

ZOMATO DATA ANALYSIS PROJECT USING PYTHON



Description:

This project focuses on analyzing customer behavior and restaurant performance on Zomato, a leading food delivery and review platform. Using Python and data analysis libraries like Pandas, NumPy, and Matplotlib, the project explores key business questions to uncover insights about customer preferences, restaurant popularity, and order patterns.

The analysis covers:

- Identifying the most preferred restaurant types by customers
- Aggregating customer votes across restaurant categories
- Understanding common rating trends among restaurants
- Estimating average spending per order by couples who order online
- Comparing ratings between online and offline ordering modes
- Detecting restaurant types with high offline order volumes to target promotional offers

This project demonstrates practical skills in data cleaning, grouping, aggregation, and visualization, and provides actionable insights that can help Zomato optimize its marketing and customer engagement strategies.

Data Analysis Project Using Python — Zomato Case Study Problems:

1. What type of restaurant do the majority of customers order from?
2. How many votes has each type of restaurant received from customers?
3. What are the ratings that the majority of restaurants have received?
4. Zomato has observed that most couples order most of their food online. What is their average spending on each order?
5. Which mode (online or offline) has received the maximum rating?
6. Which type of restaurant received more offline orders, so that Zomato can provide those customers with some good offers?

Zomato data analysis project

Importing libraries

```
In [2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

Read data

```
In [4]: df = pd.read_csv("Zomato data .csv")
print(df)
```

	name	online_order	book_table	rate	votes	\
0	Jalsa	Yes	Yes	4.1/5	775	
1	Spice Elephant	Yes	No	4.1/5	787	
2	San Churro Cafe	Yes	No	3.8/5	918	
3	Addhuri Udupi Bhojana	No	No	3.7/5	88	
4	Grand Village	No	No	3.8/5	166	
..	
143	Melting Melodies	No	No	3.3/5	0	
144	New Indraprasta	No	No	3.3/5	0	
145	Anna Kuteera	Yes	No	4.0/5	771	
146	Darbar	No	No	3.0/5	98	
147	Vijayalakshmi	Yes	No	3.9/5	47	

	approx_cost(for two people)	listed_in(type)
0	800	Buffet
1	800	Buffet
2	800	Buffet
3	300	Buffet
4	600	Buffet
..
143	100	Dining
144	150	Dining
145	450	Dining
146	800	Dining
147	200	Dining

[148 rows x 7 columns]

Show data

```
In [12]: df
```

Out[12]:

	name	online_order	book_table	rate	votes	approx_cost(for two people)	listed_in(type)
0	Jalsa	Yes	Yes	4.1	775	800	Buffet
1	Spice Elephant	Yes	No	4.1	787	800	Buffet
2	San Churro Cafe	Yes	No	3.8	918	800	Buffet
3	Addhuri Udupi Bhojana	No	No	3.7	88	300	Buffet
4	Grand Village	No	No	3.8	166	600	Buffet
...
143	Melting Melodies	No	No	3.3	0	100	Dining
144	New Indraprasta	No	No	3.3	0	150	Dining
145	Anna Kuteera	Yes	No	4.0	771	450	Dining
146	Darbar	No	No	3.0	98	800	Dining
147	Vijayalakshmi	Yes	No	3.9	47	200	Dining

148 rows × 7 columns

Convert the datatype of rate by removing denominator

```
In [10]: def handleRate(value):
          value = str(value).split('/')
          value = value[0];
          return float(value)

          df['rate']=df['rate'].apply(handleRate)
          print(df.head())
```

	name	online_order	book_table	rate	votes	\
0	Jalsa	Yes	Yes	4.1	775	
1	Spice Elephant	Yes	No	4.1	787	
2	San Churro Cafe	Yes	No	3.8	918	
3	Addhuri Udupi Bhojana	No	No	3.7	88	
4	Grand Village	No	No	3.8	166	

	approx_cost(for two people)	listed_in(type)
0	800	Buffet
1	800	Buffet
2	800	Buffet
3	300	Buffet
4	600	Buffet

Checking null and info

