

```
In [5]: # Importing Libraries
```

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

```
In [6]: df=pd.read_csv("zomato.csv",encoding='latin-1')
df.head()
```

```
Out[6]:
```

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Locality Verbose	Longitude
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...	121.0275
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...	121.0141
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...	121.0568
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...	121.0564
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...	121.0575

5 rows × 21 columns

```
In [7]: df.columns
```

```
Out[7]: Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address',
             'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines',
             'Average Cost for two', 'Currency', 'Has Table booking',
             'Has Online delivery', 'Is delivering now', 'Switch to order menu',
             'Price range', 'Aggregate rating', 'Rating color', 'Rating text',
             'Votes'],
            dtype='object')
```

```
In [8]: 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
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
```

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

```
Out[9]:
```

	Restaurant ID	Country Code	Longitude	Latitude	Average Cost for two	Price range	Aggregate rating
count	9.551000e+03	9551.000000	9551.000000	9551.000000	9551.000000	9551.000000	9551.000000
mean	9.051128e+06	18.365616	64.126574	25.854381	1199.210763	1.804837	2.6663
std	8.791521e+06	56.750546	41.467058	11.007935	16121.183073	0.905609	1.5163
min	5.300000e+01	1.000000	-157.948486	-41.330428	0.000000	1.000000	0.0000
25%	3.019625e+05	1.000000	77.081343	28.478713	250.000000	1.000000	2.5000
50%	6.004089e+06	1.000000	77.191964	28.570469	400.000000	2.000000	3.2000
75%	1.835229e+07	1.000000	77.282006	28.642758	700.000000	2.000000	3.7000
max	1.850065e+07	216.000000	174.832089	55.976980	800000.000000	4.000000	4.9000

In data analysis all Things we do

- 1.missing values
- 2.explore about the numerical variable
- 3.explore about categorical variable
- 4.finding relationship between features

```
In [10]: df.isnull().sum()
```

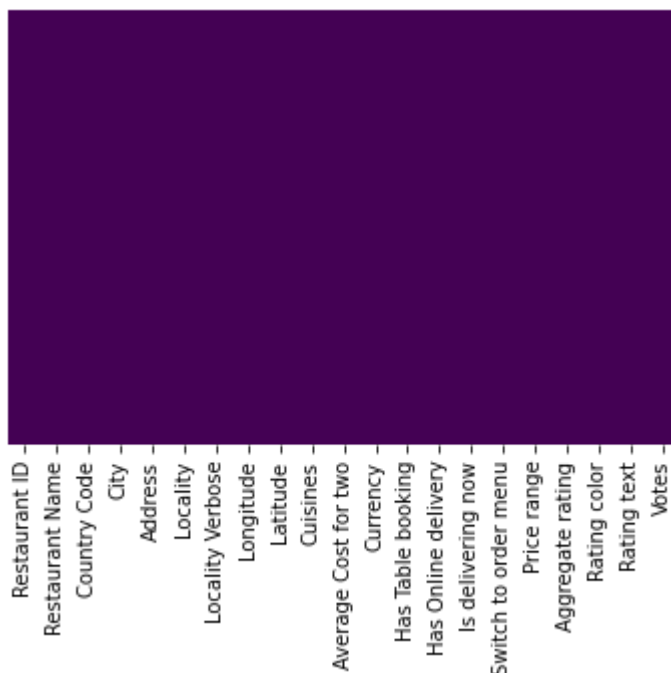
```
Out[10]: Restaurant ID      0
Restaurant Name      0
Country Code      0
City      0
Address      0
Locality      0
Locality Verbose      0
Longitude      0
Latitude      0
Cuisines      9
Average Cost for two      0
Currency      0
Has Table booking      0
Has Online delivery      0
Is delivering now      0
Switch to order menu      0
Price range      0
Aggregate rating      0
Rating color      0
Rating text      0
Votes      0
dtype: int64
```

```
In [11]: [features for features in df.columns if df[features].isnull().sum()>0]
```

```
Out[11]: ['Cuisines']
```

```
In [12]: sns.heatmap(df.isnull(),yticklabels=False,cbar=False,cmap='viridis')
```

```
Out[12]: <AxesSubplot: >
```



```
In [13]: df_country=pd.read_excel('Country-Code.xlsx')
df_country.head()
```

Out[13]:

	Country Code	Country
0	1	India
1	14	Australia
2	30	Brazil
3	37	Canada
4	94	Indonesia

	Country Code	Country
0	1	India
1	14	Australia
2	30	Brazil
3	37	Canada
4	94	Indonesia

In [14]: `df.columns`

Out[14]: Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address', 'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines', 'Average Cost for two', 'Currency', 'Has Table booking', 'Has Online delivery', 'Is delivering now', 'Switch to order menu', 'Price range', 'Aggregate rating', 'Rating color', 'Rating text', 'Votes'], dtype='object')

In []:

In [15]: `final_df=pd.merge(df,df_country,on='Country Code',how='left')`

In [16]: `final_df.head(2)`

Out[16]:

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Locality Verbose	Longitude	Latitude
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
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

2 rows × 22 columns

In [17]: `## To check data types
final_df.dtypes`

```
Out[17]: 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
```

```
In [18]:
```

```
—
```

```
Out[18]: 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
```

```
In [19]: final_df.columns
```

```
Out[19]: Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address',
               'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines',
               'Average Cost for two', 'Currency', 'Has Table booking',
               'Has Online delivery', 'Is delivering now', 'Switch to order menu',
               'Price range', 'Aggregate rating', 'Rating color', 'Rating text',
               'Votes', 'Country'],
              dtype='object')
```

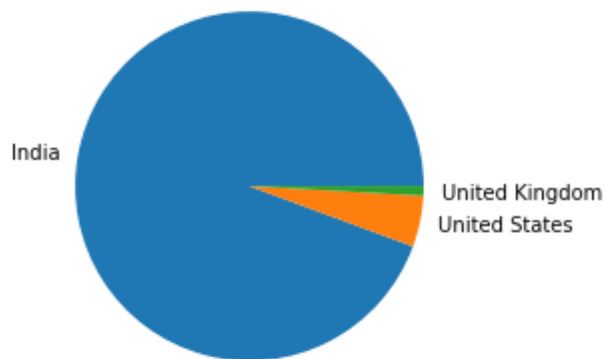
```
In [20]: country_names=final_df.Country.value_counts().index
```

```
In [21]: country_val=final_df.Country.value_counts().values
```

```
In [ ]:
```

```
In [22]: ## pie chart_ Top 3 countries that uses Zomato  
plt.pie(country_val[:3],labels=country_names[:3])
```

```
Out[22]: ([<matplotlib.patches.Wedge at 0x7f30f902b7f0>,  
          <matplotlib.patches.Wedge at 0x7f30f902bc70>,  
          <matplotlib.patches.Wedge at 0x7f30f8e64130>],  
          [Text(-1.0829742700952103, 0.19278674827836725, 'India'),  
          Text(1.077281715838356, -0.22240527134123297, 'United States'),  
          Text(1.0995865153823035, -0.03015783794312073, 'United Kingdom')])
```



```
In [23]: ## Numrical Variable
```

```
In [24]: final_df.columns
```

```
Out[24]: Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address',  
               'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines',  
               'Average Cost for two', 'Currency', 'Has Table booking',  
               'Has Online delivery', 'Is delivering now', 'Switch to order menu',  
               'Price range', 'Aggregate rating', 'Rating color', 'Rating text',  
               'Votes', 'Country'],  
              dtype='object')
```

```
In [25]: ratings=final_df.groupby(['Aggregate rating','Rating color','Rating text']).size().
```

```
In [26]: ratings
```

Out[26]:

	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	519
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

Observation

1.when Rating is between 4.5 to 4.9 -->Excellent 2.when Rating is between 4.0 to 3.4 -->very good 3.when Rating is between 3.5 to 3.9 --> good 4.when Rating is between 3.0 to 3.4 -->average 5.when rating is between 3.0 to 3.4 -->average 6.when rating is between 2.5.0 to 2.9-->average 5.when rating is between 2.0 to 2.4 -->poor

In [27]: `ratings.head()`

Out[27]:

	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

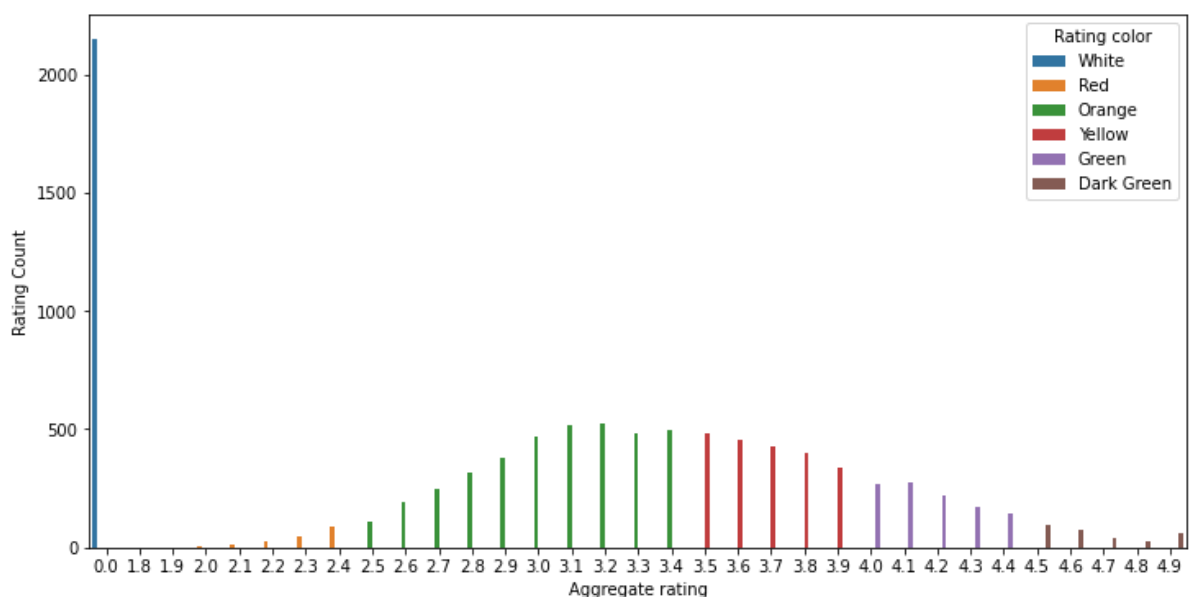
In [28]: `ratings.head()`

Out[28]:

	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

In [29]: `import matplotlib`
`matplotlib.rcParams['figure.figsize']=(12, 6)`
`sns.barplot(x="Aggregate rating",y="Rating Count",hue='Rating color',data=ratings)`

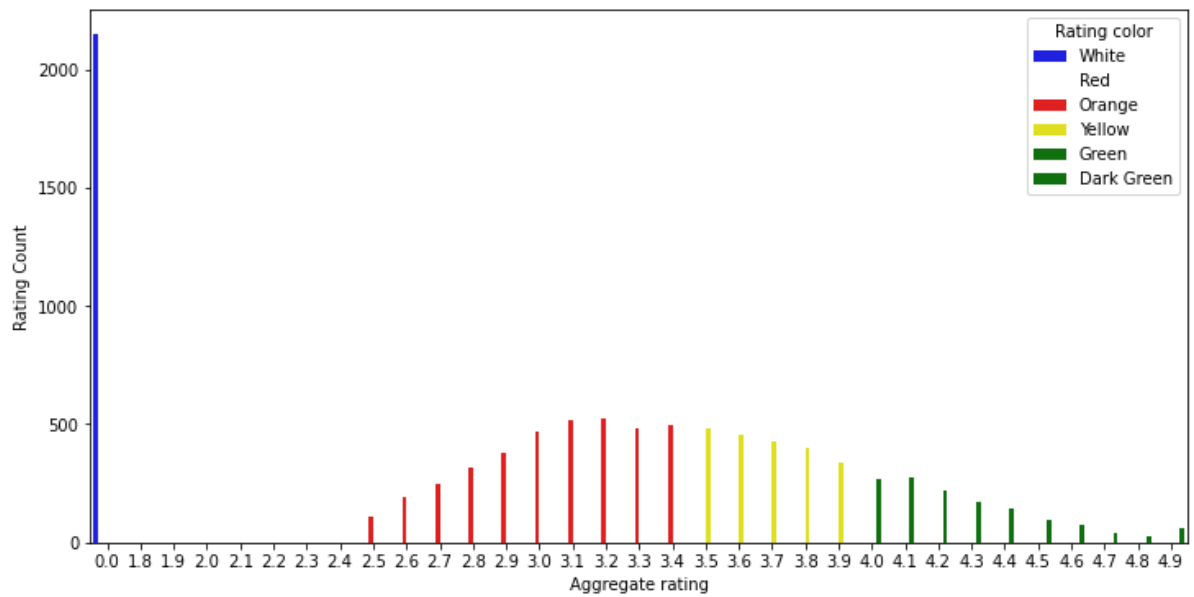
Out[29]: `<AxesSubplot: xlabel='Aggregate rating', ylabel='Rating Count'>`



Obervation 1.Not Rated count is very high 2.Maximum number of rating are between 2.5 to 3.4


```
In [33]: # bar plot
sns.barplot(x="Aggregate rating",y="Rating Count",hue='Rating color',data=ratings,
```

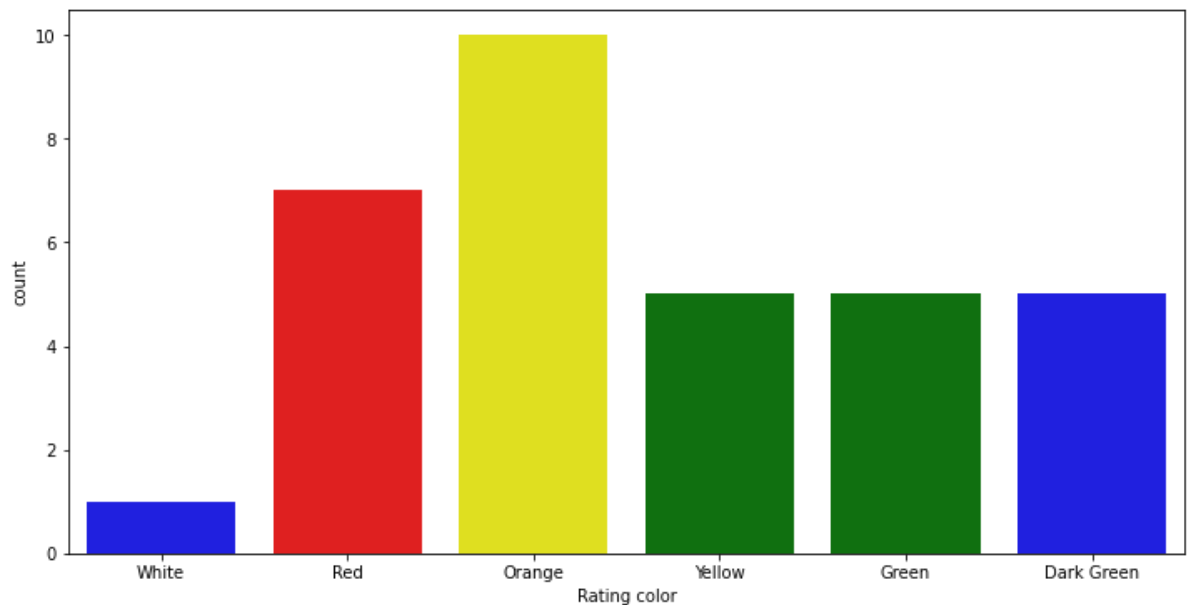
```
Out[33]: <AxesSubplot: xlabel='Aggregate rating', ylabel='Rating Count'>
```



Obervation : 1.Not rated count is very high 2.maximum numbers ofv rating are between 2.5 to 3.4

```
In [38]: ## count plot
sns.countplot(x="Rating color",data=ratings,palette= ['blue','red','yellow','green']
```

```
Out[38]: <AxesSubplot: xlabel='Rating color', ylabel='count'>
```



```
In [39]: ratings
```

Out[39]:

	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	519
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 [44]: `## find the countries name that has given o rating
final_df[final_df['Rating color']=='White'].groupby('Country').size().reset_index()`

```
Out[44]:
```

	Country	0
0	Brazil	5
1	India	2139
2	United Kingdom	1
3	United States	3

```
In [52]: final_df.groupby(['Aggregate rating', 'Country']).size().reset_index().head(5)
```

```
Out[52]:
```

	Aggregate rating	Country	0
0	0.0	Brazil	5
1	0.0	India	2139
2	0.0	United Kingdom	1
3	0.0	United States	3
4	1.8	India	1

Obervation 1.Maximum number of 0 ratings are from Indian customer

```
In [53]: final_df.columns
```

```
Out[53]: Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address',
                'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines',
                'Average Cost for two', 'Currency', 'Has Table booking',
                'Has Online delivery', 'Is delivering now', 'Switch to order menu',
                'Price range', 'Aggregate rating', 'Rating color', 'Rating text',
                'Votes', 'Country'],
                dtype='object')
```

```
In [56]: final_df[['Country', 'Currency']].groupby(['Country', 'Currency']).size().reset_index
```

Out[56]:

	Country	Currency	0
0	Australia	Dollar(\$)	24
1	Brazil	Brazilian Real(R\$)	60
2	Canada	Dollar(\$)	4
3	India	Indian Rupees(Rs.)	8652
4	Indonesia	Indonesian Rupiah(IDR)	21
5	New Zealand	NewZealand(\$)	40
6	Phillipines	Botswana Pula(P)	22
7	Qatar	Qatari Rial(QR)	20
8	Singapore	Dollar(\$)	20
9	South Africa	Rand(R)	60
10	Sri Lanka	Sri Lankan Rupee(LKR)	20
11	Turkey	Turkish Lira(TL)	34
12	UAE	Emirati Diram(AED)	60
13	United Kingdom	Pounds(£)	80
14	United States	Dollar(\$)	434

In [61]: `## Which countries do have online deliveries option`
`final_df.groupby(['Has Online delivery']).size().reset_index()`

Out[61]:

	Has Online delivery	0
0	No	7100
1	Yes	2451

In [67]: `final_df[final_df['Has Online delivery'] == "Yes"].Country.value_counts()`

Out[67]: India 2423
UAE 28
Name: Country, dtype: int64

In [74]: `final_df[['Has Online delivery','Country']].groupby(['Has Online delivery','Country`

```
Out[74]:
```

	Has Online delivery	Country	0
0	No	Australia	24
1	No	Brazil	60
2	No	Canada	4
3	No	India	6229
4	No	Indonesia	21
5	No	New Zealand	40
6	No	Phillipines	22
7	No	Qatar	20
8	No	Singapore	20
9	No	South Africa	60
10	No	Sri Lanka	20
11	No	Turkey	34
12	No	UAE	32
13	No	United Kingdom	80
14	No	United States	434
15	Yes	India	2423
16	Yes	UAE	28

```
In [ ]:
```

Obervations: 1.Online delivaries are avaiiible in india and UAE

```
In [75]: final_df.columns
```

```
Out[75]: Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address',
               'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines',
               'Average Cost for two', 'Currency', 'Has Table booking',
               'Has Online delivery', 'Is delivering now', 'Switch to order menu',
               'Price range', 'Aggregate rating', 'Rating color', 'Rating text',
               'Votes', 'Country'],
              dtype='object')
```

```
In [ ]: ## Create a pie Chart a specific top 5 cities distibution
```

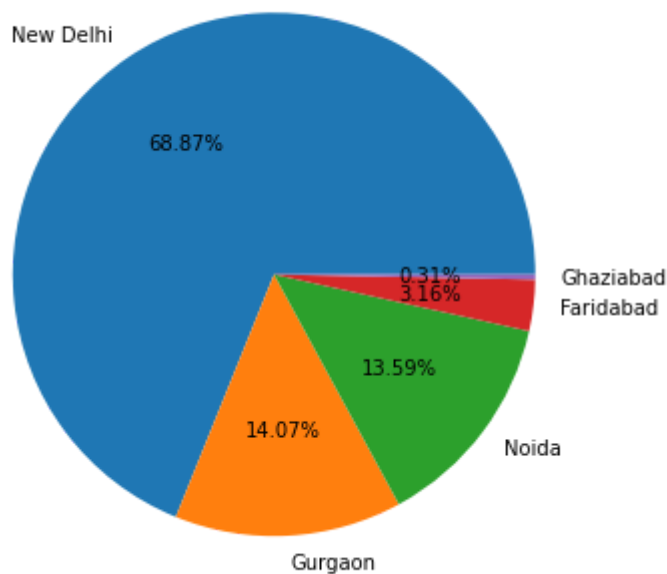
```
In [77]: final_df.City.value_counts().index
```

```
Out[77]: Index(['New Delhi', 'Gurgaon', 'Noida', 'Faridabad', 'Ghaziabad',
               'Bhubaneshwar', 'Amritsar', 'Ahmedabad', 'Lucknow', 'Guwahati',
               ...,
               'Ojo Caliente', 'Montville', 'Monroe', 'Miller', 'Middleton Beach',
               'Panchkula', 'Mc Millan', 'Mayfield', 'Macedon', 'Vineland Station'],
              dtype='object', length=141)
```

```
In [84]: city_values=final_df.City.value_counts().values
         city_labels=final_df.City.value_counts().index
```

```
In [87]: plt.pie(city_values[:5],labels=city_labels[:5],autopct='%1.2f%%')
```

```
Out[87]: ([<matplotlib.patches.Wedge at 0x7f30f43f9390>,
<matplotlib.patches.Wedge at 0x7f30f43f9a20>,
<matplotlib.patches.Wedge at 0x7f30f43fa0b0>,
<matplotlib.patches.Wedge at 0x7f30f43fa740>,
<matplotlib.patches.Wedge at 0x7f30f43fadd0>],
[Text(-0.6145352824185932, 0.9123301960708633, 'New Delhi'),
Text(0.0623675251198054, -1.0982305276263407, 'Gurgaon'),
Text(0.8789045225625368, -0.6614581167535246, 'Noida'),
Text(1.0922218418223437, -0.13058119407559224, 'Faridabad'),
Text(1.099946280005612, -0.010871113182029924, 'Ghaziabad')],
[Text(-0.3352010631374145, 0.497634652402289, '68.87%'),
Text(0.0340186500653484, -0.5990348332507311, '14.07%'),
Text(0.47940246685229276, -0.36079533641101336, '13.59%'),
Text(0.5957573682667329, -0.07122610585941394, '3.16%'),
Text(0.5999706981848791, -0.005929698099289049, '0.31%')])
```



```
In [ ]:
```