

Zomato Data Analysis Project

Step-1 : Importing libraries

pandas for data manipulation and analysis

numpy for numerical operations

matplotlib and seaborn for data visualization

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

Step-2 : Creating the data frame

```
data = pd.read_csv("ZomatoData.csv")
```

```
data.head(5)
```

	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 Udipi Bhojana	No	No	3.7/5	88	
4	Grand Village	No	No	3.8/5	166	

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

Step-3 : Data Cleaning and Preprocessing

(a) convert data type of column - rate

```
def handleRate(value):
    value = str(value).split('/')
    value = value[0];
    return float(value)
data['rate'] = data['rate'].apply(handleRate)
data.head(5)
```

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(b) Check for missing values

```
data.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

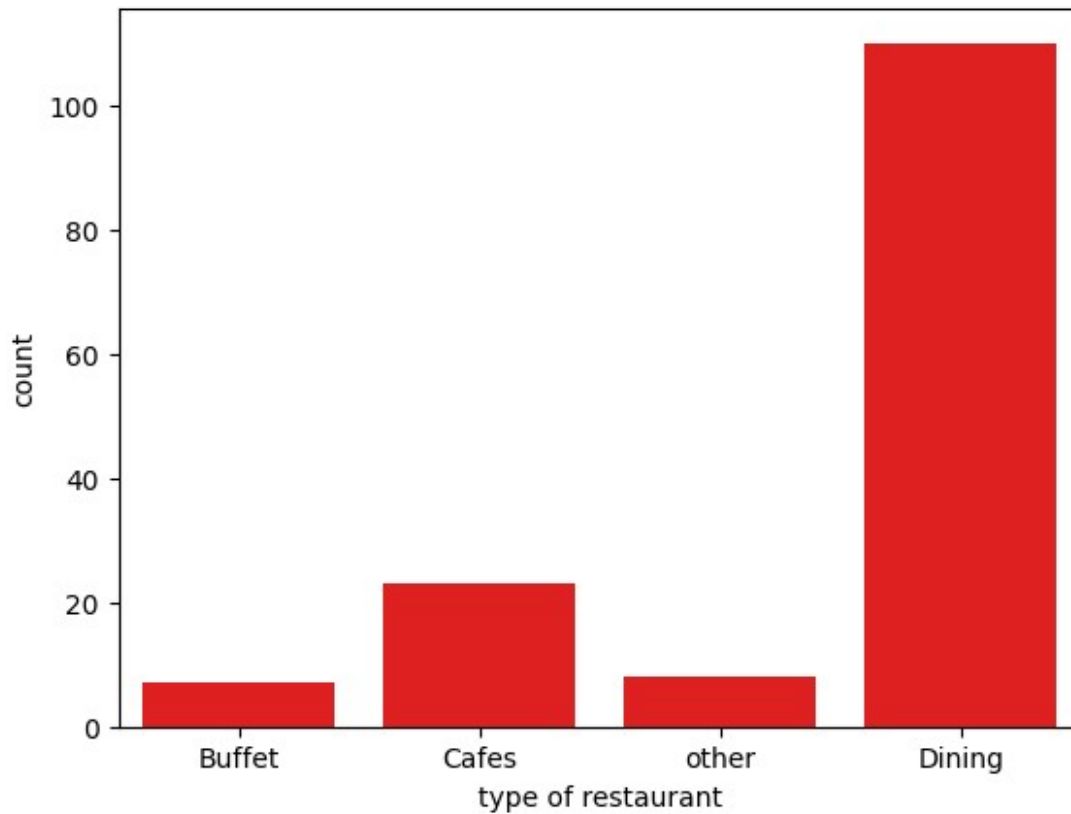
```
dtypes: float64(1), int64(2), object(4)
```

```
memory usage: 8.2+ KB
```

Step-4 : Exploratory Data Analysis, Visualisations and Insights

(1) Types of restaurant

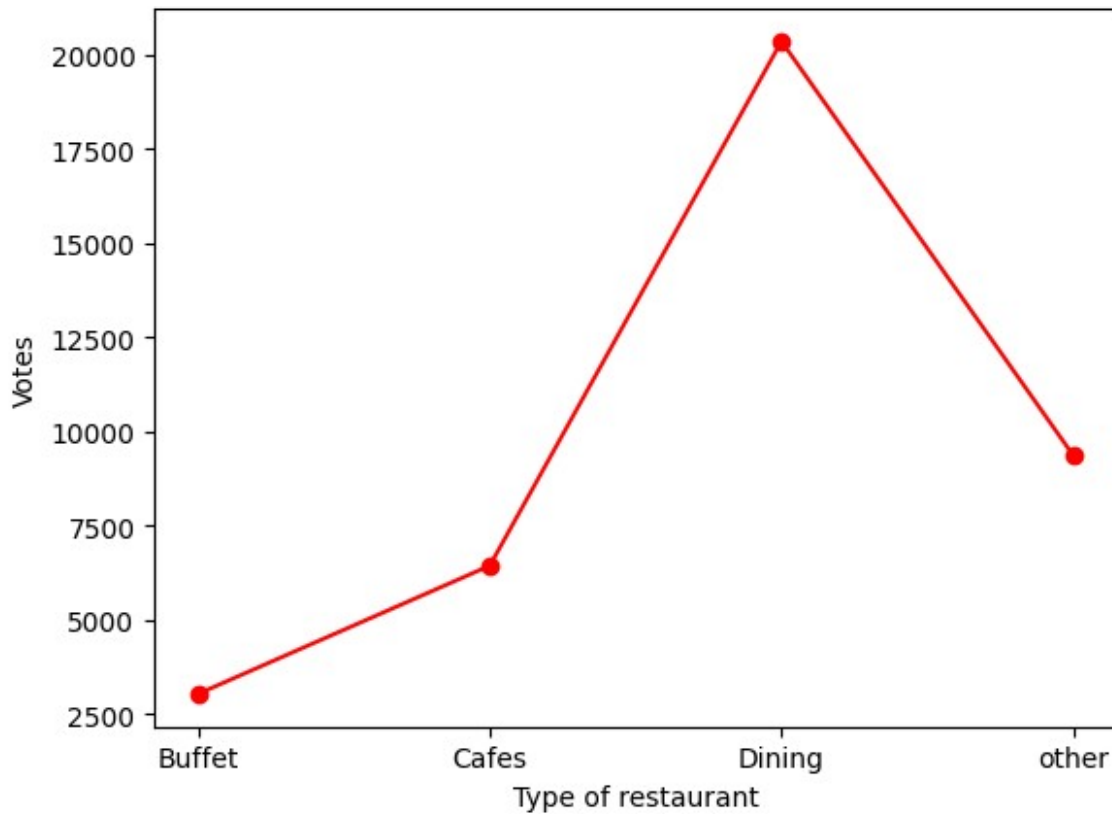
```
sns.countplot(x=data['listed_in(type)'], color="red")
plt.xlabel("type of restaurant")
plt.show()
```



conclusion - majority of the restaurants falls in dining category.

(2) Votings according to restaurant type

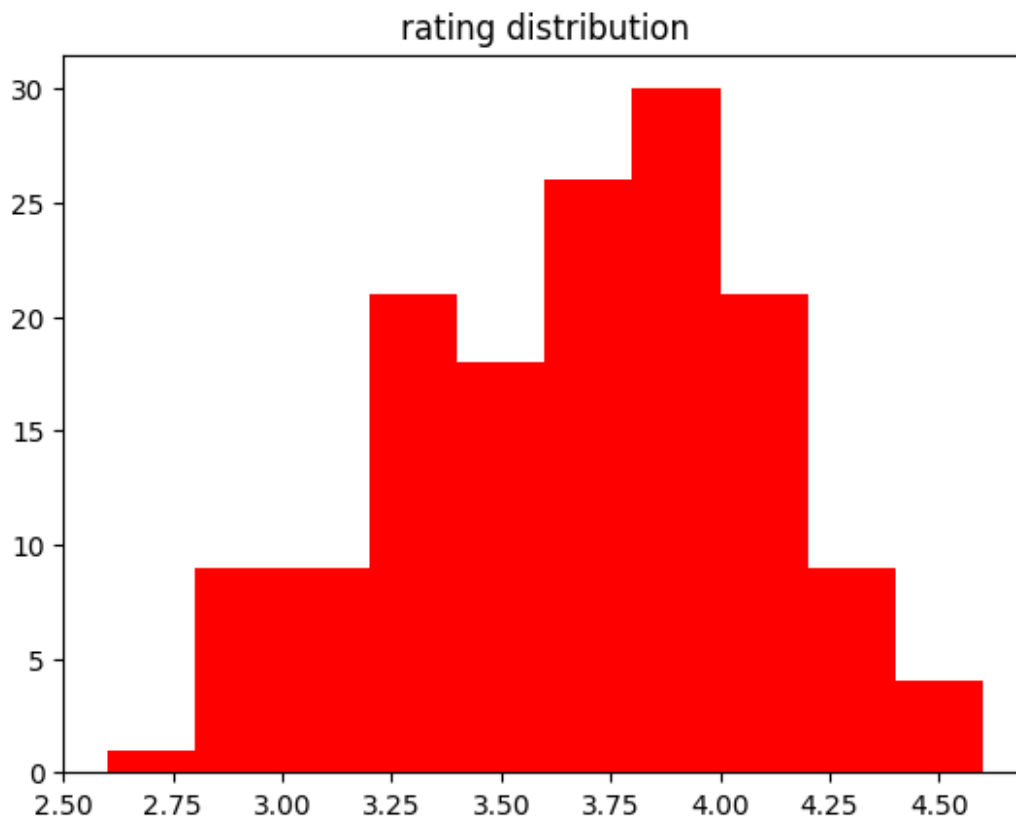
```
grouped_data = data.groupby("listed_in(type)")["votes"].sum()
result = pd.DataFrame({"votes": grouped_data})
plt.plot(result, c= "red", marker ="o")
plt.xlabel("Type of restaurant", size = 10)
plt.ylabel("Votes", size = 10)
Text(0, 0.5, 'Votes')
```



conclusion- Dining restaurants has recieved maximum votes

(3) Ratings of Majority of restaurants

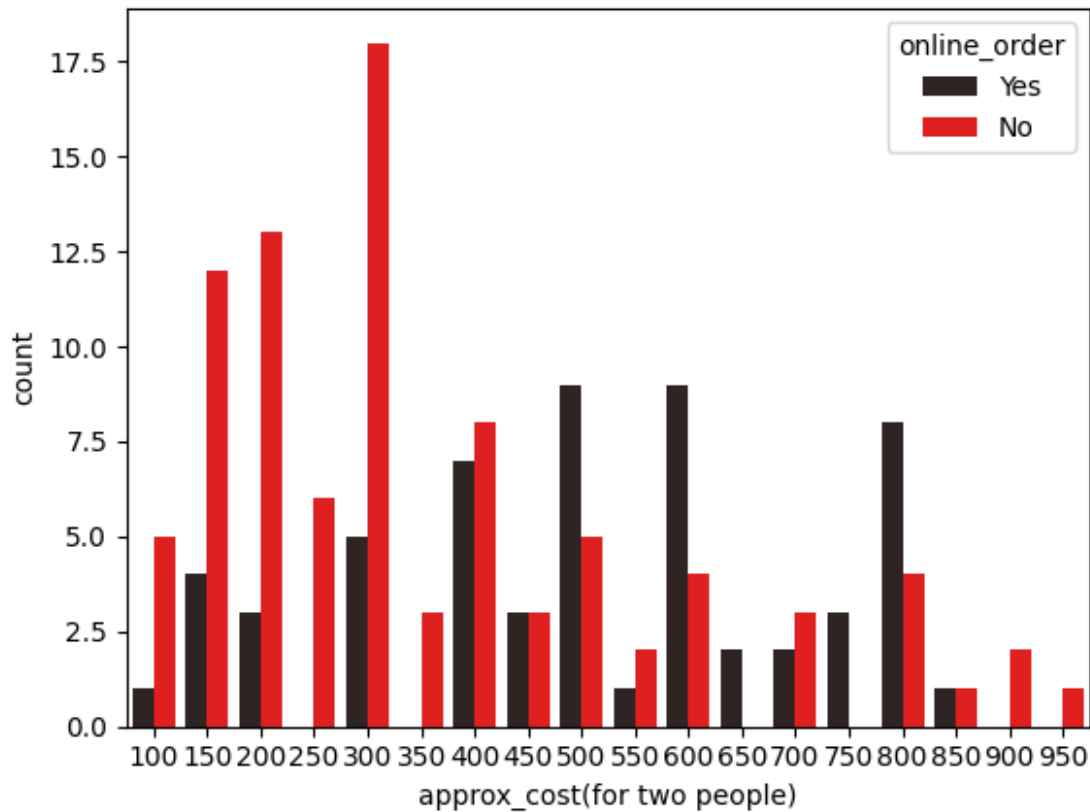
```
plt.hist(data["rate"],bins =10, color= 'red')  
plt.title("rating distribution")  
plt.show()
```



conclusion- the majority restaurant receives rating from 3.5 - 4.0

(4) Spendings on online and offline orders

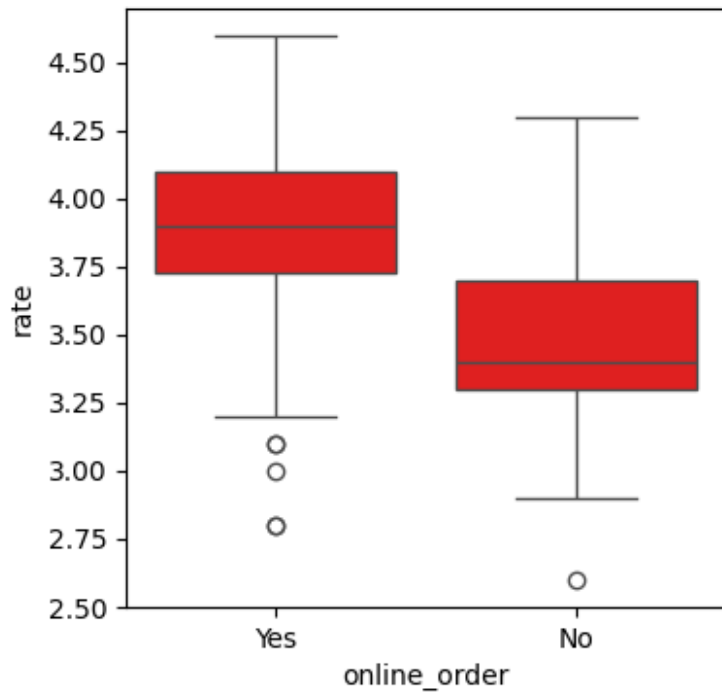
```
couple_data = data['approx_cost(for two people)']  
online = data['online_order']  
sns.countplot(x= couple_data, hue = online, palette = 'dark:red')  
<Axes: xlabel='approx_cost(for two people)', ylabel='count'>
```



conclusion - most of the offline orders (for 2 people) amounts to Rs.300 approx.
and most of the online orders (for 2 people) amounts between Rs.500 - Rs.600.

(5) Which mode (Online or offline) has recieved maximum ratings

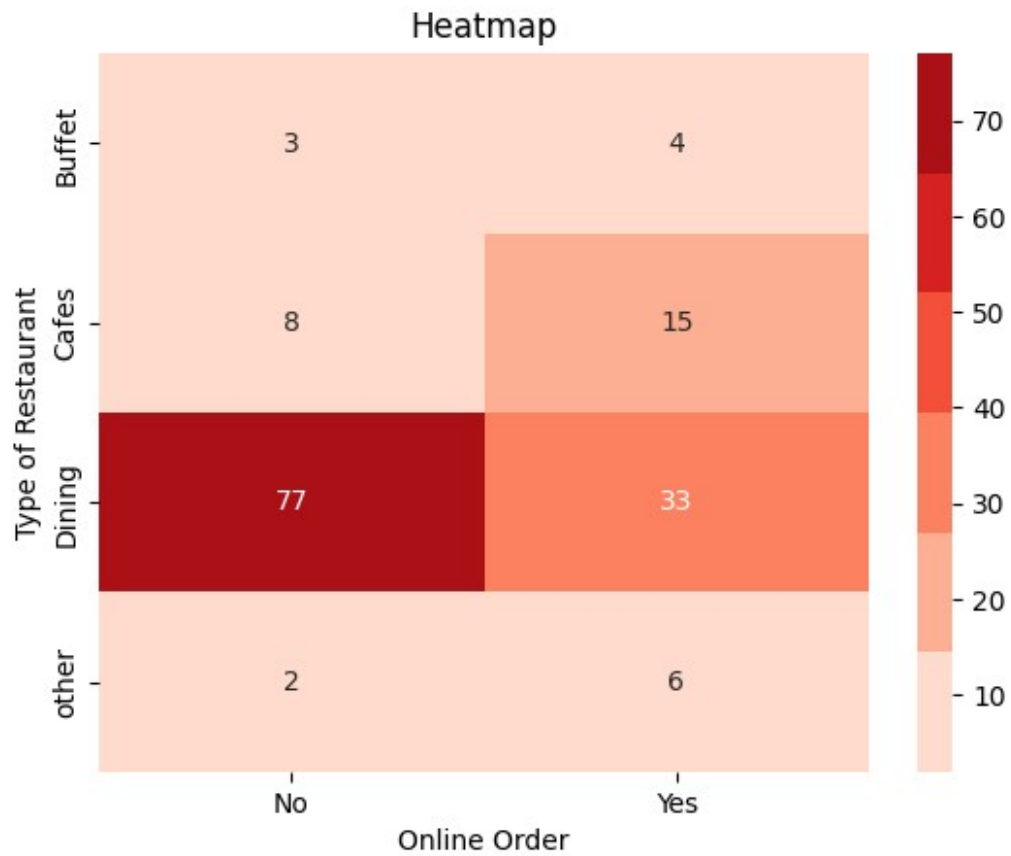
```
plt.figure(figsize = (4,4))
sns.boxplot(x='online_order', y='rate',data = data, color = 'red')
<Axes: xlabel='online_order', ylabel='rate'>
```



conclusion - Online order receives higher ratings in comparison to Offline orders

(6) which type of restaurant receives more offline order?

```
pivot_table = data.pivot_table(index = 'listed_in(type)', columns =  
'online_order', aggfunc='size',  
                                fill_value=0)  
colormap = sns.color_palette("Reds")  
sns.heatmap(pivot_table, annot=True, cmap =colormap,fmt='d')  
plt.title("Heatmap")  
plt.xlabel("Online Order")  
plt.ylabel("Type of Restaurant")  
plt.show()
```



conclusion- Dining restaurants receives more online and offline orders.