```
In [11]: df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 148 entries, 0 to 147
Data columns (total 7 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   name                                  148 non-null    object
1   online_order                         148 non-null    object
2   book_table                           148 non-null    object
3   rate                                 148 non-null    float64
4   votes                                148 non-null    int64
5   approx_cost(for two people)          148 non-null    int64
6   listed_in(type)                      148 non-null    object
dtypes: float64(1), int64(2), object(4)
memory usage: 8.2+ KB

```

Type of restaurant

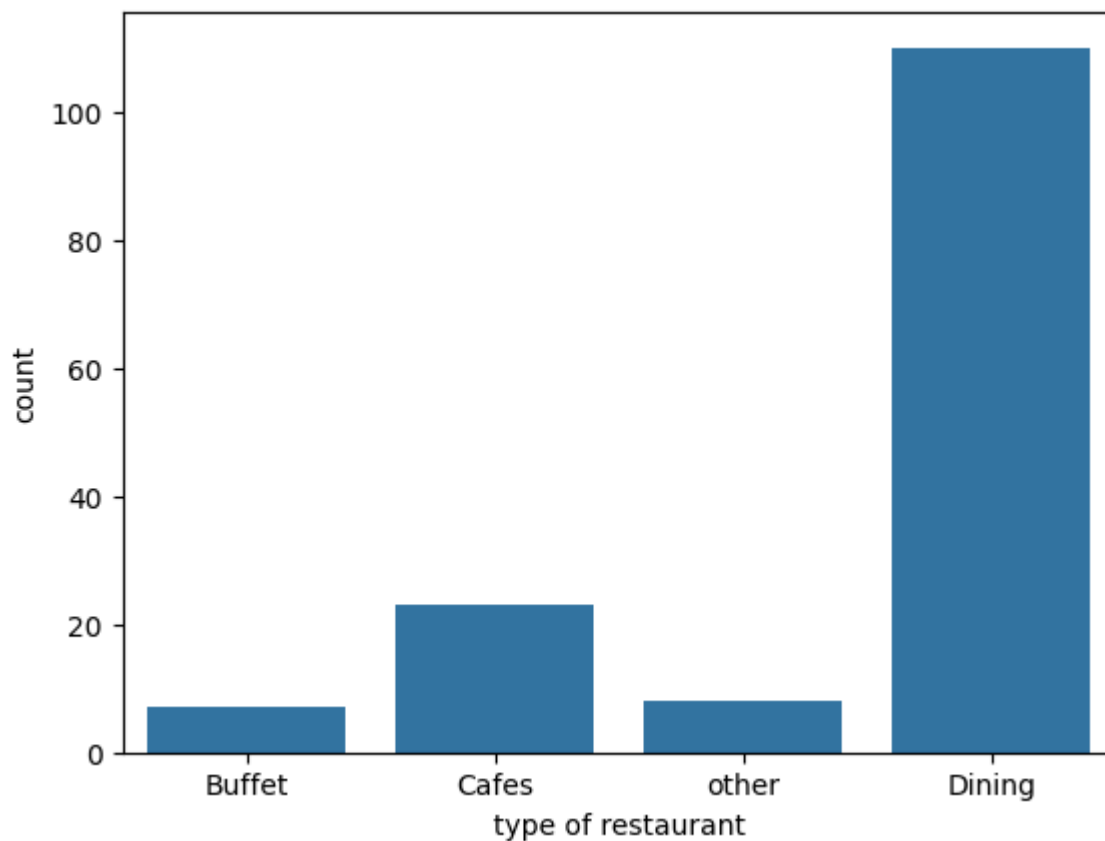
```
In [14]: df.head()
```

```
Out[14]:
```

	name	online_order	book_table	rate	votes	approx_cost(for two people)	listed_in(type)
0	Jalsa	Yes	Yes	4.1	775	800	Buffet
1	Spice Elephant	Yes	No	4.1	787	800	Buffet
2	San Churro Cafe	Yes	No	3.8	918	800	Buffet
3	Addhuri Udupi Bhojana	No	No	3.7	88	300	Buffet
4	Grand Village	No	No	3.8	166	600	Buffet

```
In [15]: sns.countplot(x=df['listed_in(type)'])
plt.xlabel("type of restaurant")
```

```
Out[15]: Text(0.5, 0, 'type of restaurant')
```

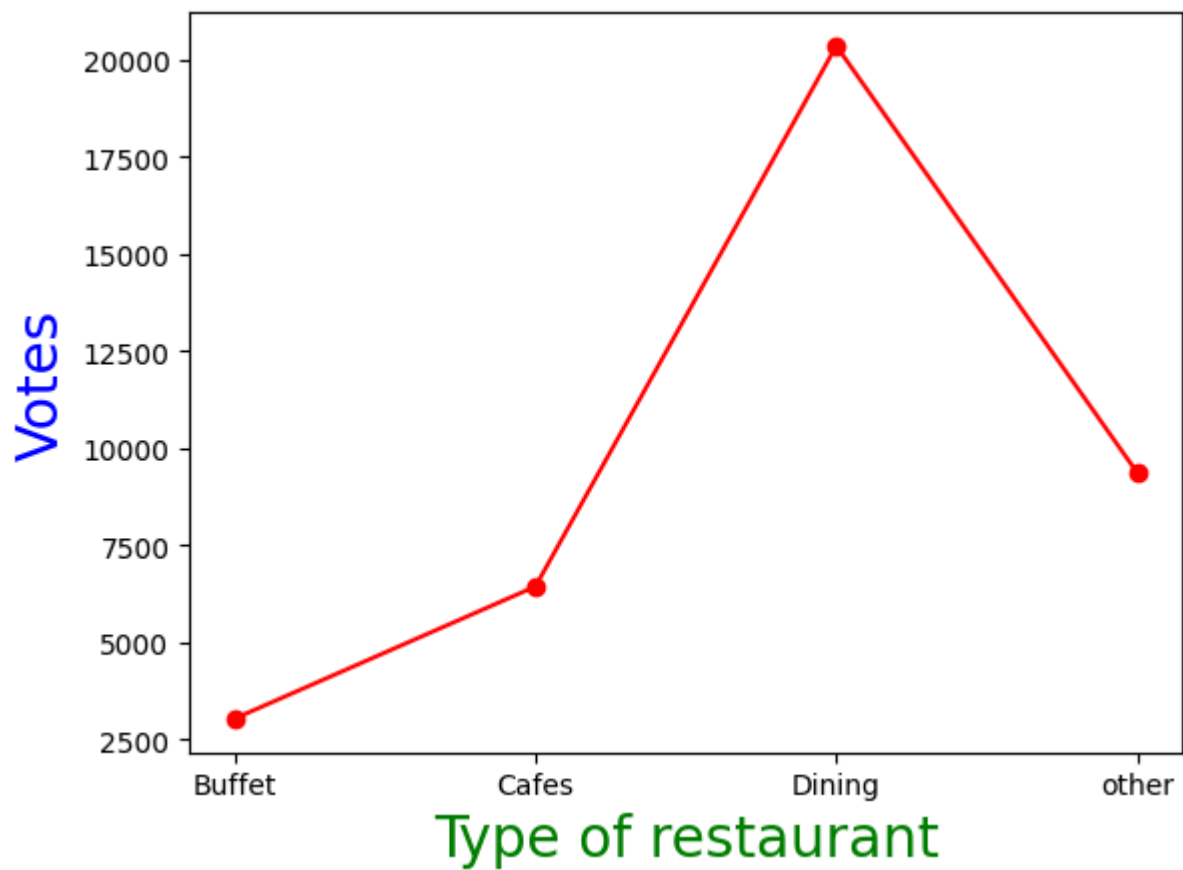


Conclusion: majority of the restaurant falls in dinning category

Group by restaurant type and sum of votes(line graph)

```
In [20]: grouped_data = df.groupby('listed_in(type)')['votes'].sum()
result = pd.DataFrame({'votes': grouped_data})
plt.plot(result, c="red", marker="o")
plt.xlabel("Type of restaurant", c="green", size=20)
plt.ylabel("Votes", c="blue", size=20)
```

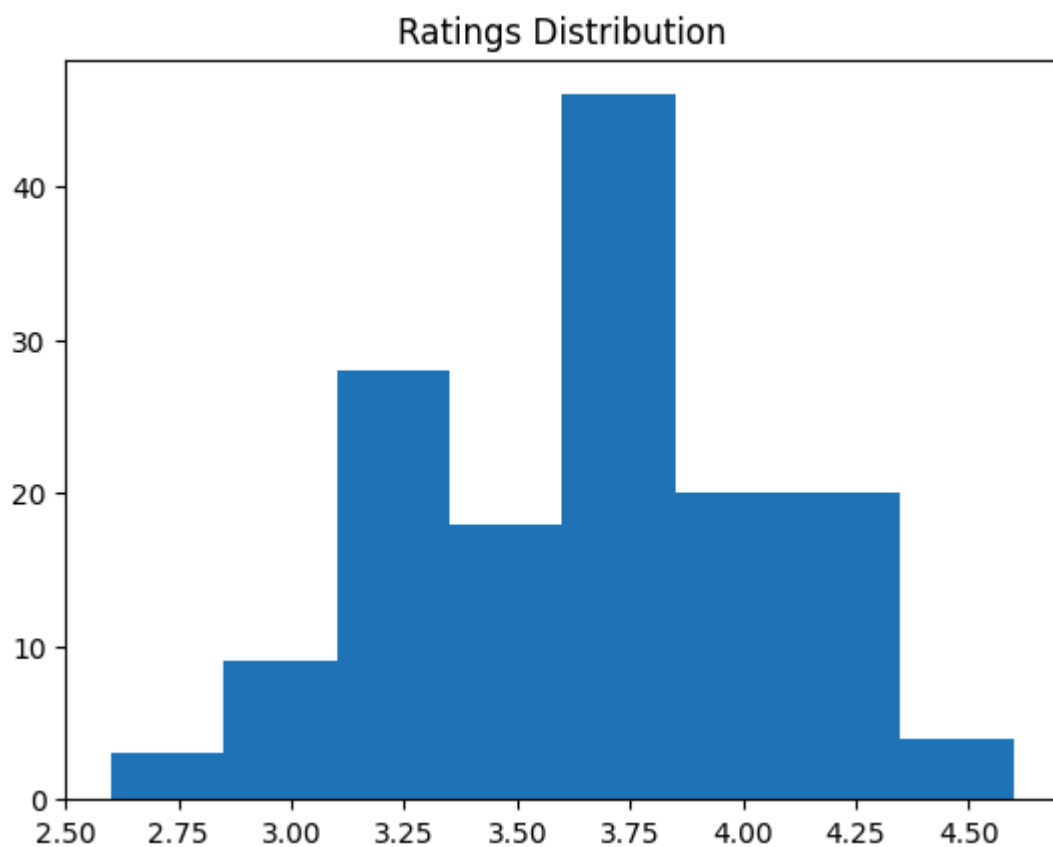
```
Out[20]: Text(0, 0.5, 'Votes')
```



Conclusion: dinning restaurants has received most votes

Histogram showing ratings distributions

```
In [29]: plt.hist(df['rate'], bins=8)
plt.title("Ratings Distribution")
plt.show()
```



Conclusion: majority rating is 3.75

Average spending in each order

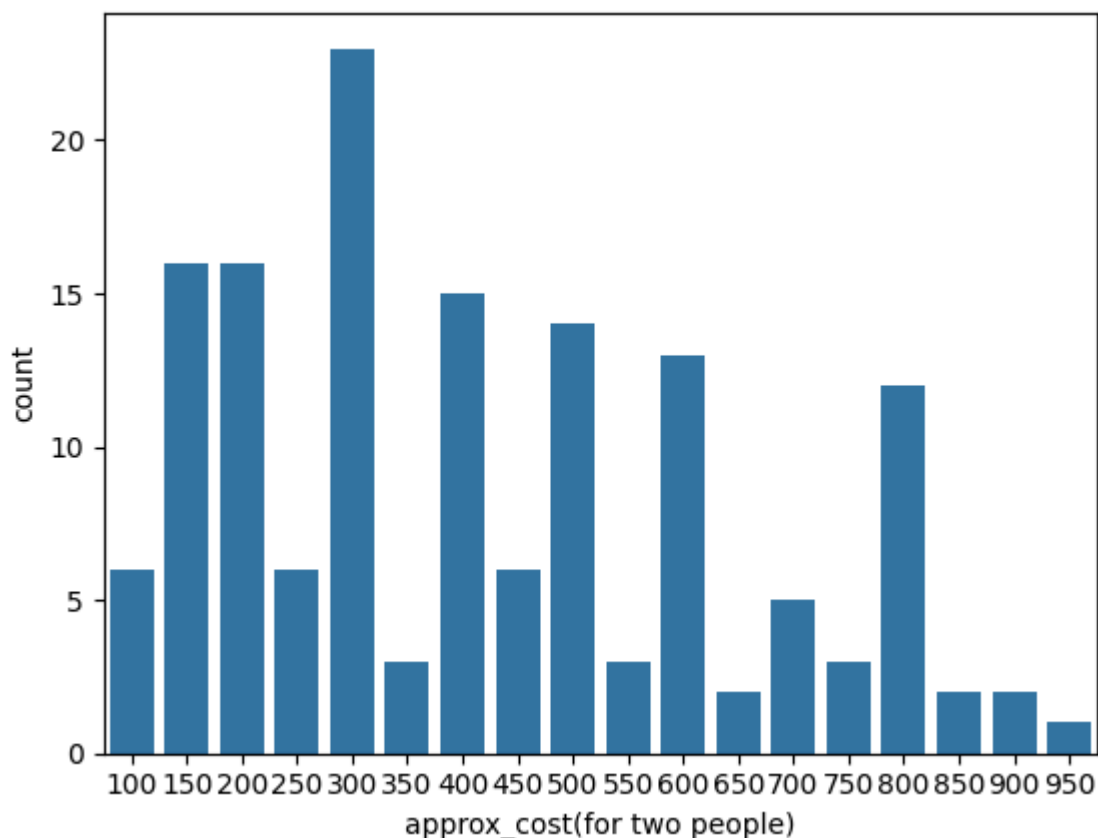
```
In [30]: df.head()
```

```
Out[30]:
```

	name	online_order	book_table	rate	votes	approx_cost(for two people)	listed_in(type)
0	Jalsa	Yes	Yes	4.1	775	800	Buffet
1	Spice Elephant	Yes	No	4.1	787	800	Buffet
2	San Churro Cafe	Yes	No	3.8	918	800	Buffet
3	Addhuri Udupi Bhojana	No	No	3.7	88	300	Buffet
4	Grand Village	No	No	3.8	166	600	Buffet

```
In [31]: couple_data = df['approx_cost(for two people)']  
sns.countplot(x=couple_data)
```

```
Out[31]: <Axes: xlabel='approx_cost(for two people)', ylabel='count'>
```



Conclusion: the majority of people prefers restaurants with an approx cost of 300 rupees.

Highest rating between online and offline mode

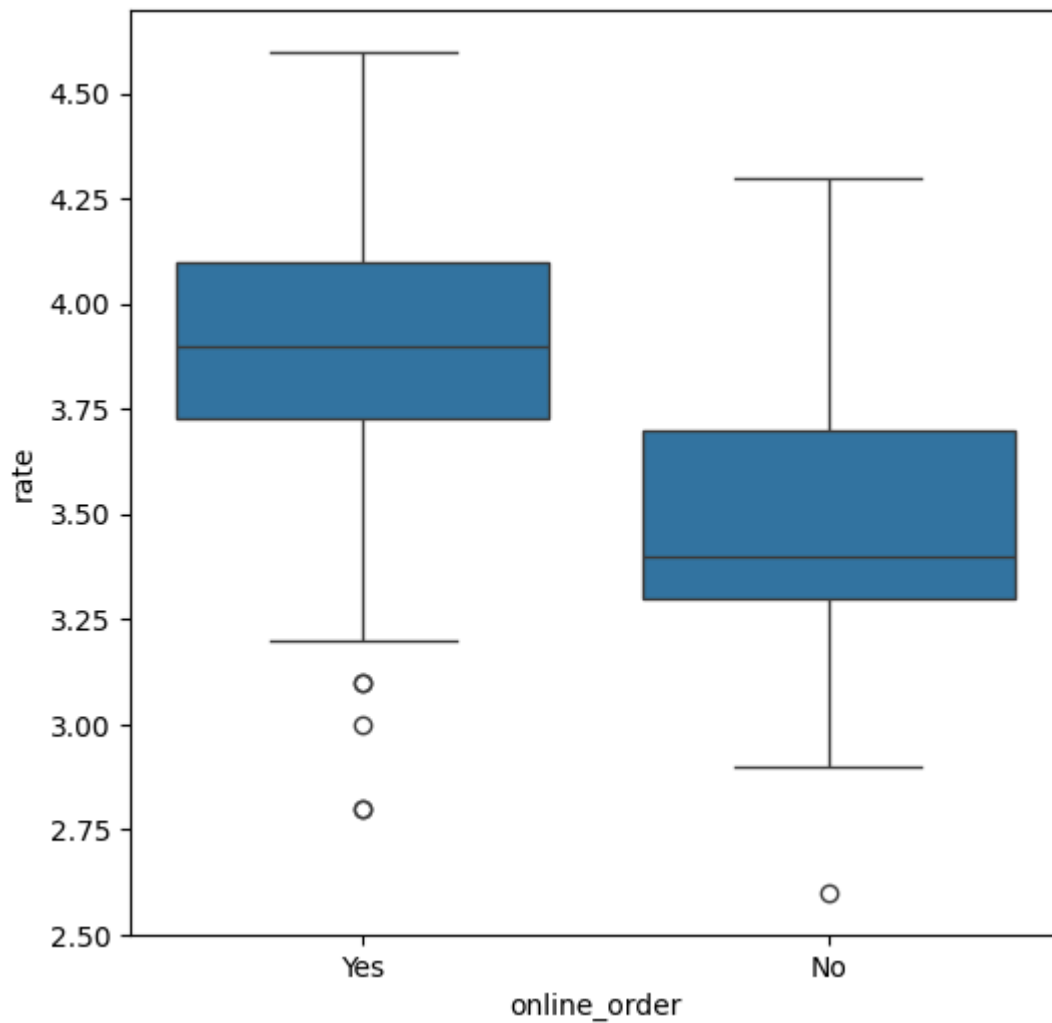
```
In [32]: df.head()
```

Out[32]:

	name	online_order	book_table	rate	votes	approx_cost(for two people)	listed_in(type)
0	Jalsa	Yes	Yes	4.1	775	800	Buffet
1	Spice Elephant	Yes	No	4.1	787	800	Buffet
2	San Churro Cafe	Yes	No	3.8	918	800	Buffet
3	Addhuri Udipi Bhojana	No	No	3.7	88	300	Buffet
4	Grand Village	No	No	3.8	166	600	Buffet

