zomato data analysis project  In [1]: import pandas as pd import numpy as np import matplotlib.pyplot as plt import soaborn as sng		
<pre>import seaborn as sns  In [3]:     dataframe = pd.read_csv("Zomato data .csv")     print(dataframe)</pre>	\	
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		
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]		
In [4]:       dataframe         Out[4]:       name online_order book_table rate votes approx_cost(for the property of the property	two people) liste 800 800	d_in(type)  Buffet Buffet
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	800 300 600 	Buffet Buffet Dining
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         148 rows × 7 columns	150 450 800 200	Dining Dining Dining Dining
data cleaning  In []:		
<pre>def handleRate(value):     value = str(value).split('/')     value = value[0];     return float(value)  dataframe['rate'] = dataframe['rate'].apply(handleRate)     print(dataframe.head())</pre>		
name online_order book_table rate votes \ 0		
approx_cost(for two people) listed_in(type) 0		
<pre> <class 'pandas.core.frame.dataframe'=""> RangeIndex: 148 entries, 0 to 147 Data columns (total 7 columns):     # Column</class></pre>		
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 resturant  In [18]: dataframe.head()  Out[18]: name online_order book_table rate votes approx_cost(for two	people) listed_i	n(type)
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	800 800 800 300 600	Buffet Buffet Buffet Buffet Buffet
<pre>In [21]: sns.countplot(x=dataframe['listed_in(type)'])   plt.xlabel("type of resturant") Out[21]: Text(0.5, 0, 'type of resturant')</pre>		
100 - 80 - tig 60 -		
Buffet Cafes other Dining type of resturant		
conclusion-majority of the restaurant falls  In [23]: dataframe.head()  Out[23]: name online_order book_table rate votes approx_cost(for two		
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	800 800 800 300 600	Buffet Buffet Buffet Buffet Buffet
<pre>In [27]:     grouped_data =dataframe.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)</pre>	)	
Out[27]: Text(0, 0.5, 'votes')  20000 - 17500 - 15000 -		
12500 - 10000 - 7500 - 2500		
Type of restaurant  conclusion - dinning restaurant has rece  In [28]: dataframe.head()	eived ma	ximum votes
Out [28]:         name         online_order         book_table         rate         votes         approx_cost(for two           0         Jalsa         Yes         Yes         4.1         775           1         Spice Elephant         Yes         No         4.1         787           2         San Churro Cafe         Yes         No         3.8         918	800 800 800	Buffet Buffet Buffet
3 Addhuri Udupi Bhojana No No 3.7 88 4 Grand Village No No 3.8 166  In [30]: plt.hist(dataframe['rate'], bins = 5) plt.title("rating distribution") plt.show()	300 600	Buffet Buffet
rating distribution  50 -  40 -		
30 - 20 - 10 -		
conclusion - the majority restaurants rec	eived ra	ating from 3.5 to 4
Average order spending by couples  In [31]: dataframe.head()  Out[31]: name online_order book_table rate votes approx_cost(for two	people) listed i	n(type)
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	800 800 800 800 300 600	Buffet Buffet Buffet Buffet Buffet
<pre>In [32]: couple_data=dataframe['approx_cost(for two people)']     sns.countplot(x=couple_data) Out[32]: <axessubplot:xlabel='approx_cost(for people)',="" two="" ylabel="count"></axessubplot:xlabel='approx_cost(for></pre>		
20 -		
8 10 - 5 - 100150 200250 300350 400450 500 550600 650 700750 800850 900 950 approx_cost(for two people)		
conclusion - the majority of couples prefer	err resta	aurants with an approximate cost of 300 rupees
which mode recevies maximum rating  In [34]: dataframe.head()  Out[34]: name online_order book_table rate votes approx_cost(for two	people) listed i	n(type)
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	800 800 800 300 600	Buffet Buffet Buffet Buffet Buffet
<pre>In [35]: plt.figure(figsize = (6,6))     sns.boxplot(x = 'online_order', y= 'rate', data = dataframe) Out[35]: <axessubplot:xlabel='online_order', ylabel="rate"></axessubplot:xlabel='online_order',></pre>		
4.50 - 4.25 - 4.00 - 3.75 -		
3.25 -		
2.75  2.50  No online_order  conclusion offline order received lower	rotin	a comparison to opline ender
Conclusion - offline order received lower  In [36]: dataframe.head()  Out[36]: name online_order book_table rate votes approx_cost(for two online_order 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	800 800 300 600	Buffet Buffet Buffet Buffet
<pre>In [39]:     pivot_table = dataframe.pivot_table(index='listed_in(type)',colum     sns.heatmap(pivot_table, annot=True, cmap="YlGnBu",fmt='d')     plt.title("heatmap")     plt.xlabel("online order")     plt.ylabel("listed in (type)")     plt.show()</pre>	mns='online_o	rder', aggfunc='size', fill_value=0)
heatmap  - 70  - 60  - 50  - 40		
- 50 - 40 - 30 - 20 - 10 No Nes online order		
	y accep es orders	t offline orders, whereas cafes primarily receive online s in person at restaurants, but prefer online ordering at cafes.