

LEVEL1- TASK3: PRICE RANGE DISTRIBUTION

--3:1 Create a histogram or bar chart to visualize the distribution of price ranges among the restaurants.

--3:2 Calculate the percentage of restaurants in each price range category.

3:1 CREATE A HISTOGRAM OR BAR CHART TO VISUALIZE THE DISTRIBUTION OF PRICE RANGES AMONG THE RESTAURANTS

```
#import libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px

#import data
dataset= pd.read_csv("dataset.csv")

#check data
dataset.head(10)
```

| Restaurant ID | Restaurant Name | Country |
|---------------|--|---------|
| 0 | Le Petit Souffle | |
| 1 | Izakaya Kikufuji | |
| 2 | Heat - Edsa Shangri-La | |
| 3 | Ooma | |
| 4 | Sambo Kojin | |
| 5 | Din Tai Fung | |
| 6 | Buffet 101 | |
| 7 | Vikings | |
| 8 | Spiral - Sofitel Philippine Plaza Manila | |
| 9 | Locavore | |

| City | Address |
|------|---|
| 0 | Makati City Third Floor, Century City Mall, Kalayaan Avenu... |

| | | |
|---|------------------|---|
| 1 | Makati City | Little Tokyo, 2277 Chino Roces Avenue, Legaspi... |
| 2 | Mandaluyong City | Edsa Shangri-La, 1 Garden Way, Ortigas, Mandal... |
| 3 | Mandaluyong City | Third Floor, Mega Fashion Hall, SM Megamall, O... |
| 4 | Mandaluyong City | Third Floor, Mega Atrium, SM Megamall, Ortigas... |
| 5 | Mandaluyong City | Ground Floor, Mega Fashion Hall, SM Megamall, ... |
| 6 | Pasay City | Building K, SM By The Bay, Sunset Boulevard, M... |
| 7 | Pasay City | Building B, By The Bay, Seaside Boulevard, Mal... |
| 8 | Pasay City | Plaza Level, Sofitel Philippine Plaza Manila, ... |
| 9 | Pasig City | Brixton Technology Center, 10 Brixton Street, ... |

| | Locality \ |
|---|---|
| 0 | Century City Mall, Poblacion, Makati City |
| 1 | Little Tokyo, Legaspi Village, Makati City |
| 2 | Edsa Shangri-La, Ortigas, Mandaluyong City |
| 3 | SM Megamall, Ortigas, Mandaluyong City |
| 4 | SM Megamall, Ortigas, Mandaluyong City |
| 5 | SM Megamall, Ortigas, Mandaluyong City |
| 6 | SM by the Bay, Mall of Asia Complex, Pasay City |
| 7 | SM by the Bay, Mall of Asia Complex, Pasay City |
| 8 | Sofitel Philippine Plaza Manila, Pasay City |
| 9 | Kapitolyo |

| | Locality Verbose | Longitude |
|------------|---|------------|
| Latitude \ | | |
| 0 | Century City Mall, Poblacion, Makati City, Mak... | 121.027535 |
| | 14.565443 | |
| 1 | Little Tokyo, Legaspi Village, Makati City, Ma... | 121.014101 |
| | 14.553708 | |
| 2 | Edsa Shangri-La, Ortigas, Mandaluyong City, Ma... | 121.056831 |
| | 14.581404 | |
| 3 | SM Megamall, Ortigas, Mandaluyong City, Mandal... | 121.056475 |
| | 14.585318 | |
| 4 | SM Megamall, Ortigas, Mandaluyong City, Mandal... | 121.057508 |
| | 14.584450 | |
| 5 | SM Megamall, Ortigas, Mandaluyong City, Mandal... | 121.056314 |
| | 14.583764 | |
| 6 | SM by the Bay, Mall of Asia Complex, Pasay Cit... | 120.979667 |
| | 14.531333 | |
| 7 | SM by the Bay, Mall of Asia Complex, Pasay Cit... | 120.979333 |
| | 14.540000 | |
| 8 | Sofitel Philippine Plaza Manila, Pasay City, P... | 120.980090 |

14.552990

9

Kapitolyo, Pasig City 121.056532

14.572041

| | Cuisines | ... | Currency | \ |
|---|------------------------------------|-----|------------------|---|
| 0 | French, Japanese, Desserts | ... | Botswana Pula(P) | |
| 1 | Japanese | ... | Botswana Pula(P) | |
| 2 | Seafood, Asian, Filipino, Indian | ... | Botswana Pula(P) | |
| 3 | Japanese, Sushi | ... | Botswana Pula(P) | |
| 4 | Japanese, Korean | ... | Botswana Pula(P) | |
| 5 | Chinese | ... | Botswana Pula(P) | |
| 6 | Asian, European | ... | Botswana Pula(P) | |
| 7 | Seafood, Filipino, Asian, European | ... | Botswana Pula(P) | |
| 8 | European, Asian, Indian | ... | Botswana Pula(P) | |
| 9 | Filipino | ... | Botswana Pula(P) | |

| | Has Table booking | Has Online delivery | Is delivering now | \ |
|---|-------------------|---------------------|-------------------|---|
| 0 | Yes | No | No | |
| 1 | Yes | No | No | |
| 2 | Yes | No | No | |
| 3 | No | No | No | |
| 4 | Yes | No | No | |
| 5 | No | No | No | |
| 6 | Yes | No | No | |
| 7 | Yes | No | No | |
| 8 | Yes | No | No | |
| 9 | Yes | No | No | |

| | Switch to order menu | Price range | Aggregate rating | Rating color | \ |
|---|----------------------|-------------|------------------|--------------|---|
| 0 | No | 3 | 4.8 | Dark Green | |
| 1 | No | 3 | 4.5 | Dark Green | |
| 2 | No | 4 | 4.4 | Green | |
| 3 | No | 4 | 4.9 | Dark Green | |
| 4 | No | 4 | 4.8 | Dark Green | |
| 5 | No | 3 | 4.4 | Green | |
| 6 | No | 4 | 4.0 | Green | |
| 7 | No | 4 | 4.2 | Green | |
| 8 | No | 4 | 4.9 | Dark Green | |
| 9 | No | 3 | 4.8 | Dark Green | |

| | Rating text | Votes |
|---|-------------|-------|
| 0 | Excellent | 314 |
| 1 | Excellent | 591 |
| 2 | Very Good | 270 |
| 3 | Excellent | 365 |
| 4 | Excellent | 229 |
| 5 | Very Good | 336 |
| 6 | Very Good | 520 |
| 7 | Very Good | 677 |
| 8 | Excellent | 621 |

```
9    Excellent    532
```

```
[10 rows x 21 columns]
```

```
#check database shape
```

```
dataset.shape
```

```
(9551, 21)
```

```
#check dataset information
```

```
dataset.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 9551 entries, 0 to 9550
```

```
Data columns (total 21 columns):
```

| # | Column | Non-Null Count | Dtype |
|----|----------------------|----------------|---------|
| 0 | 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
```

```
#check dataset column names
```

```
dataset.columns
```

```
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 Preprocessing

#check for null values

```
pd.isnull(dataset).sum()
```

```

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

```

#drop all null values

```
dataset.dropna(inplace=True)
```

#check database

```
dataset.shape
```

```
(9542, 21)
```

```
dataset.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
Index: 9542 entries, 0 to 9550
```

```
Data columns (total 21 columns):
```

| # | Column | Non-Null Count | Dtype |
|---|-----------------|----------------|--------|
| 0 | Restaurant ID | 9542 non-null | int64 |
| 1 | Restaurant Name | 9542 non-null | object |

| | | | | |
|----|----------------------|------|----------|---------|
| 2 | Country Code | 9542 | non-null | int64 |
| 3 | City | 9542 | non-null | object |
| 4 | Address | 9542 | non-null | object |
| 5 | Locality | 9542 | non-null | object |
| 6 | Locality Verbose | 9542 | non-null | object |
| 7 | Longitude | 9542 | non-null | float64 |
| 8 | Latitude | 9542 | non-null | float64 |
| 9 | Cuisines | 9542 | non-null | object |
| 10 | Average Cost for two | 9542 | non-null | int64 |
| 11 | Currency | 9542 | non-null | object |
| 12 | Has Table booking | 9542 | non-null | object |
| 13 | Has Online delivery | 9542 | non-null | object |
| 14 | Is delivering now | 9542 | non-null | object |
| 15 | Switch to order menu | 9542 | non-null | object |
| 16 | Price range | 9542 | non-null | int64 |
| 17 | Aggregate rating | 9542 | non-null | float64 |
| 18 | Rating color | 9542 | non-null | object |
| 19 | Rating text | 9542 | non-null | object |
| 20 | Votes | 9542 | non-null | int64 |

dtypes: float64(3), int64(5), object(13)

memory usage: 1.6+ MB

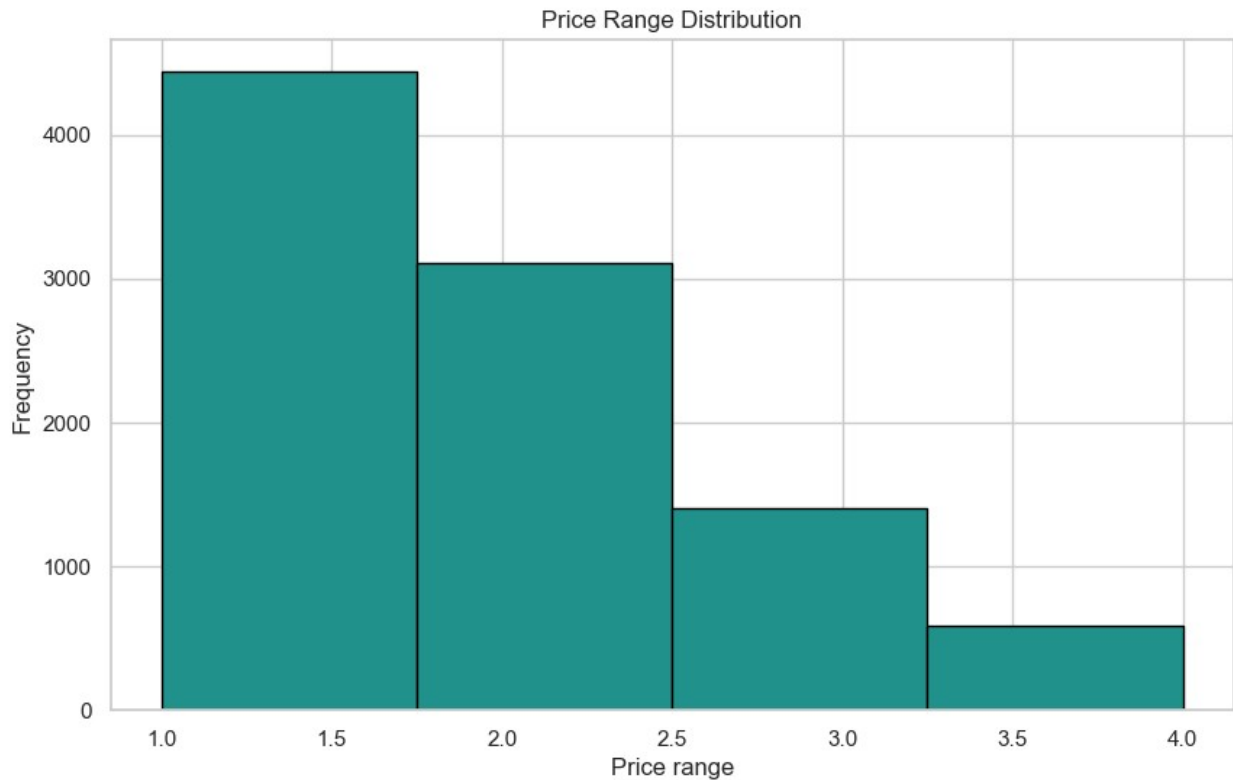
#check description of data

```
dataset[['Average Cost for two', 'Price range', 'Aggregate rating',
'Votes']].describe()
```

| | Average Cost for two | Price range | Aggregate rating |
|--------------|----------------------|-------------|------------------|
| Votes | | | |
| count | 9542.000000 | 9542.000000 | 9542.000000 |
| 9542.000000 | | | |
| mean | 1200.326137 | 1.804968 | 2.665238 |
| 156.772060 | | | |
| std | 16128.743876 | 0.905563 | 1.516588 |
| 430.203324 | | | |
| min | 0.000000 | 1.000000 | 0.000000 |
| 0.000000 | | | |
| 25% | 250.000000 | 1.000000 | 2.500000 |
| 5.000000 | | | |
| 50% | 400.000000 | 2.000000 | 3.200000 |
| 31.000000 | | | |
| 75% | 700.000000 | 2.000000 | 3.700000 |
| 130.000000 | | | |
| max | 800000.000000 | 4.000000 | 4.900000 |
| 10934.000000 | | | |

```
sns.set(style="whitegrid")
plt.subplots(figsize=(10,6))
plt.hist(dataset['Price range'],bins=dataset['Price
range'].nunique(),edgecolor='black',color='#21918c')
plt.title("Price Range Distribution")
```

```
plt.xlabel('Price range')
plt.ylabel('Frequency')
plt.show()
```



3:2 CALCULATE THE PERCENTAGE OF RESTAURANTS IN EACH PRICE RANGE CATEGORY.

```
price_range_counts= dataset['Price
range'].value_counts().reset_index()
price_range_counts
```

| | Price range | count |
|---|-------------|-------|
| 0 | 1 | 4438 |
| 1 | 2 | 3113 |
| 2 | 3 | 1405 |
| 3 | 4 | 586 |

```
#create a dataframe to hold the counts of restaurants in each price
range
```

```
price_range_counts= dataset['Price
range'].value_counts().reset_index()
```

```
#calculate the total number of restaurants
```

```
total_restaurants= len(dataset)
```

```
#calculate the percentage of restaurants in each price range category
#use the 'normalize' argument in value_counts to get percentages
```

```

directly
percentage_per_price_range= dataset['Price
range'].value_counts(normalize=True)* 100

#print result
print("Percentage of restaurants in each price range category:")
for price_range, percentage in percentage_per_price_range.items():
    print(f"{price_range}: {percentage:.2f}%")

Percentage of restaurants in each price range category:
1: 46.51%
2: 32.62%
3: 14.72%
4: 6.14%

percentage_df= percentage_per_price_range.reset_index()
percentage_df.columns= ['Price range', 'percentage']

#setting the style and color palette
sns.set(style="whitegrid")

#plotting the bar plot for percentage of restaurant in each price
range
plt.figure(figsize=(8,5))
sns.barplot(x='Price range',y='percentage', data=percentage_df,
color='#21918c')

#cutomizing the plot
plt.title("Percentage of Restaurants in Each Price Range", fontsize=
16)
plt.xlabel("Price range (1-4)", fontsize=12)
plt.ylabel("Percentage of restsaurants (%)", fontsize=12)
plt.show()

```


Percentage of Restaurants in Each Price Range