```
In [37]: plt.figure(figsize=(6,6))
sns.boxplot(x='online_order', y='rate', data=df)
```

Out[37]: <Axes: xlabel='online_order', ylabel='rate'>



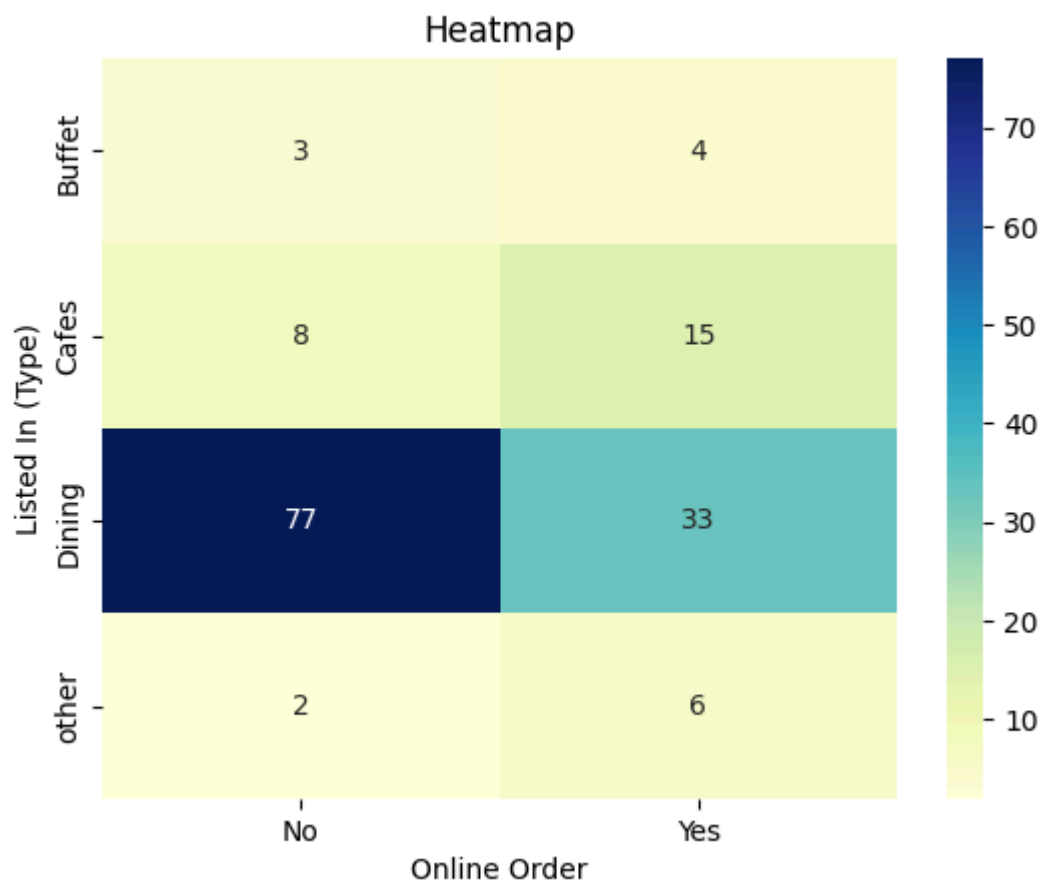
Conclusion: Offline orders received low rating in comparison to online.

Heatmap

```
In [38]: pivot_table = df.pivot_table(index="listed_in(type)", columns="online_order", aggfunc='size')

# Heatmap plot karte hain
sns.heatmap(pivot_table, annot=True, cmap="YlGnBu", fmt='d')
plt.title("Heatmap")
plt.xlabel("Online Order")
```

```
plt.ylabel("Listed In (Type)")
plt.show()
```



CONCLUSION:

Dining restaurants primarily accept offline orders, whereas cafes primarily receive online orders. This suggests that clients prefer to place orders in person at restaurants, but prefer online ordering at cafes.

In []: