



Avadheshkumar04 / Avadheshkumar04

Type / to search



&lt;&gt; Code



Pull requests



Actions



Projects



Security



Insights



Settings

Avadheshkumar04 / Zomato\_Restaurant\_Dataset\_EDA.ipynb



Avadheshkumar04 Created using Colab

5515dd9 · now



History



Avadheshkumar04 / Zomato\_Restaurant\_Dataset\_EDA.ipynb

↑ Top

Preview

Code

Blame

5629 lines (5629 loc) · 1.31 MB

Raw





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

%matplotlib inline
import warnings
warnings.filterwarnings('ignore')
```

```
In [ ]: df = pd.read_csv("/content/zomato.csv",encoding='latin-1')
```

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

Out[ ]:

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Locality Verbose	Longitude	Latitude	Cuisines	...	Currency
0	6317637	Le Petit Souffle	162	Makati City	Third Floor, Century City Mall, Kalayaan Avenu...	Century City Mall, Poblacion, Makati City	Century City Mall, Poblacion, Makati City, Mak...	121.027535	14.565443	French, Japanese, Desserts	...	Botswana Pula(P)
1	6304287	Izakaya Kikufuji	162	Makati City	Little Tokyo, 2277 Chino Roces Avenue, Legaspi...	Little Tokyo, Legaspi Village, Makati City	Little Tokyo, Legaspi Village, Makati City, Ma...	121.014101	14.553708	Japanese	...	Botswana Pula(P)
					Edsa Shangri-	Edsa Shangri-	Edsa Shangri-			Seafood		

2	6300002	Heat - Edsa Shangri-La	162	Mandaluyong City	La, 1 Garden Way, Ortigas, Mandal...	Edsa Shangri-La, Ortigas, Mandaluyong City	Edsa Shangri-La, Ortigas, Mandaluyong City, Ma...	121.056831	14.581404	Asian, Filipino, Indian	...	Botswana Pula(P)
3	6318506	Ooma	162	Mandaluyong City	Third Floor, Mega Fashion Hall, SM Megamall, O...	SM Megamall, Ortigas, Mandaluyong City	SM Megamall, Ortigas, Mandaluyong City, Mandal...	121.056475	14.585318	Japanese, Sushi	...	Botswana Pula(P)
4	6314302	Sambo Kojin	162	Mandaluyong City	Third Floor, Mega Atrium, SM Megamall, Ortigas...	SM Megamall, Ortigas, Mandaluyong City	SM Megamall, Ortigas, Mandaluyong City, Mandal...	121.057508	14.584450	Japanese, Korean	...	Botswana Pula(P)

5 rows × 21 columns



In [ ]:

```
df.sample(5)
```

Out[ ]:

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Locality Verbose	Longitude	Latitude	Cuisines	...	Cu
6992	7603	Stuti Restaurant	1	New Delhi	6, Satyaniketan, New Delhi	Satyaniketan	Satyaniketan, New Delhi	77.168962	28.587512	North Indian	...	Rupe
4902	8517	Apni Rasoi	1	New Delhi	A2, Ganesh Nagar, Laxmi Nagar, New Delhi	Laxmi Nagar	Laxmi Nagar, New Delhi	77.277690	28.630719	North Indian	...	Rupe

141	17294850	Thai Kitchen	216	Augusta	4357 Washington Road, Evans, GA 30809	Evans	Evans, Augusta	-82.132800	33.540600	Thai	...	Dc
743	58882	Big Brewsky	1	Bangalore	Behind MK Retail, Before WIPRO Corporate Offic...	Sarjapur Road	Sarjapur Road, Bangalore	77.683237	12.913041	Finger Food, North Indian, Italian, Continenta...	...	Rupe
3581	18381663	Khan Chacha	1	New Delhi	Lower Ground Floor, FCCE_8, Epicuria, Nehru Pla...	Epicuria Food Mall, Nehru Place	Epicuria Food Mall, Nehru Place, New Delhi	77.251426	28.551456	North Indian, Mughlai	...	Rupe

5 rows × 21 columns



In [ ]: df.shape

Out[ ]: (9551, 21)

In [ ]: df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9551 entries, 0 to 9550
Data columns (total 21 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Restaurant ID          9551 non-null   int64
1   Restaurant Name        9551 non-null   object
2   Country Code           9551 non-null   int64
3   City                   9551 non-null   object
4   Address                9551 non-null   object
5   Locality               9551 non-null   object
6   Locality Verbose       9551 non-null   object
7   Longitude              9551 non-null   float64
```

```

7  Longitude      9551 non-null    float64
8  Latitude       9551 non-null    float64
9  Cuisines       9542 non-null    object
10 Average Cost for two  9551 non-null    int64
11 Currency       9551 non-null    object
12 Has Table booking  9551 non-null    object
13 Has Online delivery  9551 non-null    object
14 Is delivering now  9551 non-null    object
15 Switch to order menu  9551 non-null    object
16 Price range     9551 non-null    int64
17 Aggregate rating  9551 non-null    float64
18 Rating color     9551 non-null    object
19 Rating text      9551 non-null    object
20 Votes           9551 non-null    int64

```

dtypes: float64(3), int64(5), object(13)

memory usage: 1.5+ MB

Observation

1. The Dataset contains some null value in the cuisines coloumn.
2. There are total 13 categorical varibales in the datasets .

In [ ]: df.describe()

Out[ ]:

	Restaurant ID	Country Code	Longitude	Latitude	Average Cost for two	Price range	Aggregate rating	Votes
<b>count</b>	9.551000e+03	9551.000000	9551.000000	9551.000000	9551.000000	9551.000000	9551.000000	9551.000000
<b>mean</b>	9.051128e+06	18.365616	64.126574	25.854381	1199.210763	1.804837	2.666370	156.909748
<b>std</b>	8.791521e+06	56.750546	41.467058	11.007935	16121.183073	0.905609	1.516378	430.169145
<b>min</b>	5.300000e+01	1.000000	-157.948486	-41.330428	0.000000	1.000000	0.000000	0.000000
<b>25%</b>	3.019625e+05	1.000000	77.081343	28.478713	250.000000	1.000000	2.500000	5.000000
<b>50%</b>	6.004089e+06	1.000000	77.191964	28.570469	400.000000	2.000000	3.200000	31.000000
<b>75%</b>	1.835229e+07	1.000000	77.282006	28.642758	700.000000	2.000000	3.700000	131.000000
<b>max</b>	1.850065e+07	216.000000	174.832089	55.976980	800000.000000	4.000000	4.900000	10934.000000

. Have a look on Avg,Min,Max values of Prices , Rating and votes

```
In [ ]: df.isnull().sum()
```

```
Out[ ]: Restaurant ID      0
Restaurant Name      0
Country Code        0
City                0
Address             0
Locality            0
Locality Verbose    0
Longitude           0
Latitude            0
Cuisines            9
Average Cost for two 0
Currency            0
Has Table booking   0
Has Online delivery 0
Is delivering now   0
Switch to order menu 0
Price range         0
Aggregate rating     0
Rating color        0
Rating text         0
Votes              0
dtype: int64
```

```
In [ ]: #Remove the duplicates
df.duplicated().sum()
```

```
Out[ ]: 0
```

```
In [ ]: df.dropna(inplace=True)
```

```
In [ ]: # Correlation between required features
df[["Average Cost for two", "Price range", "Aggregate rating", "Votes"]].corr()
```

```
Out[ ]:
```

	Average Cost for two	Price range	Aggregate rating	Votes
Average Cost for two	1.000000	0.075111	0.051864	0.067833

<b>Price range</b>	0.075111	1.000000	0.438356	0.309474
<b>Aggregate rating</b>	0.051864	0.438356	1.000000	0.313474
<b>Votes</b>	0.067833	0.309474	0.313474	1.000000

In [ ]: `df.iloc[0: 1]`

Out[ ]:

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Locality Verbose	Longitude	Latitude	Cuisines	...	Currency	Has Table booking	...
0	6317637	Le Petit Souffle	162	Makati City	Third Floor, Century City Mall, Kalayaan Avenu...	Century City Mall, Poblacion, Makati City	Century City Mall, Poblacion, Makati City, Mak...	121.027535	14.565443	French, Japanese, Desserts	...	Botswana Pula(P)	Yes	

1 rows × 21 columns



Warning: Total number of columns (21) exceeds max\_columns (20) limiting to first (20) columns.

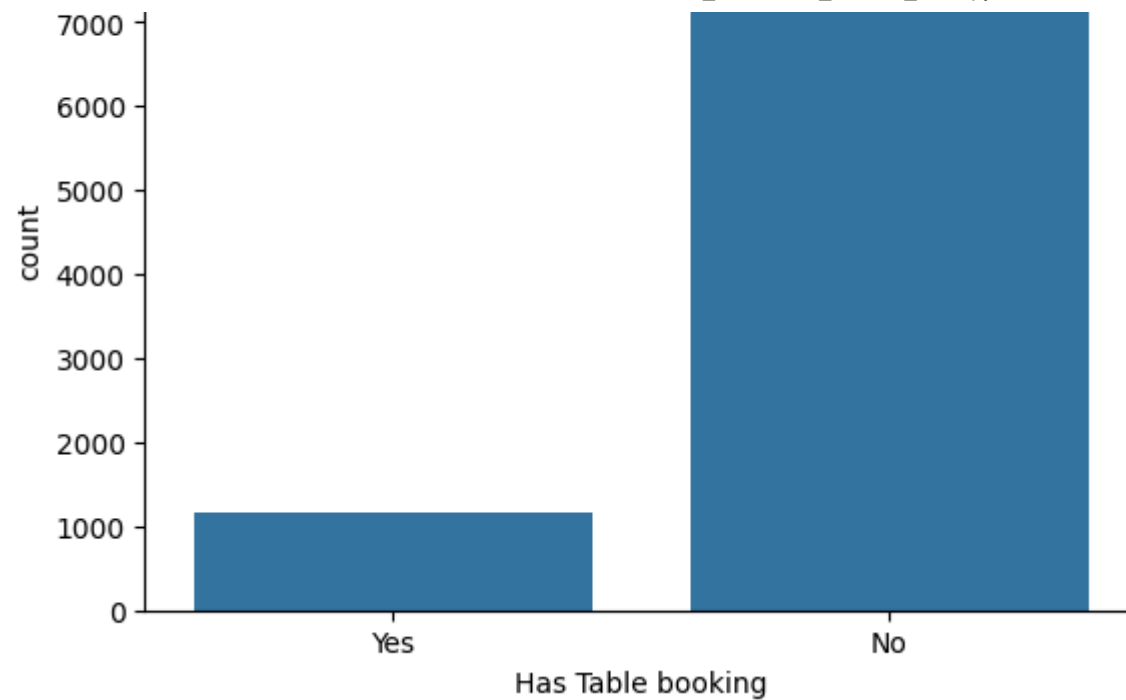
Warning: Total number of columns (21) exceeds max\_columns (20) limiting to first (20) columns.

Observation ,countplot for Table booking and not ,or customer online Delivery and not

In [ ]: `# Analysing costumer has table bookings or not`  
`sns.countplot(x='Has Table booking', data=df)`

Out[ ]: <Axes: xlabel='Has Table booking', ylabel='count'>





```
In [ ]: # Analysing the customer has online delivery and not  
sns.countplot(x='Has Online delivery', data=df)
```

```
Out[ ]: <Axes: xlabel='Has Online delivery', ylabel='count'>
```



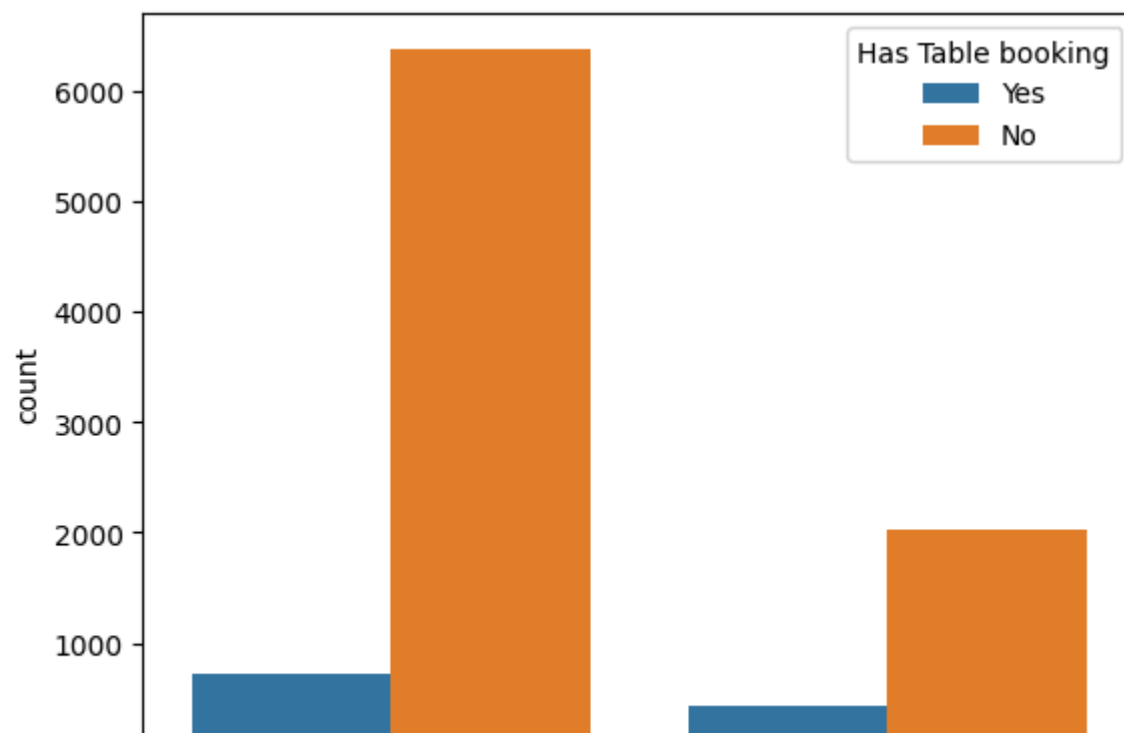




Some of the customer has online delivery

```
In [ ]: #Analysing the no of delivery in present  
  
sns.countplot(x='Has Online delivery', hue='Has Table booking', data=df)
```

```
Out[ ]: <Axes: xlabel='Has Online delivery', ylabel='count'>
```

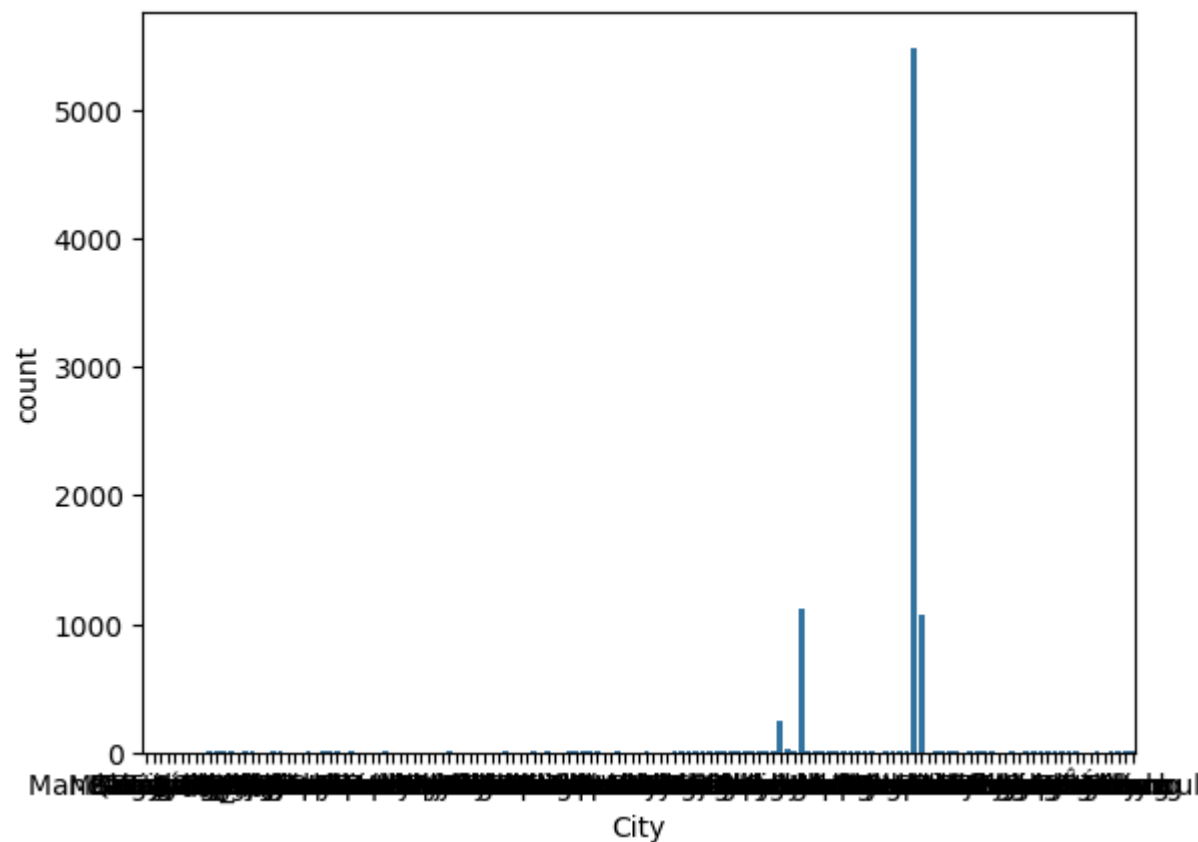




Observation : Very less deliveries

```
In [ ]: sns.countplot(x='City', data=df)
```

```
Out[ ]: <Axes: xlabel='City', ylabel='count'>
```



```
In [ ]: df_country = pd.read_excel("/content/Country-Code.xlsx")
```

```
In [ ]: df_country
```

Out[ ]:

	Country Code	Country
0	1	India
1	14	Australia
2	30	Brazil
3	37	Canada
4	94	Indonesia
5	148	New Zealand
6	162	Phillipines
7	166	Qatar
8	184	Singapore
9	189	South Africa
10	191	Sri Lanka
11	208	Turkey
12	214	UAE
13	215	United Kingdom
14	216	United States

In [ ]:

```
final_df = pd.merge(df, df_country, on='Country Code', how='left')
```

In [ ]:

```
final_df
```

Out[ ]:

Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Locality Verbose	Longitude	Latitude	Cuisines	...	b
				Third Floor,		Century City					
						Century City					

0	6317637	Le Petit Souffle	162	Makati City	Century City Mall, Kalayaan Avenu...	Century City Mall, Poblacion, Makati City	Mall, Poblacion, Makati City, Mak...	121.027535	14.565443	French, Japanese, Desserts	...
1	6304287	Izakaya Kikufuji	162	Makati City	Little Tokyo, 2277 Chino Roces Avenue, Legaspi...	Little Tokyo, Legaspi Village, Makati City	Little Tokyo, Legaspi Village, Makati City, Ma...	121.014101	14.553708	Japanese	...
2	6300002	Heat - Edsa Shangri-La	162	Mandaluyong City	Edsa Shangri-La, 1 Garden Way, Ortigas, Mandal...	Edsa Shangri-La, Ortigas, Mandaluyong City	Edsa Shangri-La, Ortigas, Mandaluyong City, Ma...	121.056831	14.581404	Seafood, Asian, Filipino, Indian	...
3	6318506	Ooma	162	Mandaluyong City	Third Floor, Mega Fashion Hall, SM Megamall, O...	SM Megamall, Ortigas, Mandaluyong City	SM Megamall, Ortigas, Mandaluyong City, Mandal...	121.056475	14.585318	Japanese, Sushi	...
4	6314302	Sambo Kojin	162	Mandaluyong City	Third Floor, Mega Atrium, SM Megamall, Ortigas...	SM Megamall, Ortigas, Mandaluyong City	SM Megamall, Ortigas, Mandaluyong City, Mandal...	121.057508	14.584450	Japanese, Korean	...
...	...	...	...	...	...	...	...	...	...	...	...
9546	5915730	NamlÛ± Gurme	208	ÛÁstanbul	Kemanke±ô Karamustafa Pa±ôa Mahallesi, RÛ±htÛ±...	Karakí_y	Karakí_y, ÛÁstanbul	28.977392	41.022793	Turkish	...
9547	5908749	Ceviz AÛôacÛ±	208	ÛÁstanbul	Ko±ôuyolu Mahallesi, Muhittin îstí_ndaÛô Cadd...	Ko±ôuyolu	Ko±ôuyolu, ÛÁstanbul	29.041297	41.009847	World Cuisine, Patisserie, Cafe	...

9548	5915807	Huqqa	208	Üstanbul	Kuruí_eðme Mahallesi, Muallim Naci Caddesi, N...	Kuruí_eðme	Kuruí_eðme, Üstanbul	29.034640	41.055817	Italian, World Cuisine	...
9549	5916112	Αððððk Kahve	208	Üstanbul	Kuruí_eðme Mahallesi, Muallim Naci Caddesi, N...	Kuruí_eðme	Kuruí_eðme, Üstanbul	29.036019	41.057979	Restaurant Cafe	...
9550	5927402	Walter's Coffee Roastery	208	Üstanbul	CafeaÜöa Mahallesi, BademaltÜ± Sokak, No 21/B,...	Moda	Moda, Üstanbul	29.026016	40.984776	Cafe	...

9551 rows × 22 columns



## Display the datatype

```
In [ ]: final_df.dtypes
```

```
Out[ ]: Restaurant ID      int64
Restaurant Name      object
Country Code        int64
City                object
Address             object
Locality            object
Locality Verbose    object
Longitude           float64
Latitude            float64
Cuisines            object
Average Cost for two  int64
Currency            object
Has Table booking    object
Has Online delivery  object
Is delivering now    object
```

```
Switch to order menu    object
Price range             int64
Aggregate rating        float64
Rating color            object
Rating text             object
Votes                  int64
Country                 object
dtype: object
```

## Top 3 Country with Maximum number of order

```
In [ ]: final_df.Country.value_counts()
```

```
Out[ ]: Country
India            8652
United States    434
United Kingdom   80
Brazil           60
UAE              60
South Africa     60
New Zealand      40
Turkey           34
Australia        24
Phillipines      22
Indonesia        21
Singapore        20
Qatar            20
Sri Lanka        20
Canada           4
Name: count, dtype: int64
```

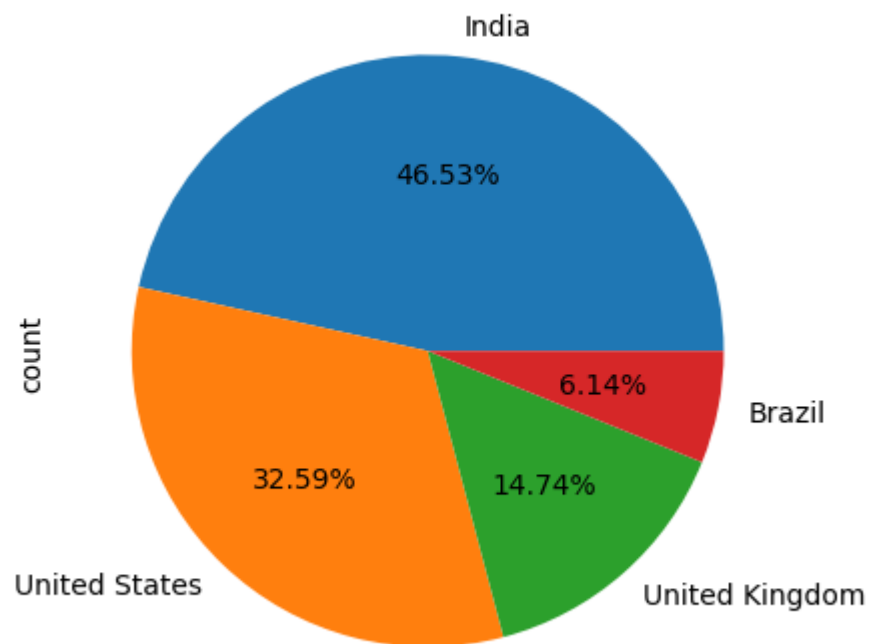
```
In [ ]: #final_df['Price range'].value_counts().plot(kind='pie', autopct='%.2f%%')

# Assuming final_df is your DataFrame containing 'Price range' and 'Country' columns

final_df['Price range'].value_counts().plot(kind='pie', autopct='%.2f%%', labels=final_df.Country.value_counts().index)
plt.title('Price range distribution by Country')
```

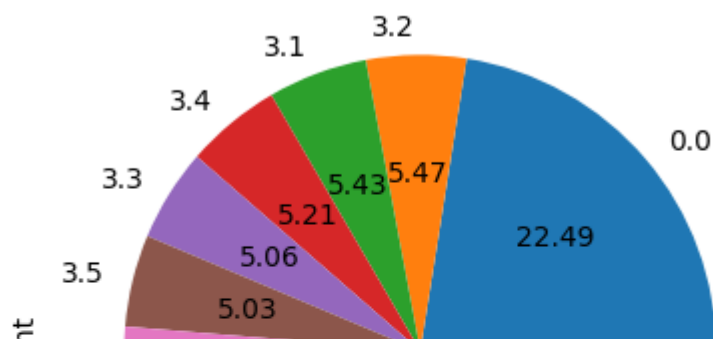
```
Out[ ]: Text(0.5, 1.0, 'Price range distribution by Country')
```

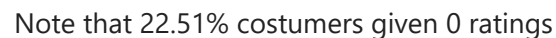
## Price range distribution by Country



```
In [ ]: final_df['Aggregate rating'].value_counts().plot(kind='pie', autopct='%.2f')
plt.title('Aggregate rating distribution')
plt.show()
```

## Aggregate rating distribution





Out[ ]:

	Aggregate rating	Rating color	Rating text	0
0	0.0	White	Not rated	2148
1	1.8	Red	Poor	1
2	1.9	Red	Poor	2
3	2.0	Red	Poor	7
4	2.1	Red	Poor	15
5	2.2	Red	Poor	27
6	2.3	Red	Poor	47
7	2.4	Red	Poor	87
8	2.5	Orange	Average	110
9	2.6	Orange	Average	191



<b>10</b>	2.7	Orange	Average	250
<b>11</b>	2.8	Orange	Average	315
<b>12</b>	2.9	Orange	Average	381
<b>13</b>	3.0	Orange	Average	468
<b>14</b>	3.1	Orange	Average	519
<b>15</b>	3.2	Orange	Average	522
<b>16</b>	3.3	Orange	Average	483
<b>17</b>	3.4	Orange	Average	498
<b>18</b>	3.5	Yellow	Good	480
<b>19</b>	3.6	Yellow	Good	458
<b>20</b>	3.7	Yellow	Good	427
<b>21</b>	3.8	Yellow	Good	400
<b>22</b>	3.9	Yellow	Good	335
<b>23</b>	4.0	Green	Very Good	266
<b>24</b>	4.1	Green	Very Good	274
<b>25</b>	4.2	Green	Very Good	221
<b>26</b>	4.3	Green	Very Good	174
<b>27</b>	4.4	Green	Very Good	144
<b>28</b>	4.5	Dark Green	Excellent	95
<b>29</b>	4.6	Dark Green	Excellent	78
<b>30</b>	4.7	Dark Green	Excellent	42
<b>31</b>	4.8	Dark Green	Excellent	25
<b>32</b>	4.9	Dark Green	Excellent	61

INCIDENCE

## INFERENCE

Rating color is WHITE when it is NOT RATED Rating color is RED when it is POOR Rating color is ORANGE when it is AVERAGE Rating color is YELLOW when it is GOOD Rating color is GREEN when it is VERY GOOD Rating color is DARK GREEN when it is EXCELLENT

```
In [ ]: #ratings = final_df.groupby(['Aggregate rating', 'Rating color', 'Rating text']).size().reset_index().rename(columns={0: 'Rating Count'})
#ratings
```

```
ratings = final_df.groupby(['Aggregate rating', 'Rating color', 'Rating text']).size().reset_index().rename(columns={0: 'Rating Count'})
ratings
```

```
Out[ ]:
```

	Aggregate rating	Rating color	Rating text	Rating Count
0	0.0	White	Not rated	2148
1	1.8	Red	Poor	1
2	1.9	Red	Poor	2
3	2.0	Red	Poor	7
4	2.1	Red	Poor	15
5	2.2	Red	Poor	27
6	2.3	Red	Poor	47
7	2.4	Red	Poor	87
8	2.5	Orange	Average	110
9	2.6	Orange	Average	191
10	2.7	Orange	Average	250
11	2.8	Orange	Average	315
12	2.9	Orange	Average	381
13	3.0	Orange	Average	468
14	3.1	Orange	Average	510

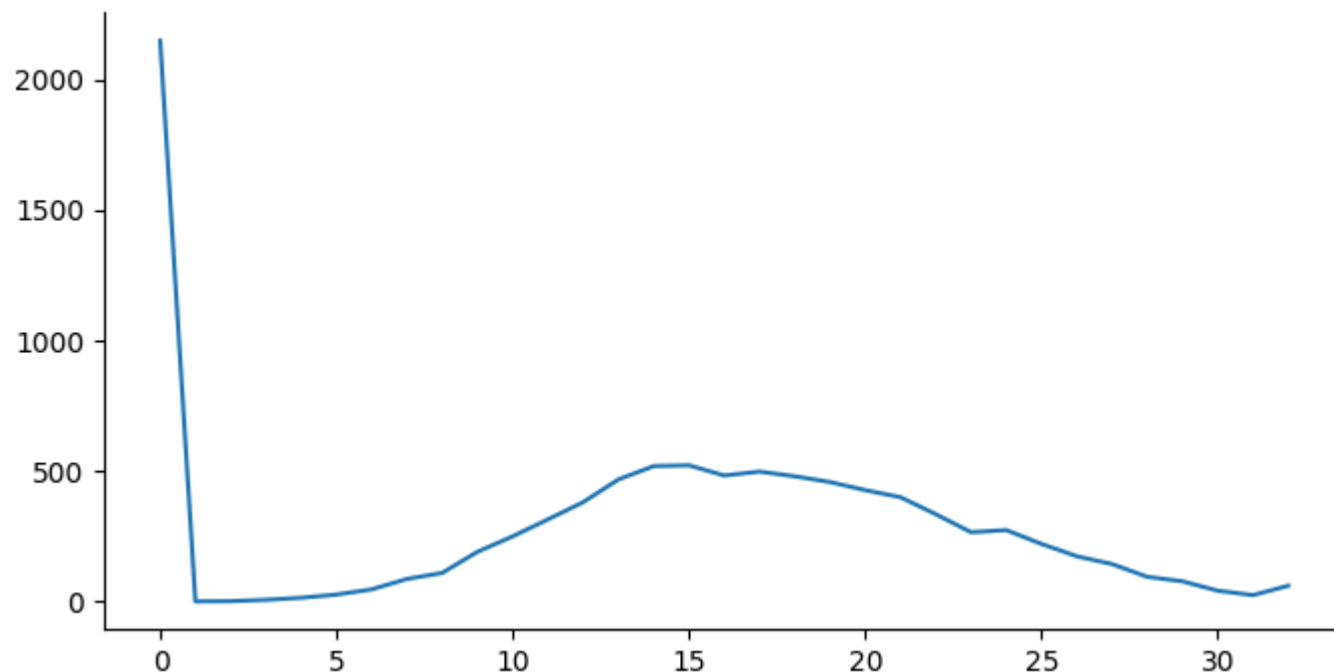
14	3.1	Orange	Average	512
15	3.2	Orange	Average	522
16	3.3	Orange	Average	483
17	3.4	Orange	Average	498
18	3.5	Yellow	Good	480
19	3.6	Yellow	Good	458
20	3.7	Yellow	Good	427
21	3.8	Yellow	Good	400
22	3.9	Yellow	Good	335
23	4.0	Green	Very Good	266
24	4.1	Green	Very Good	274
25	4.2	Green	Very Good	221
26	4.3	Green	Very Good	174
27	4.4	Green	Very Good	144
28	4.5	Dark Green	Excellent	95
29	4.6	Dark Green	Excellent	78
30	4.7	Dark Green	Excellent	42
31	4.8	Dark Green	Excellent	25
32	4.9	Dark Green	Excellent	61

In [ ]:

```
# @title Rating Count

from matplotlib import pyplot as plt
ratings['Rating Count'].plot(kind='line', figsize=(8, 4), title='Rating Count')
plt.gca().spines[['top', 'right']].set_visible(False)
```

## Rating Count



INFERENCE--> Around 2k people have given zero ratings. ratings (4.5-4.9)=excellent. 4.0 to 4.4=v.good. 3.5 to 3.9=good. 2.5 to 3.4 = average and so on .

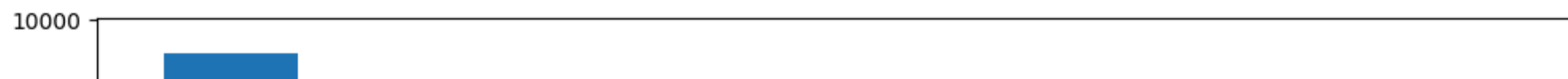
In [ ]:

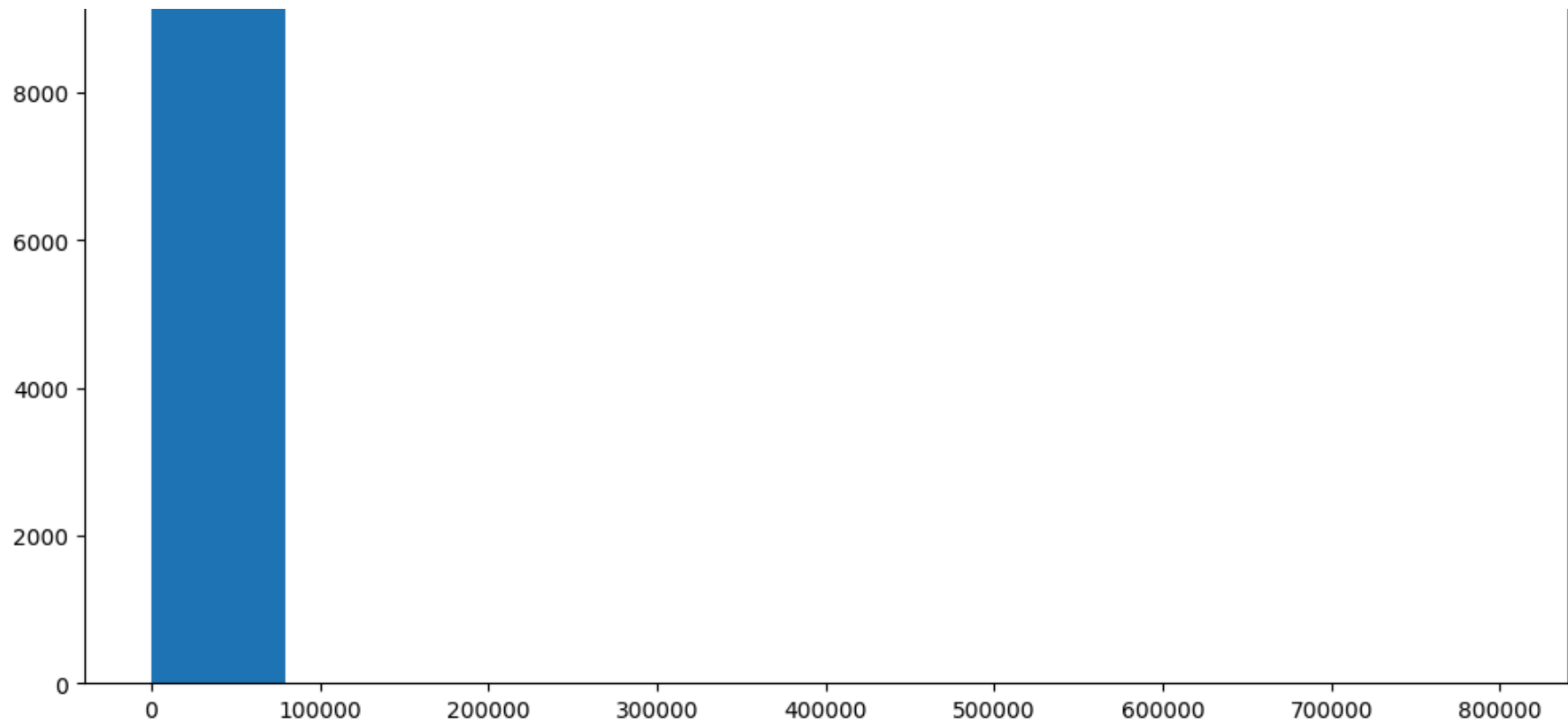
## Histogram

In [ ]:

```
plt.hist(final_df['Average Cost for two'])
```

Out[ ]: (array([9.531e+03, 4.000e+00, 7.000e+00, 4.000e+00, 1.000e+00, 1.000e+00,  
1.000e+00, 0.000e+00, 0.000e+00, 2.000e+00]),  
array([ 0., 80000., 160000., 240000., 320000., 400000., 480000.,  
560000., 640000., 720000., 800000.]),  
<BarContainer object of 10 artists>)



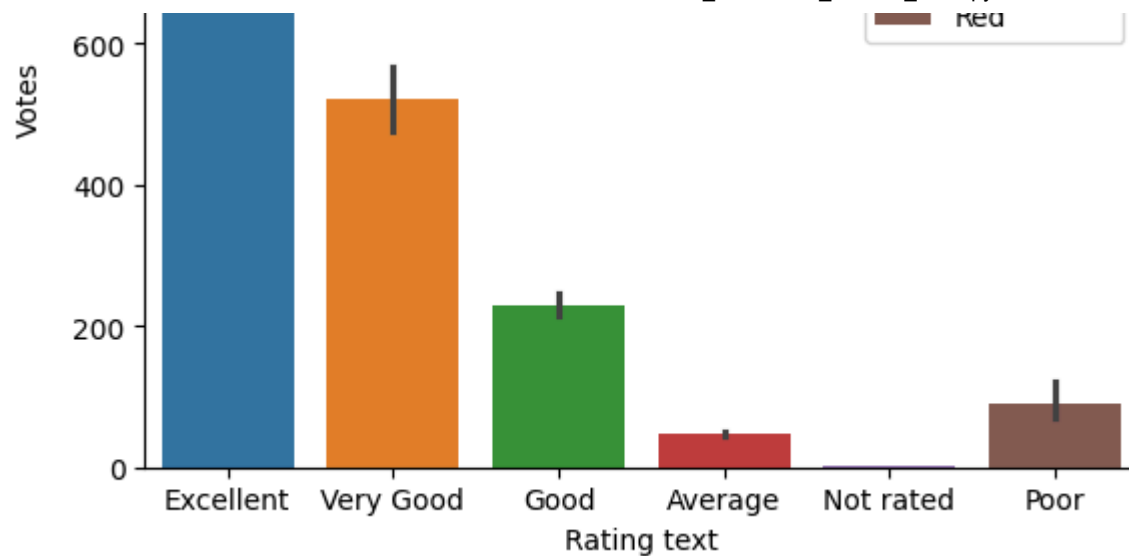


## Bar Plot

```
In [ ]: sns.barplot(x=final_df["Rating text"],y=df["Votes"],hue =df["Rating color"])
```

```
Out[ ]: <Axes: xlabel='Rating text', ylabel='Votes'>
```





```
In [ ]: hue =final_df["Rating color"]
hue
```

```
Out[ ]: 0      Dark Green
1      Dark Green
2       Green
3      Dark Green
4      Dark Green
...
9546    Green
9547    Green
9548    Yellow
9549    Green
9550    Green
Name: Rating color, Length: 9551, dtype: object
```

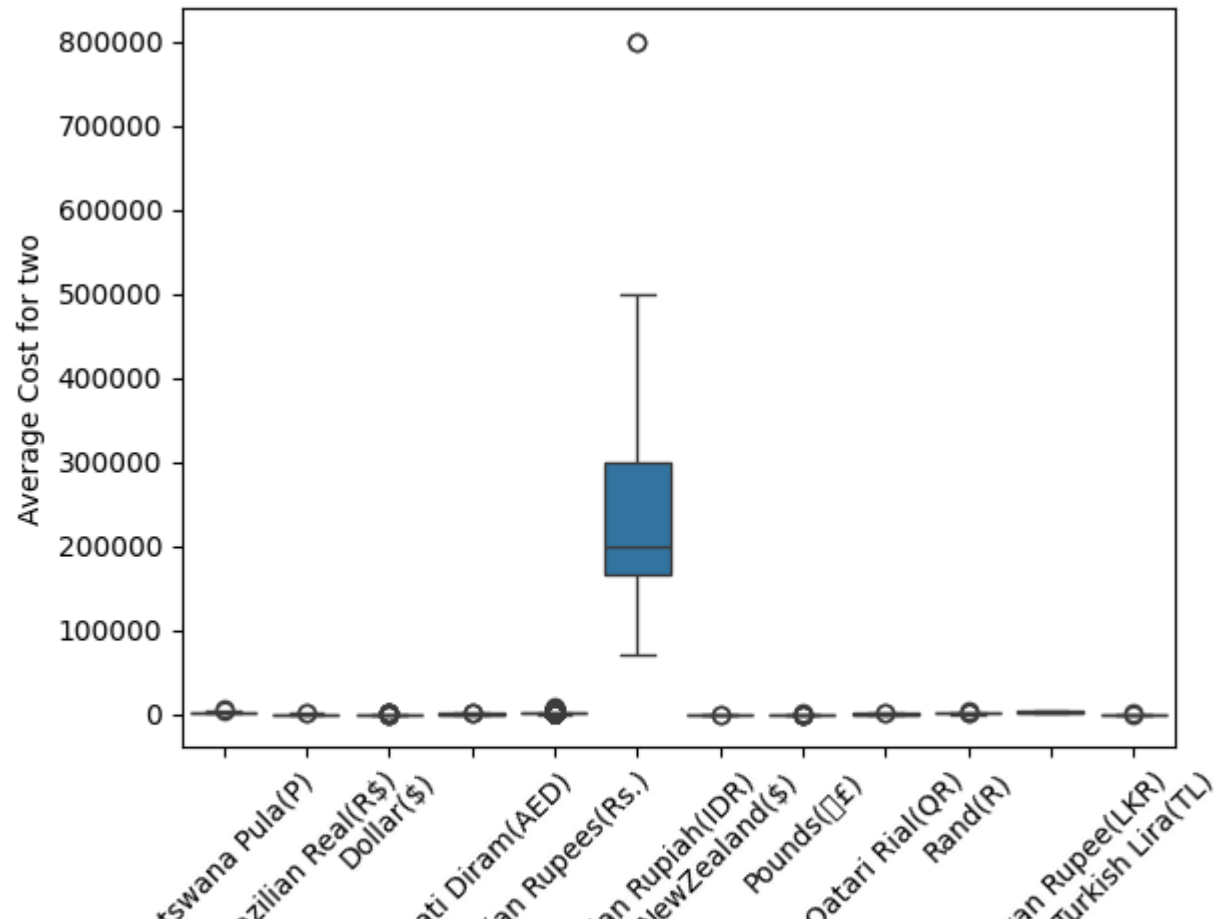
## New Section

## box plot

```
In [ ]: sns.boxplot(x=final_df['Currency'], y=final_df['Average Cost for two'])
```

```
plt.xticks(rotation=45)
```

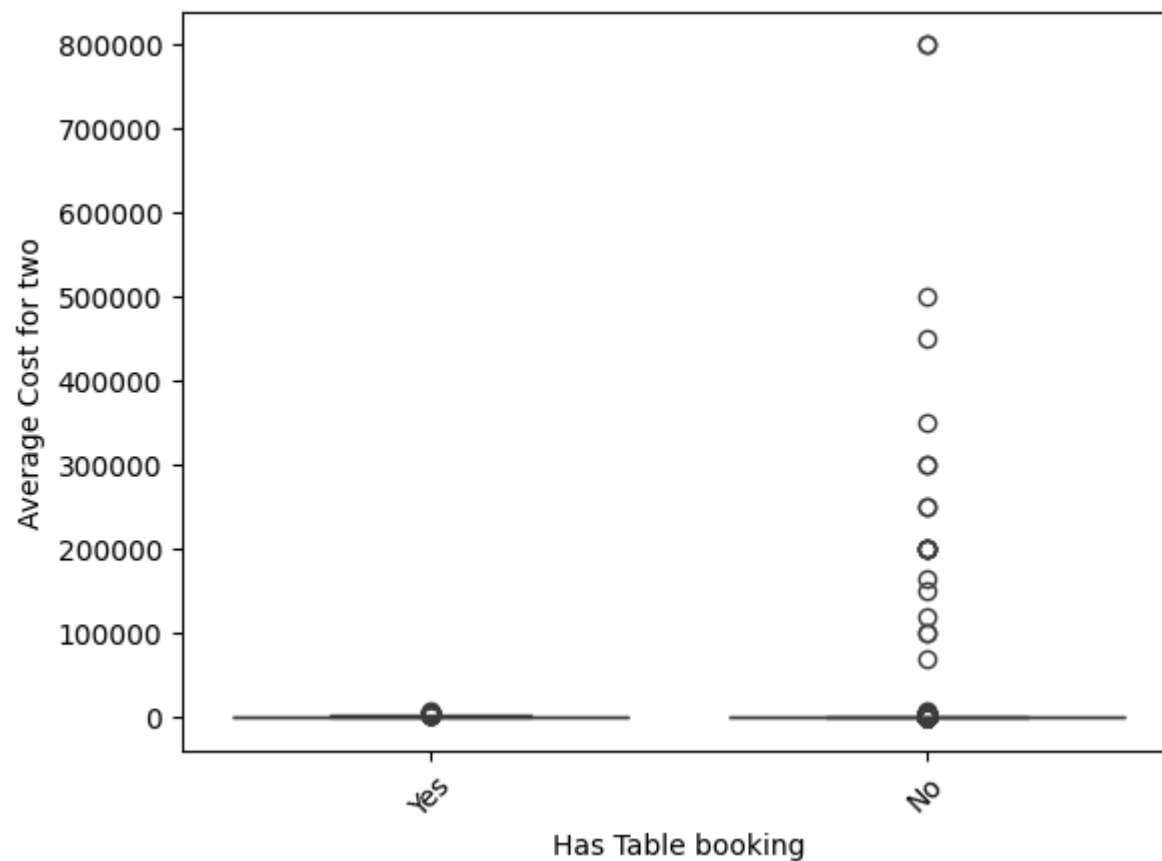
```
Out[ ]: ([0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11],
 [Text(0, 0, 'Botswana Pula(P)'),
  Text(1, 0, 'Brazilian Real(R$)'),
  Text(2, 0, 'Dollar($)'),
  Text(3, 0, 'Emirati Diram(AED)'),
  Text(4, 0, 'Indian Rupees(Rs.)'),
  Text(5, 0, 'Indonesian Rupiah(IDR)'),
  Text(6, 0, 'NewZealand($)'),
  Text(7, 0, 'Pounds(\x8c\x99)'),
  Text(8, 0, 'Qatari Rial(QR)'),
  Text(9, 0, 'Rand(R)'),
  Text(10, 0, 'Sri Lankan Rupee(LKR)'),
  Text(11, 0, 'Turkish Lira(TL)')])
```



Bot  
Bra  
Emira  
India  
Indonesia  
N  
Sri Lanka  
Currency

```
In [ ]: sns.boxplot(x=final_df['Has Table booking'], y=final_df['Average Cost for two'])  
plt.xticks(rotation=45)
```

```
Out[ ]: ([0, 1], [Text(0, 0, 'Yes'), Text(1, 0, 'No')])
```

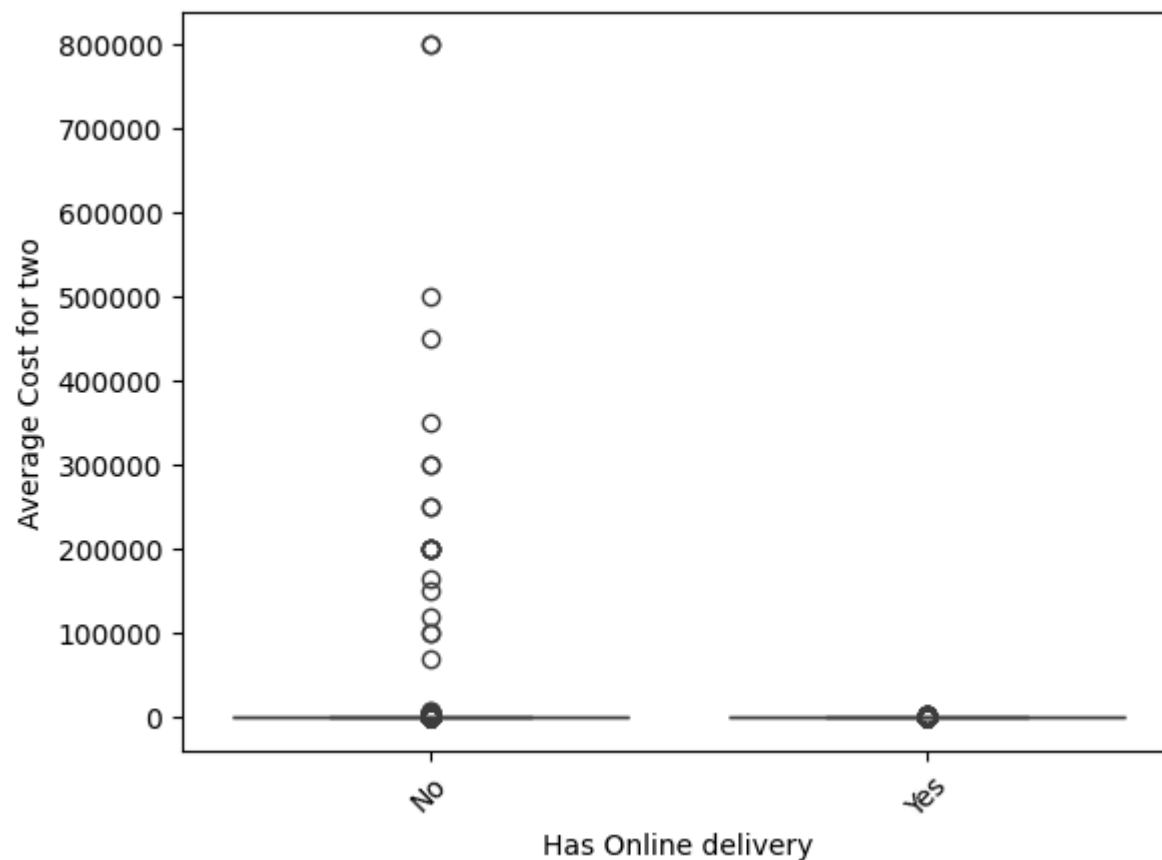


```
In [ ]: sns.boxplot(x=final_df['Has Online delivery'], y=final_df['Average Cost for two'])  
plt.xticks(rotation=45)
```

```
Out[ ]: ([0, 1], [Text(0, 0, 'No'), Text(1, 0, 'Yes')])
```



```
Out[ ]: ([0, 1], [text(0, 0, 'No'), text(1, 0, 'Yes')])
```

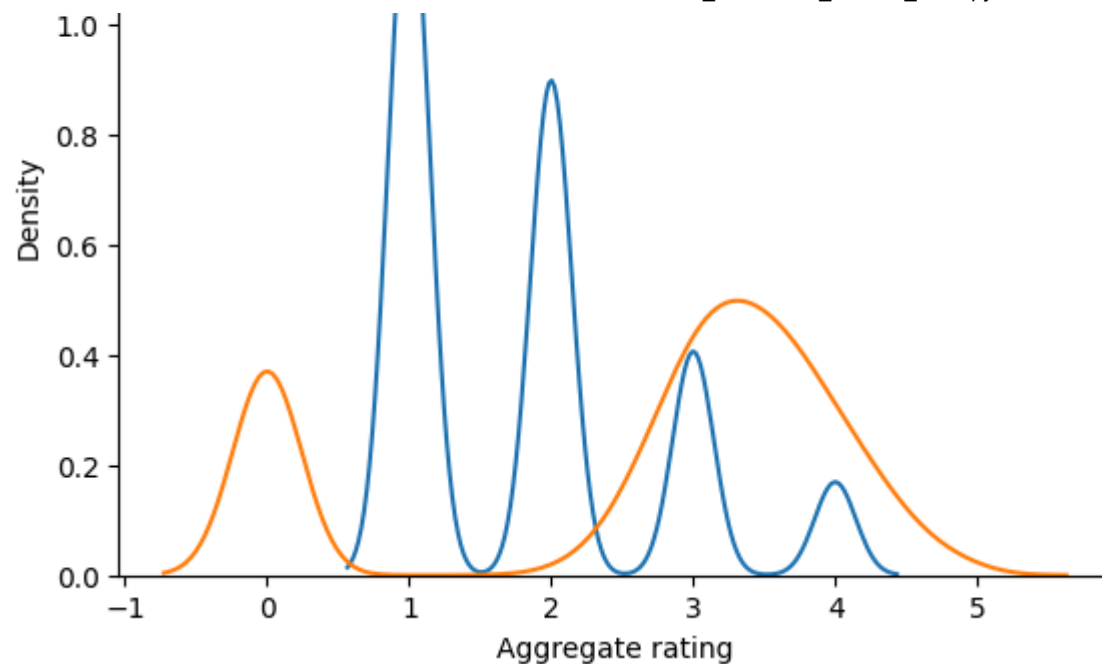


## Distplot

```
In [ ]: sns.distplot(final_df['Price range'],hist= False)
sns.distplot(final_df['Aggregate rating'],hist= False)
```

```
Out[ ]: <Axes: xlabel='Aggregate rating', ylabel='Density'>
```





Observation :-> Amazing informatios by this Probability Density Function Price range "1" has given no ratings Price range "3 and 4" has given ratings between "2.0 to 4.9

```
In [ ]: pd.crosstab(final_df['Is delivering now'],final_df['Has Online delivery'])
```

```
Out[ ]:  Has Online delivery  No  Yes
         Is delivering now
         No  7100  2417
         Yes   0    34
```

## HeatMap

```
In [ ]: sns.heatmap(pd.crosstab(final_df['Is delivering now'],final_df['Has Online delivery']))
```

```
Out[ ]: <Axes: xlabel='Has Online delivery', ylabel='Is delivering now'>
```

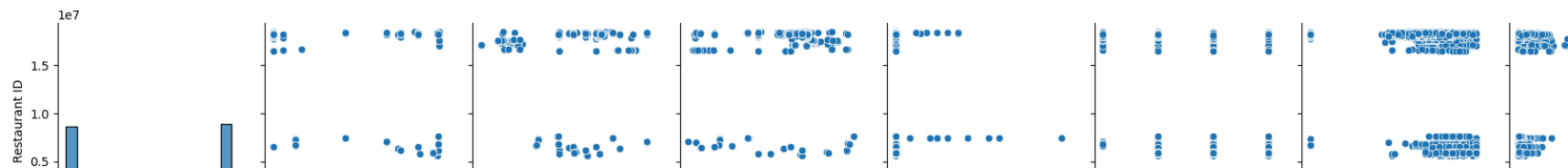


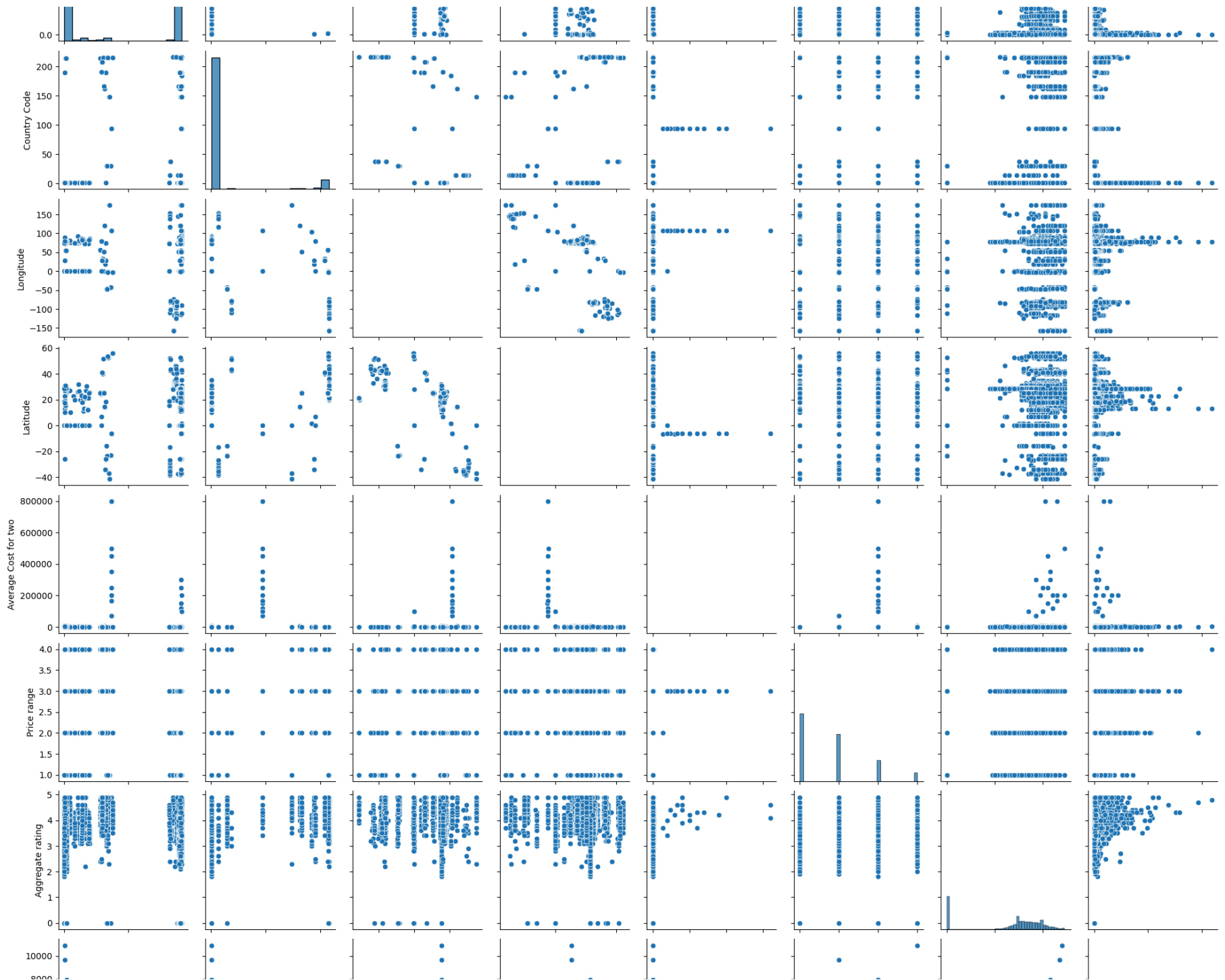
Observation:- this heatmap gives a better understanding , less no. of delivering the order recive

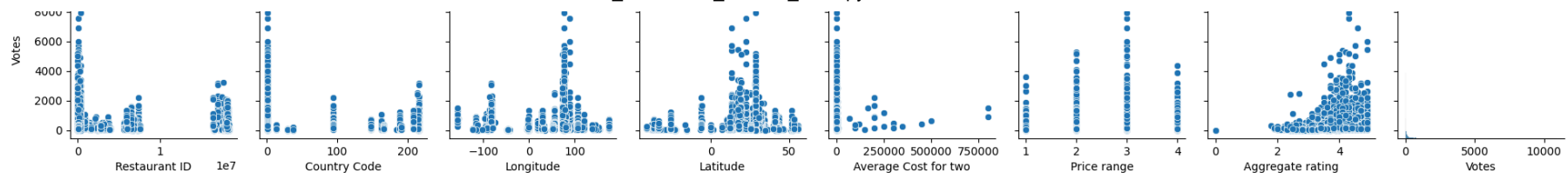
## Pairplot

```
In [ ]: sns.pairplot(final_df)
```

```
Out[ ]: <seaborn.axisgrid.PairGrid at 0x7db1c45c05b0>
```





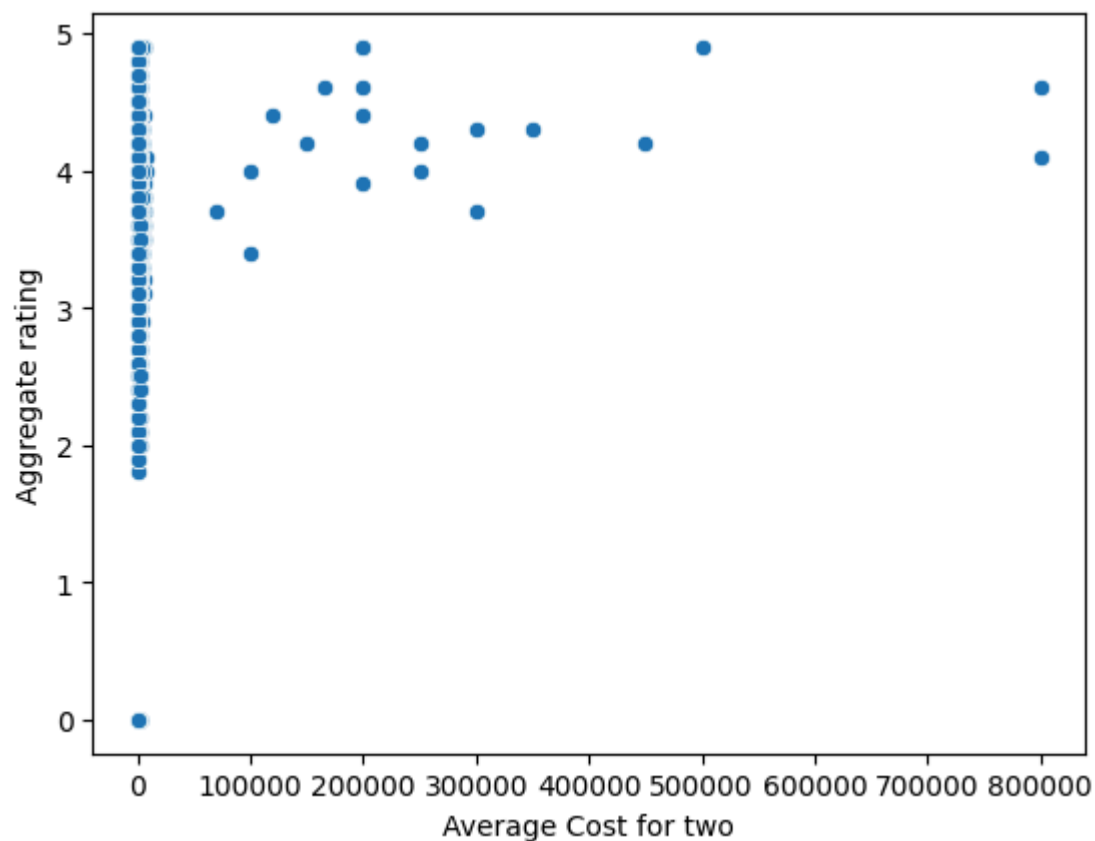


Since pairplot is not required here. But just to give a Bird Eye of Scatter plots

## scatterplot

```
In [ ]: sns.scatterplot(x=final_df['Average Cost for two'],y=final_df['Aggregate rating'])
```

```
Out[ ]: <Axes: xlabel='Average Cost for two', ylabel='Aggregate rating'>
```



```
In [ ]: sns.lineplot(y=final_df['Votes'],x=final_df['Has Online delivery'])
```

```
Out[ ]: <Axes: xlabel='Has Online delivery', ylabel='Votes'>
```



In this notebook, we've delved into Exploratory Data Analysis (EDA), aiming to thoroughly understand our dataset's different features, whether numerical or categorical. To achieve this, we've employed a variety of visualization techniques like count plots, histograms, box plots, and scatter plots, among others.

EDA serves as a vital initial step in comprehending the inherent patterns and connections within our data. Through visualization, we're able to glean significant insights, facilitating informed decision-making. It enables us to pinpoint trends, anomalies, correlations, and other critical aspects of our dataset, laying the groundwork for further analysis and interpretation.

```
In [ ]:
```

