N:\Dataset .csv')
```

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

```
Out[4]: 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')
```

Data characteristics

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

Out[5]:

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Locality Verbose	Longitu
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...	121.0275
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...	121.0141
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...	121.0568
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...	121.0564
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...	121.0575

5 rows × 21 columns

```
In [6]: df.tail()
```

Out[6]:

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Location Ver
9546	5915730	Namlı Gurme	208	İstanbul	Kemankeş Karamustafa Paşa Mahallesi, Rıhtım ...	Karaköy	Karaköy
9547	5908749	Açık Ceviz	208	İstanbul	Koşuyolu Mahallesi, Muhittin ...	Koşuyolu	Koşuyolu
9548	5915807	Huqqa	208	İstanbul	Kuruçeşme Mahallesi, Muallim Naci Caddesi, N...	Kuruçeşme	Kuruçeşme
9549	5916112	Açık Kahve	208	İstanbul	Kuruçeşme Mahallesi, Muallim Naci Caddesi, N...	Kuruçeşme	Kuruçeşme
9550	5927402	Walter's Coffee Roastery	208	İstanbul	Cafea Mahallesi, Bademaltı Sokak, No 21/B, ...	Moda	Moda

5 rows × 21 columns

```
In [7]: df.describe()
```

Out[7]:

	Restaurant ID	Country Code	Longitude	Latitude	Average Cost for two	Price range	Aggr
count	9.551000e+03	9551.000000	9551.000000	9551.000000	9551.000000	9551.000000	9551.0
mean	9.051128e+06	18.365616	64.126574	25.854381	1199.210763	1.804837	2.6
std	8.791521e+06	56.750546	41.467058	11.007935	16121.183073	0.905609	1.5
min	5.300000e+01	1.000000	-157.948486	-41.330428	0.000000	1.000000	0.0
25%	3.019625e+05	1.000000	77.081343	28.478713	250.000000	1.000000	2.5
50%	6.004089e+06	1.000000	77.191964	28.570469	400.000000	2.000000	3.2
75%	1.835229e+07	1.000000	77.282006	28.642758	700.000000	2.000000	3.7
max	1.850065e+07	216.000000	174.832089	55.976980	800000.000000	4.000000	4.9

Calculate basic statistical measures (mean,median, standard deviation, etc.) for

numerical columns

```
In [8]: print(df['Country Code'].mean())  
print(df['Country Code'].median())  
print(ST.stdev(df['Country Code']))
```

```
18.365616165846507  
1.0  
56.75054560094657
```

```
In [9]: print(df['Longitude'].mean())  
print(df['Longitude'].median())  
print(ST.stdev(df['Longitude']))
```

```
64.12657446168706  
77.1919642  
41.46705784761728
```

```
In [10]: print(df['Latitude'].mean())  
print(df['Latitude'].median())  
print(ST.stdev(df['Latitude']))
```

```
25.854380700074756  
28.57046888  
11.007935124784668
```

```
In [11]: print(df['Average Cost for two'].mean())  
print(df['Average Cost for two'].median())  
print(ST.stdev(df['Average Cost for two']))
```

```
1199.2107632708617  
400.0  
16121.183073499646
```

```
In [12]: print(df['Price range'].mean())  
print(df['Price range'].median())  
print(ST.stdev(df['Price range']))
```

```
1.804837189823055  
2.0  
0.905608847397614
```

```
In [13]: print(df['Aggregate rating'].mean())  
print(df['Aggregate rating'].median())  
print(ST.stdev(df['Aggregate rating']))
```

```
2.66637001361114  
3.2  
1.5163775396521326
```

```
In [14]: print(df['Votes'].mean())  
print(df['Votes'].median())  
print(ST.stdev(df['Votes']))
```

```
156.909747670401  
31.0  
430.1691453762912
```

Explore the distribution of categorical variables

s

```
In [15]: top_cuis=df['Cuisines'].value_counts()  
print(top_cuis.head())
```

```
Cuisines  
North Indian          936  
North Indian, Chinese  511  
Chinese               354  
Fast Food             354  
North Indian, Mughlai  334  
Name: count, dtype: int64
```

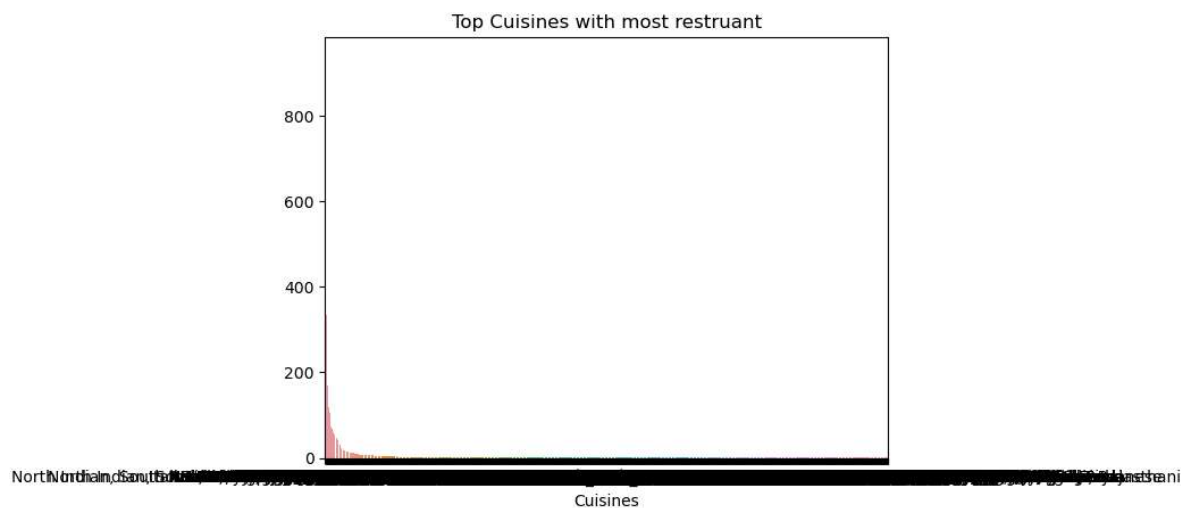
```
In [16]: top_city=df['City'].value_counts()  
print(top_city.head())
```

```
City  
New Delhi    5473  
Gurgaon      1118  
Noida        1080  
Faridabad    251  
Ghaziabad    25  
Name: count, dtype: int64
```

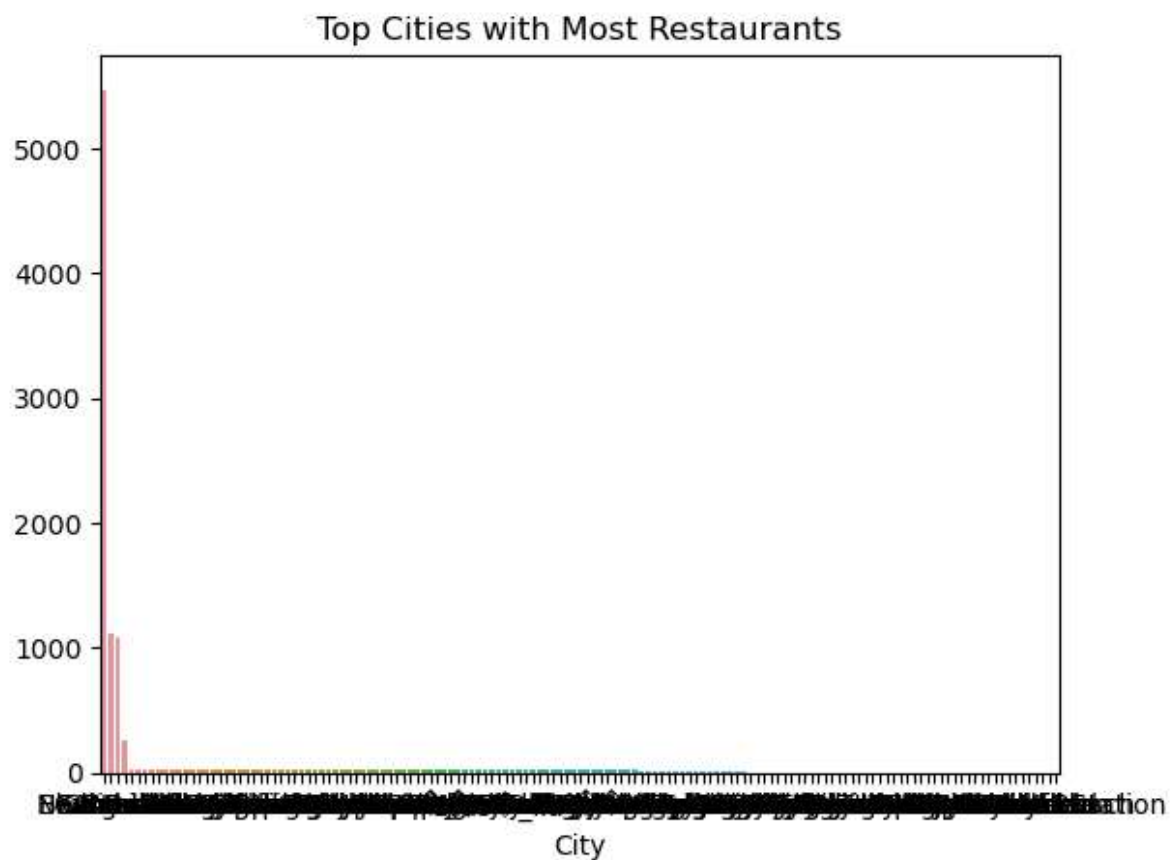
```
In [17]: top_restu=df['Restaurant Name'].value_counts()  
print(top_restu.head())
```

```
Restaurant Name  
Cafe Coffee Day    83  
Domino's Pizza     79  
Subway             63  
Green Chick Chop   51  
McDonald's         48  
Name: count, dtype: int64
```

```
In [21]: sns.barplot(x=top_cuis.index ,y=top_cuis.values)
plt.title('Top Cuisines with most restruant')
plt.show()
```



```
In [19]: sns.barplot(x=top_city.index, y=top_city.values)
plt.title('Top Cities with Most Restaurants')
plt.show()
```



In []: