```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
In [2]: path=r"C:\Users\Sruth\Downloads\cognifyz dataset.csy"
```

In [2]: path=r"C:\Users\Sruth\Downloads\cognifyz dataset.csv"
 df=pd.read\_csv(path)
 df

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality
0	6317637	Le Petit Souffle	162	Makati City	Third Floor, Century City Mall, Kalayaan Avenu	Century City Mall, Poblacion, Makati City
1	6304287	Izakaya Kikufuji	162	Makati City	Little Tokyo, 2277 Chino Roces Avenue, Legaspi	Little Tokyo, Legaspi Village, Makati City
2	6300002	Heat - Edsa Shangri-La	162	Mandaluyong City	Edsa Shangri-La, 1 Garden Way, Ortigas, Mandal	Edsa Shangri- La, Ortigas, Mandaluyong City
3	6318506	Ooma	162	Mandaluyong City	Third Floor, Mega Fashion Hall, SM Megamall, O	SM Megamall, Ortigas, Mandaluyong City
4	6314302	Sambo Kojin	162	Mandaluyong City	Third Floor, Mega Atrium, SM Megamall, Ortigas	SM Megamall, Ortigas, Mandaluyong City
•••						
9546	5915730	Namll Gurme	208	<b>��</b> stanbul	Kemanke�� Karamustafa Pa��a Mahallesi, Rlhtlm	Karak <b>∳</b> _y
9547	5908749	Ceviz A��acl	208	<b>��</b> stanbul	Ko��uyolu Mahallesi, Muhittin ��st�_nda�� Cadd	Ko��uyolu
9548	5915807	Huqqa	208	<b>��</b> stanbul	Kuru�_e��me Mahallesi, Muallim Naci Caddesi, N	Kuru <b>∳</b> _e��me
9549	5916112	A���k Kahve	208	<b>�</b> �stanbul	Kuru�_e��me Mahallesi, Muallim Naci Caddesi, N	Kuru�_e��me
9550	5927402	Walter's Coffee Roastery	208	<b>� �</b> stanbul	Cafea��a Mahallesi, Bademaltl Sokak, No 21/B, 	Moda

Determine the top three most common cuisines in the dataset.

```
In [3]:
        df[['Cuisines']].head()
Out[3]:
                               Cuisines
         0
                French, Japanese, Desserts
         1
                               Japanese
         2
            Seafood, Asian, Filipino, Indian
         3
                         Japanese, Sushi
         4
                        Japanese, Korean
In [4]:
        df[['Cuisines']].tail()
Out[4]:
                                  Cuisines
         9546
                                   Turkish
         9547 World Cuisine, Patisserie, Cafe
         9548
                       Italian, World Cuisine
         9549
                            Restaurant Cafe
         9550
                                      Cafe
In [5]:
        Total=df['Cuisines'].str.split(',').explode()
         Total
Out[5]: 0
                            French
                          Japanese
         0
                         Desserts
         1
                         Japanese
                          Seafood
         2
         9547
                              Cafe
                          Italian
         9548
                    World Cuisine
         9548
         9549
                  Restaurant Cafe
         9550
         Name: Cuisines, Length: 19719, dtype: object
In [6]: top_3_cusines=Total.value_counts().head(3)
         top_3_cusines
Out[6]: Cuisines
         North Indian
                          2992
          Chinese
                          1880
                          1314
          Fast Food
         Name: count, dtype: int64
         pd.DataFrame(top_3_cusines)
```

```
North Indian
                        2992
                        1880
              Chinese
             Fast Food
                        1314
         Calculate the percentage of restaurants that serve each of the top cusines
 In [8]: # Calculate the percentage of restaurants that serve each of the top cuisines
         top_3_percentage = (top_3_cusines / len(df)) * 100
         top_3_percentage
 Out[8]: Cuisines
          North Indian 31.326563
          Chinese 19.683803
Fast Food 13.757722
          Name: count, dtype: float64
 In [9]: pd.DataFrame(top_3_percentage)
 Out[9]:
                          count
              Cuisines
          North Indian 31.326563
              Chinese 19.683803
             Fast Food 13.757722
 In [ ]:
         Level1- Task-2
         city analysis
In [10]: df.columns
Out[10]: 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 [11]: city=df['City'].value_counts()
         city
```

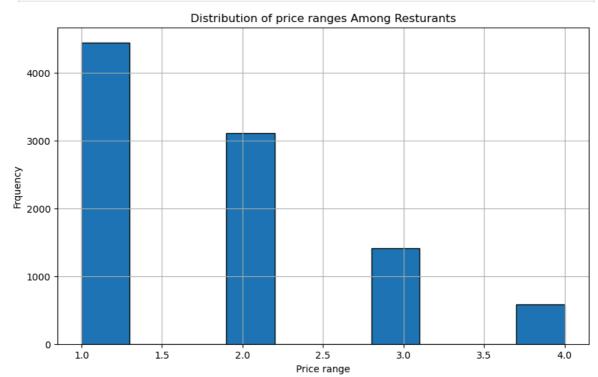
Out[7]:

count

**Cuisines** 

```
Out[11]: City
         New Delhi
                             5473
         Gurgaon
                             1118
         Noida
                             1080
         Faridabad
                             251
         Ghaziabad
                               25
                             . . .
         Panchkula
                               1
         Mc Millan
                               1
                                1
         Mayfield
         Macedon
                                1
         Vineland Station
                                1
         Name: count, Length: 141, dtype: int64
In [20]: top_city=city.idxmax()
         top_city
Out[20]: 'New Delhi'
In [17]: #identify the city with the highest number of resturants in data set
         city_counts=df['City'].value_counts()
         most_resturants_city=city_counts.idxmax()
         most_resturants_count=city_counts.max()
         print(f"The city with highest number of resturants is {most_resturants_city} wit
        The city with highest number of resturants is New Delhi with 5473 resturants.
In [23]: # calculate average rating for each city
         Average_ratings=df.groupby('City')['Aggregate rating'].mean()
         print(Average_ratings)
        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
        ♦♦stanbul
                            4.292857
        Name: Aggregate rating, Length: 141, dtype: float64
In [24]: #Determine the city with highest avg ratings.
         Average_ratings=df.groupby('City')['Aggregate rating'].mean()
         city with highest rating=Average ratings.idxmax()
         highest_rating=Average_ratings.max()
         print(f"city with the highest average rating : {city_with_highest_rating} with a
        city with the highest average rating : Inner City with an average rating 4.90
         TASK-3
In [25]: # create a histogram Distribution of price ranges Among Resturants
         price_ranges=df['Price range']
```

```
plt.figure(figsize=(10,6))
plt.hist(price_ranges,bins=10,edgecolor='black')
plt.title('Distribution of price ranges Among Resturants')
plt.xlabel('Price range')
plt.ylabel('Frquency')
plt.grid(True)
plt.show()
```



```
In [26]: # calculate the percentage of resturants in each price range category

price_ranges=df['Price range']
price_ranges_counts=price_ranges.value_counts()
total_resturants=len(price_ranges)
price_ranges_percentage=(price_ranges_counts/total_resturants)*100
print(price_ranges_percentage)
```

Price range
1 46.529159
2 32.593446
3 14.741912
4 6.135483

Name: count, dtype: float64

## **TASK-4 Online Delivery**

```
In [30]: #find count of hotels which provides online delivery
    online_delivery = df['Has Online delivery'].value_counts()
    online_delivery

Out[30]: Has Online delivery
    No     7100
    Yes    2451
    Name: count, dtype: int64

In [31]: #find total restaurants
    total_restaurants = online_delivery.sum()
    total_restaurants
```

```
Out[31]: 9551

In [32]: #find online_delivery_percentage
    online_delivery_percentage = round((online_delivery['Yes'] / total_restaurants)
    online_delivery_percentage

Out[32]: 25.66

In [33]: # find data of restaurant has online delivery
    df[df['Has Online delivery']=='Yes']
```

		Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Lc V€
	565	5704255	Famous Dave's Barbecue	214	Abu Dhabi	Near The One, Level 3, Abu Dhabi Mall, Tourist	Abu Dhabi Mall, Tourist Club Area (Al Zahiyah)	Abu Mall, <sup>*</sup> Club A Zał
	566	5701978	Pizza Di Rocco	214	Abu Dhabi	Near Corner of Salam and Al Falah Street (9th	Al Dhafrah	Al Dł Abu
	571	5700052	Cho Gao - Crowne Plaza Abu Dhabi	214	Abu Dhabi	Crowne Plaza Abu Dhabi, Sheikh Hamdan Bin Moha	Crowne Plaza Abu Dhabi, Al Markaziya	Crowne Abu Dh Marl Abu
	572	5702418	Gazebo	214	Abu Dhabi	Ground Level, Next to E- Max, Dalma Mall, Mussa	Dalma Mall, Mussafah Sanaiya	Dalma Mu Sanaiy
	573	5700386	Sangeetha Vegetarian Restaurant	214	Abu Dhabi	Opposite Cristal Hotel, Behind KM Trading, Ele	Madinat Zayed	M Zaye
	•••						•••	
910	9166	11371	Chili's	1	Pune	UG 49, Phoenix Market City, Nagar Road, Viman	Phoenix Market City, Viman Nagar	Pl Marke Viman l
	9168	18292672	Blue Water	1	Pune	Punawale, Near Basket Bridge,Off Aundh-Ravet B	Ravet	Ravet
	9170	6507967	Tales & Spirits	1	Pune	Plot 64, Shivaji Housing Society, Senapati Bap	Senapati Bapat Road	Se Bapat
	9171	6506206	18 Degrees Resto Lounge	1	Pune	8th & 9th Floor, Spot 18 Mall, Pimple Saudagar	Spot 18 Mall, Pimple Saudagar	Spot 18 F Sau
	9194	96814	Saffron Mantra	1	Secunderabad	The Purple Leaf Hotel,	Karkhana	Kar Secunde

Restaurant	Restaurant	Country	City	Addross	Locality	Lc
ID	Name	Code		Address		Ve

Karkhana, Secunderabad

2451 rows × 21 columns

In [34]: # find data of restaurant doesn't has online delivery
df[df['Has Online delivery']=='No']

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality
0	6317637	Le Petit Souffle	162	Makati City	Third Floor, Century City Mall, Kalayaan Avenu	Century City Mall, Poblacion, Makati City
1	6304287	Izakaya Kikufuji	162	Makati City	Little Tokyo, 2277 Chino Roces Avenue, Legaspi	Little Tokyo, Legaspi Village, Makati City
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4	6314302	Sambo Kojin	162	Mandaluyong City	Third Floor, Mega Atrium, SM Megamall, Ortigas	SM Megamall, Ortigas, Mandaluyong City
•••						
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9547	5908749	Ceviz A��acl	208	<b>��</b> stanbul	Ko��uyolu Mahallesi, Muhittin ��st�_nda�� Cadd	Ko��uyolu
9548	5915807	Huqqa	208	<b>��</b> stanbul	Kuru�_e��me Mahallesi, Muallim Naci Caddesi, N	Kuru�_e��me
9549	5916112	A���k Kahve	208		Kuru�_e��me Mahallesi, Muallim Naci Caddesi, N	Kuru�_e��me
9550	5927402	Walter's Coffee Roastery	208	<b>� �</b> stanbul	Cafea��a Mahallesi, Bademaltl Sokak, No 21/B, 	Moda

In [37]: # comparing the average ratings of resturants with online devlivery and without

avg\_ratings\_with\_onlinedelivery=df[df['Has Online delivery']=='Yes']['Aggregate
avg\_ratings\_without\_onlinedelivery=df[df['Has Online delivery']=='No']['Aggregat
print(f"Average rating for resturants with online delivery : {avg\_ratings\_with\_o
print(f"Average rating for resturants with out online delivery : {avg\_ratings\_wi

Average rating for resturants with online delivery : 3.25
Average rating for resturants with out online delivery : 2.47

In []: