

```
In [1]: import pandas as pd
```

```
In [2]: df=pd.read_csv("dataset restrarant new.csv" )
```

```
In [3]: df
```

Out[3]:

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Locality Verbose	Longitude	Latitude	Cuisines	Average Cost for two
0	51040	Truffles	1	Bangalore	28, 4th 'B' Cross, Koramangala 5th Block, Bang...	Koramangala 5th Block	Koramangala 5th Block, Bangalore	77.614293	12.933298	American, Burger, Cafe	800
1	51705	Toit	1	Bangalore	298, Namma Metro Pillar 62, 100 Feet Road, Ind...	Indiranagar	Indiranagar, Bangalore	77.640709	12.979166	Italian, American, Pizza	2000
2	308322	Hauz Khas Social	1	New Delhi	9A & 12, Hauz Khas Village, New Delhi	Hauz Khas Village	Hauz Khas Village, New Delhi	77.194471	28.554285	Continental, American, Asian, North Indian	1600
3	20404	Peter Cat	1	Kolkata	18A, Park Street, Park Street Area, Kolkata	Park Street Area	Park Street Area, Kolkata	88.352885	22.552672	Continental, North Indian	1000
4	56618	AB's Absolute Barbecues	1	Bangalore	90/4, 3rd Floor, Outer Ring Road, Munnekollaly...	Marathahalli	Marathahalli, Bangalore	77.699386	12.949934	European, Mediterranean, North Indian	1400
...
9543	18486858	6 Packs Momos	1	Noida	Spice World Mall, Sector 25, Noida	Spice World Mall, Sector 25	Spice World Mall, Sector 25, Noida	77.340602	28.586000	Chinese	300
9544	18431152	Cafe' Wow	1	Noida	Food Court, 3rd Floor, The Great India Palace ...	The Great India Place, Sector 38	The Great India Place, Sector 38, Noida	77.325600	28.567514	Fast Food	200
9545	18439721	Chef's Basket Pop Up Cafe	1	Noida	Inside Big Bazaar, The Great India Place, Sect...	The Great India Place, Sector 38	The Great India Place, Sector 38, Noida	0.000000	0.000000	Italian, Chinese	200

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Locality Verbose	Longitude	Latitude	Cuisines	Average Cost for two
9546	18428504	The HangoutDeli	1	Noida	320A, 3rd Floor, The Great India Place Mall, S...	The Great India Place, Sector 38	The Great India Place, Sector 38, Noida	77.323213	28.567751	Continental, Lebanese, Mexican	100
9547	18254559	Platters	1	Noida	Shop 1, Tulip Mall, Near HDFC Bank, Sector 48,...	Tulip Mall, Sector 48, Noida	Tulip Mall, Sector 48, Noida, Noida	77.367322	28.557930	North Indian, Chinese	50

9548 rows × 20 columns

level 1

TASK 1: TOP CUISINES

```
In [4]: top_cuisines = df['Cuisines'].value_counts().head(3)
print(top_cuisines)
```

North Indian 936
North Indian, Chinese 511
Chinese 354
Name: Cuisines, dtype: int64

```
In [5]: Cuisines_counts = df['Cuisines'].value_counts()
Cuisines_percentages = (Cuisines_counts / len(df)) * 100
top_Cuisines = Cuisines_percentages.head(3)
print(top_Cuisines)
```

North Indian 9.803100
North Indian, Chinese 5.351906
Chinese 3.707583
Name: Cuisines, dtype: float64

TASK 2: City Analysis

```
In [6]: top_City = df['City'].value_counts().idxmax()

print("City with the highest number of restaurants:", top_City)
```

City with the highest number of restaurants: New Delhi

```
In [7]: df["Aggregate rating"] = pd.to_numeric(df["Aggregate rating"], errors="coerce")
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
estambul       4.292857
Name: Aggregate rating, Length: 141, dtype: float64
```

```
In [8]: df["Aggregate rating"] = pd.to_numeric(df["Aggregate rating"], errors="coerce")

average_ratings = df.groupby("City")["Aggregate rating"].mean()

top_city = average_ratings.idxmax()

print("City with the highest average rating:", top_city)
```

City with the highest average rating: Inner City

Task 3: Price Range Distribution

```
In [9]: import matplotlib.pyplot as plt
```

```
In [10]: price_range_counts = df["Price range"].value_counts().sort_index()

# Plot the bar chart
plt.figure(figsize=(10, 5))
plt.bar(price_range_counts.index, price_range_counts.values, color="purple")
plt.xlabel("Price Range")
plt.ylabel("Number of Restaurants")
plt.title("Distribution of Price Ranges Among Restaurants")
plt.xticks(price_range_counts.index)
plt.grid(axis="y", linestyle="--", alpha=0.9)

# Show the plot
plt.show()
```



```
In [11]: # Count occurrences of each price range and calculate percentages
price_range_percentages = (df["Price range"].value_counts(normalize=True) * 100).sort_index()

# Display the result
print(price_range_percentages)

1    46.533305
2    32.593213
3    14.736070
4     6.137411
Name: Price range, dtype: float64
```

Task 4: ONLINE DELIVERY

```
In [12]: online_delivery_count = df["Has Online delivery"].value_counts(normalize=True) * 100

print(online_delivery_count)

False    74.329703
True     25.670297
Name: Has Online delivery, dtype: float64
```

```
In [13]: df["Aggregate rating"] = pd.to_numeric(df["Aggregate rating"], errors="coerce")

average_ratings = df.groupby("Has Online delivery")["Aggregate rating"].mean()
print(average_ratings)

Has Online delivery
False    3.467453
True     3.381274
Name: Aggregate rating, dtype: float64
```

LEVEL 2

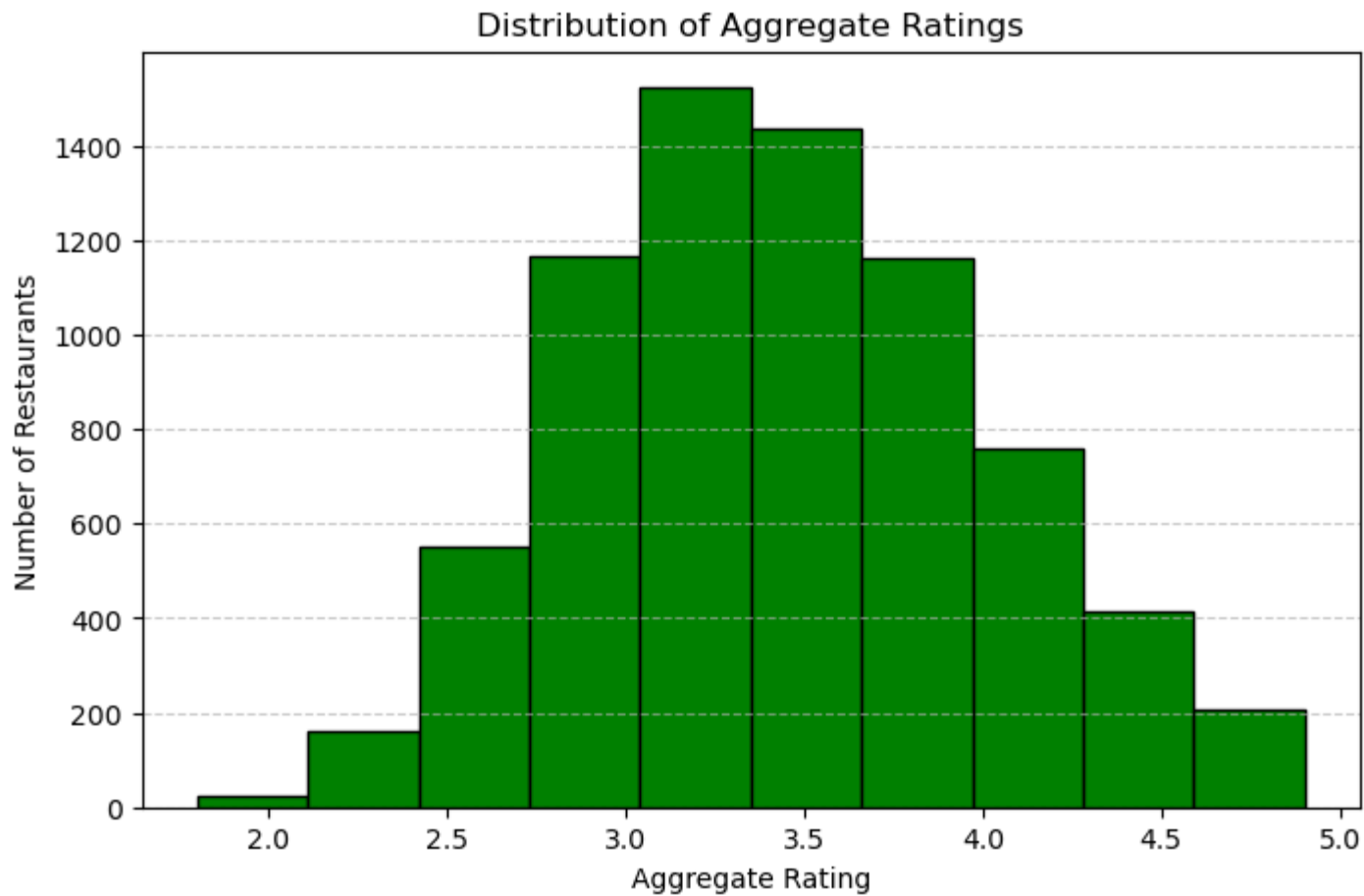
Task 1: Restrarant rating

```
In [14]: df["Aggregate rating"] = pd.to_numeric(df["Aggregate rating"], errors="coerce")

plt.figure(figsize=(8, 5))
plt.hist(df["Aggregate rating"].dropna(), bins=10, color="Green", edgecolor="black")
plt.xlabel("Aggregate Rating")
plt.ylabel("Number of Restaurants")
plt.title("Distribution of Aggregate Ratings")
plt.grid(axis="y", linestyle="--", alpha=0.7)

# Show the plot
plt.show()

most_common_rating_range = df["Aggregate rating"].round(1).value_counts().idxmax()
print("Most common rating:", most_common_rating_range)
```



Most common rating: 3.2

```
In [15]: df["Votes"] = pd.to_numeric(df["Votes"], errors="coerce")

average_votes = df["Votes"].mean()

average_votes = round(df["Votes"].mean(), 2)

print("Average number of votes received by restaurants:", average_votes)
```

Average number of votes received by restaurants: 156.78

Task 2: Cuisines Combination


```
In [16]: cuisine_combinations = df["Cuisines"].value_counts()
print("Most common cuisine combinations:")
print(cuisine_combinations.head(5)) # Show top 10
```

```
Most common cuisine combinations:
North Indian          936
North Indian, Chinese  511
Chinese               354
Fast Food             353
North Indian, Mughlai  333
Name: Cuisines, dtype: int64
```

```
In [17]: df["Aggregate rating"] = pd.to_numeric(df["Aggregate rating"], errors="coerce")

cuisine_ratings = df.groupby("Cuisines")["Aggregate rating"].mean()

top_cuisine_combinations = cuisine_ratings.sort_values(ascending=False)
print("Cuisine combinations with highest ratings:")
print(top_cuisine_combinations.head(5))
```

```
Cuisine combinations with highest ratings:
Cuisines
Burger, Bar Food, Steak          4.9
American, BBQ, Sandwich          4.9
Sunda, Indonesian               4.9
European, Asian, Indian          4.9
Italian, Bakery, Continental     4.9
Name: Aggregate rating, dtype: float64
```

Task 3: Geographic Analysis

```
In [18]: df["Longitude"] = pd.to_numeric(df["Longitude"], errors="coerce")
df["Latitude"] = pd.to_numeric(df["Latitude"], errors="coerce")

# Plot the locations
plt.figure(figsize=(10, 6))
plt.scatter(df["Longitude"], df["Latitude"], alpha=0.5, c="skyblue", edgecolors="black")
plt.xlabel("Longitude")
plt.ylabel("Latitude")
plt.title("Restaurant Locations on Map")
plt.grid(True)

# Show the plot
plt.show()
```



```
In [19]: from sklearn.cluster import KMeans
```

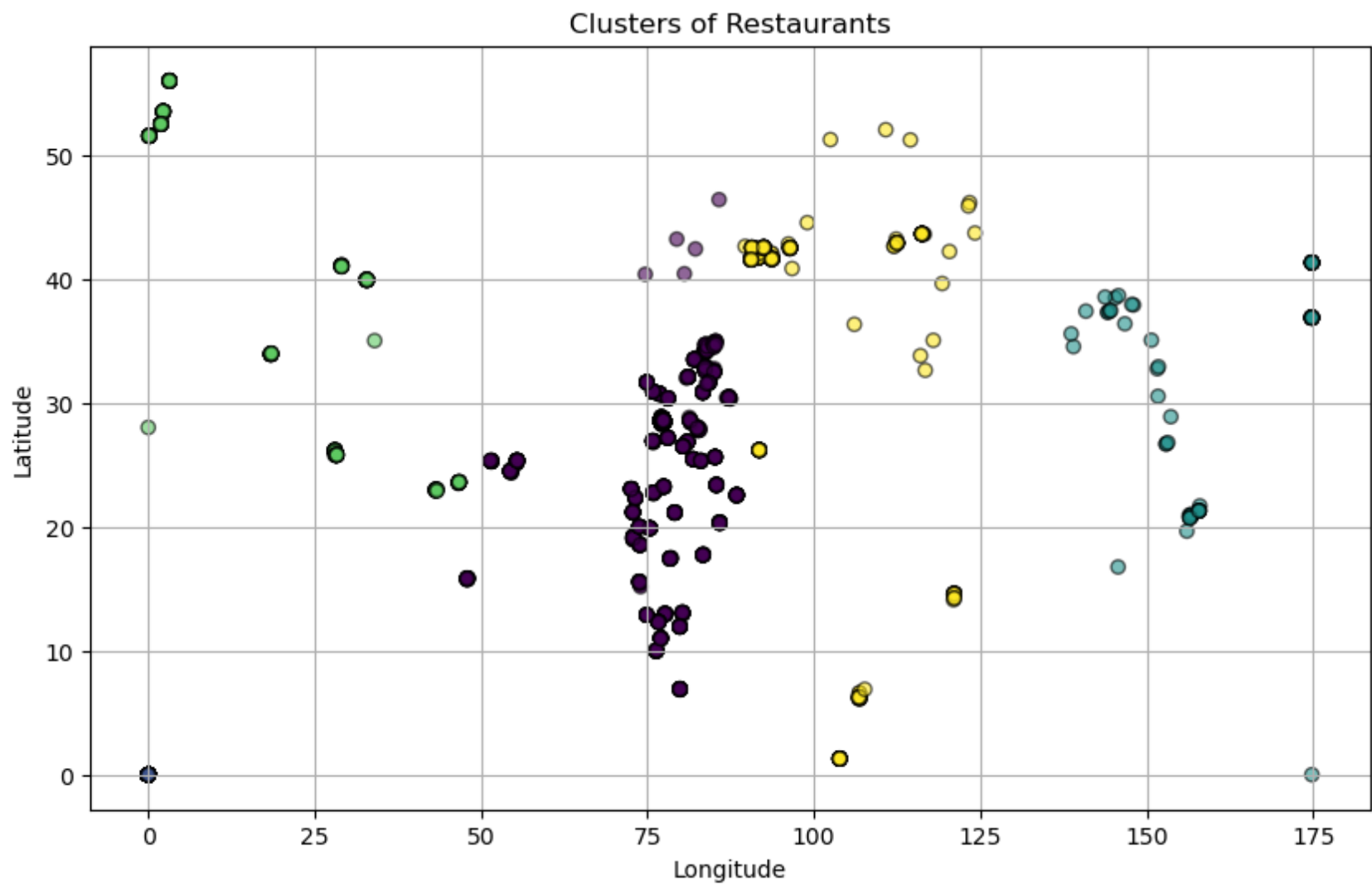
```
In [20]: df["Longitude"] = pd.to_numeric(df["Longitude"], errors="coerce")
df["Latitude"] = pd.to_numeric(df["Latitude"], errors="coerce")

df = df.dropna(subset=["Longitude", "Latitude"])

kmeans = KMeans(n_clusters=5, random_state=42, n_init=10)
df["Cluster"] = kmeans.fit_predict(df[["Longitude", "Latitude"]])

#Plot the clusters
plt.figure(figsize=(10, 6))
plt.scatter(df["Longitude"], df["Latitude"], c=df["Cluster"], cmap="viridis", alpha=0.6, edgecolors="black")
plt.xlabel("Longitude")
plt.ylabel("Latitude")
plt.title("Clusters of Restaurants")
plt.grid(True)

# Show the plot
plt.show()
```



Task 4:Restrarant Chains

```
In [21]: restaurant_counts = df["Restaurant Name"].value_counts()

restaurant_chains = restaurant_counts[restaurant_counts > 1]

#Display the result
print("Restaurant chains found in the dataset:")
print(restaurant_chains)
```

Restaurant chains found in the dataset:

Cafe Coffee Day	83
Domino's Pizza	79
Subway	63
Green Chick Chop	51
McDonald's	48

..

Chaap Point	2
Punjabi Chicken	2
Shawarma King's	2
Italiano	2
Wood Box Cafe	2

Name: Restaurant Name, Length: 736, dtype: int64

```
In [22]: df["Aggregate rating"] = pd.to_numeric(df["Aggregate rating"], errors="coerce")
df["Votes"] = pd.to_numeric(df["Votes"], errors="coerce")

restaurant_counts = df["Restaurant Name"].value_counts()
restaurant_chains = restaurant_counts[restaurant_counts > 1].index

chain_data = df[df["Restaurant Name"].isin(restaurant_chains)]
chain_analysis = chain_data.groupby("Restaurant Name").agg(
    Average_Rating=("Aggregate rating", "mean"),
    Total_Votes=("Votes", "sum")
).sort_values(by="Average_Rating", ascending=False)

# Display the result
print("Ratings and Popularity of Restaurant Chains:")
print(chain_analysis.head(10)) # Show top 10 chains
```

Ratings and Popularity of Restaurant Chains:

Restaurant Name	Average_Rating	Total_Votes
Talaga Sampireun	4.900	5514
Silantro FilMex	4.850	1364
AB's Absolute Barbecues	4.850	3151
AB's Absolute Barbecues	4.825	13400
Naturals Ice Cream	4.800	3094
Gymkhana	4.700	328
The Cheesecake Factory	4.650	3010
Garota de Ipanema	4.600	59
Dishoom	4.600	1269
Chili's	4.580	8156

In []: