

ZOMATO DATA ANALYSIS

IMPORTING LIBRARIES

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

Load Dataset

```
In [3]: df=pd.read_csv('Zomato data .csv')
```

```
In [4]: df
```

Out[4]:

	name	online_order	book_table	rate	votes	approx_cost(for two people)	listed_in(type)
0	Jalsa	Yes	Yes	4.1/5	775	800	Buffet
1	Spice Elephant	Yes	No	4.1/5	787	800	Buffet
2	San Churro Cafe	Yes	No	3.8/5	918	800	Buffet
3	Addhuri Udupi Bhojana	No	No	3.7/5	88	300	Buffet
4	Grand Village	No	No	3.8/5	166	600	Buffet
...
143	Melting Melodies	No	No	3.3/5	0	100	Dining
144	New Indraprasta	No	No	3.3/5	0	150	Dining
145	Anna Kuteera	Yes	No	4.0/5	771	450	Dining
146	Darbar	No	No	3.0/5	98	800	Dining
147	Vijayalakshmi	Yes	No	3.9/5	47	200	Dining

148 rows × 7 columns

Convert the data type of column-rate

```
In [5]: 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 Udipi 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

```
In [6]: 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
```

What type of Restaurant do majority customer order from ?

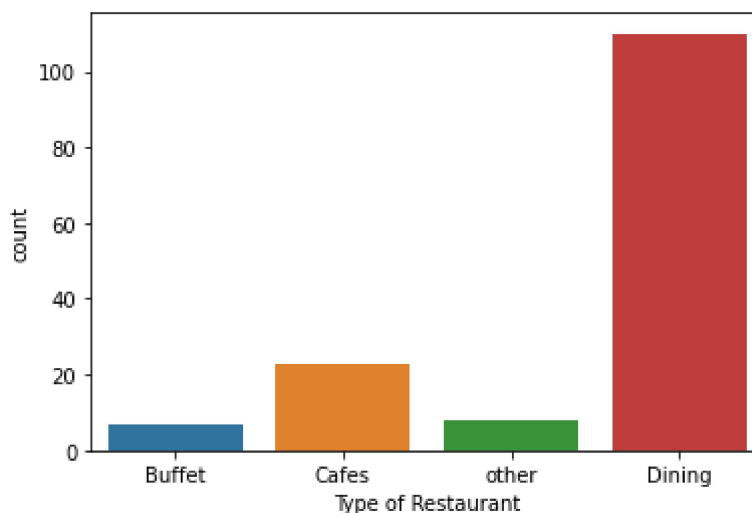
```
In [7]: df.head()
```

```
Out[7]:
```

	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 [8]: sns.countplot(x=df['listed_in(type)'])  
plt.xlabel("Type of Restaurant")
```

```
Out[8]: Text(0.5, 0, 'Type of Restaurant')
```



Conclusion : majority of the restaurant falls in Dinning Category so majority people order from dining restaurants

```
In [ ]:
```

How many votes has each type of restaurants recieved from customer

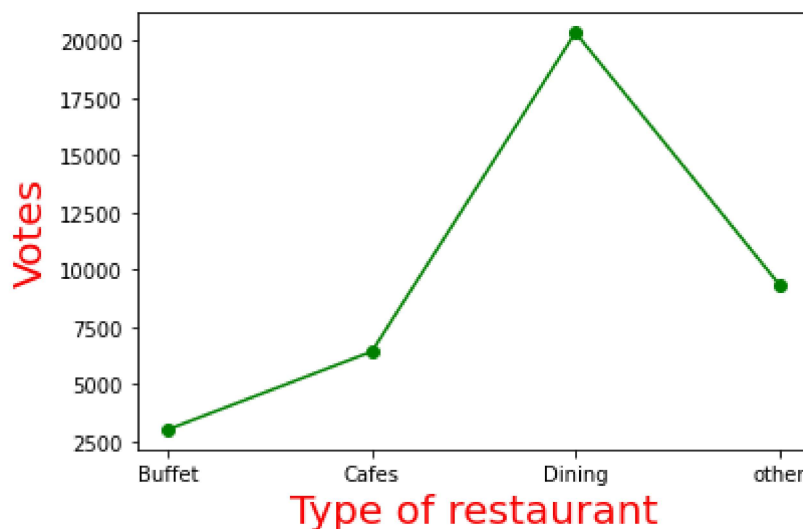
In [9]: `df.head()`

Out[9]:

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

Out[13]: Text(0, 0.5, 'Votes')



Conclusion: Dinning Restaurants has recieved maximum votes

In []:

What are the ratings that the majority of restaurants have recieved

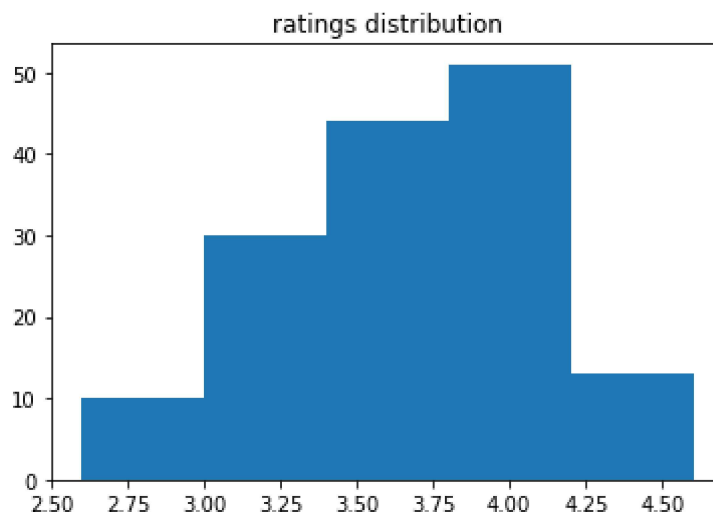
In [15]: `df.head()`

Out[15]:

	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 [16]: `plt.hist(df['rate'],bins=5)
plt.title("ratings distribution")
plt.show`

Out[16]: `<function matplotlib.pyplot.show(close=None, block=None)>`



Conclusion: The majority Restaurants recieved ratings from 3.5 to 4

In []:

In []:

Average order spending by Couples

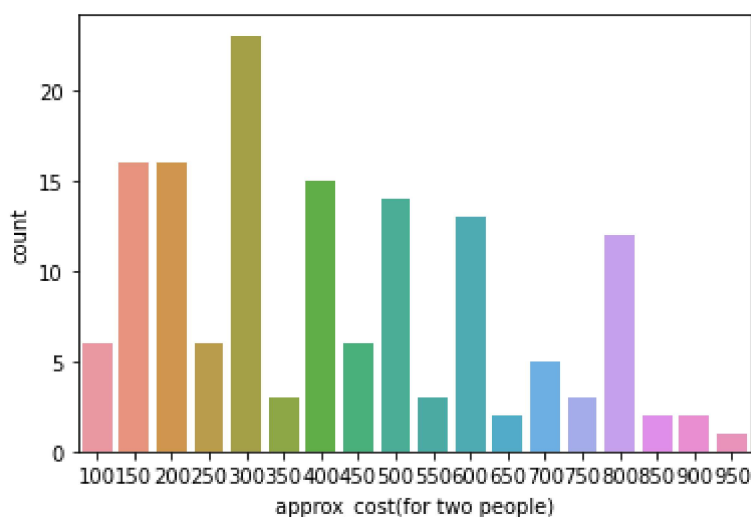
In [18]: `df.head()`

Out[18]:

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

Out[21]: <AxesSubplot:xlabel='approx_cost(for two people)', ylabel='count'>



Conclusion: The majority of couples prefer restaurants with an approximate cost of 300 rupees for 2 people

In []:

In []:

Which mode receives maximum rating

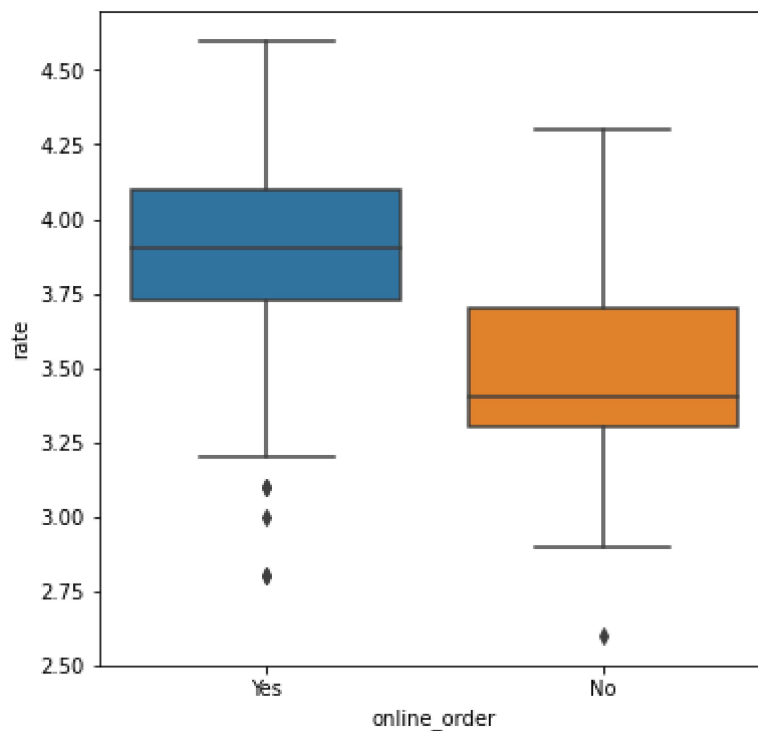
```
In [23]: df.head()
```

```
Out[23]:
```

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

```
Out[24]: <AxesSubplot:xlabel='online_order', ylabel='rate'>
```



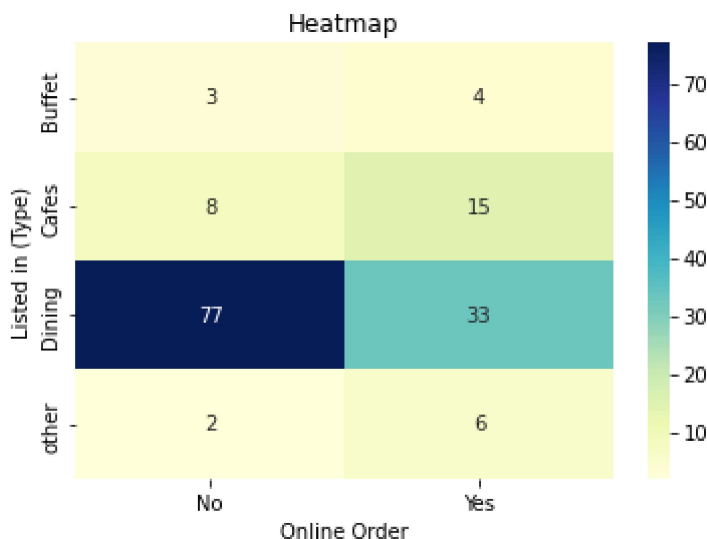
Conclusion: Offline order recieved lower rating in comparison to online order

In [26]: `df.head()`

Out[26]:

	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 [28]: `pivot_table=df.pivot_table(index='listed_in(type)', columns='online_order',
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 order. This suggests that clients prefer orders in person at restaurants, but prefer online order in Cafes.

In []: