

Exploratory data analysis

Zomato Data Set

In [2]:

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

In [9]:

```
df=pd.read_csv("zomato.csv",encoding='latin-1')
df.head()
```

Out[9]:

| | 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 Avenue... | Century City Mall, Poblacion, Makati City | Century City Mall, Poblacion, Makati City, Mak... | 121.027535 | 14.650000 |
| 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 | 14.650000 |
| 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 | 14.650000 |
| 3 | 6318506 | Ooma | 162 | Mandaluyong City | Third Floor, Mega Fashion Hall, SM Megamall, O... | SM Megamall, Ortigas, Mandaluyong City | SM Megamall, Ortigas, Mandaluyong City, Mand... | 121.056475 | 14.650000 |
| 4 | 6314302 | Sambo Kojin | 162 | Mandaluyong City | Third Floor, Mega Atrium, SM | SM Megamall, Ortigas, Mandaluyong City | SM Megamall, Ortigas, Mandaluyong | 121.057508 | 14.650000 |

| Restaurant ID | Restaurant Name | Country Code | City | Address | Locality | Locality Verbose | Longitude | Latitude | 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 |
|---------------|-----------------|--------------|------|-------------------------|----------|------------------|-----------|----------|----------------------|----------|-------------------|---------------------|-------------------|----------------------|-------------|------------------|--------------|-------------|--------------------|
| | | | | Megamall, Ortigas... | | | | | | | | | | | | | | | City, Mandal... |

5 rows × 21 columns

In [10]: `df.columns`

Out[10]: 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 [11]: `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 [12]: `df.describe()`

Out[12]:

| | 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 |

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

Data Analysis

missing value

Explore numerical and categorical values

Relationship between features

In [20]:

df.shape

Out[20]: (9551, 21)

In [14]:

df.isnull().sum()

```
Out[14]: 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
```

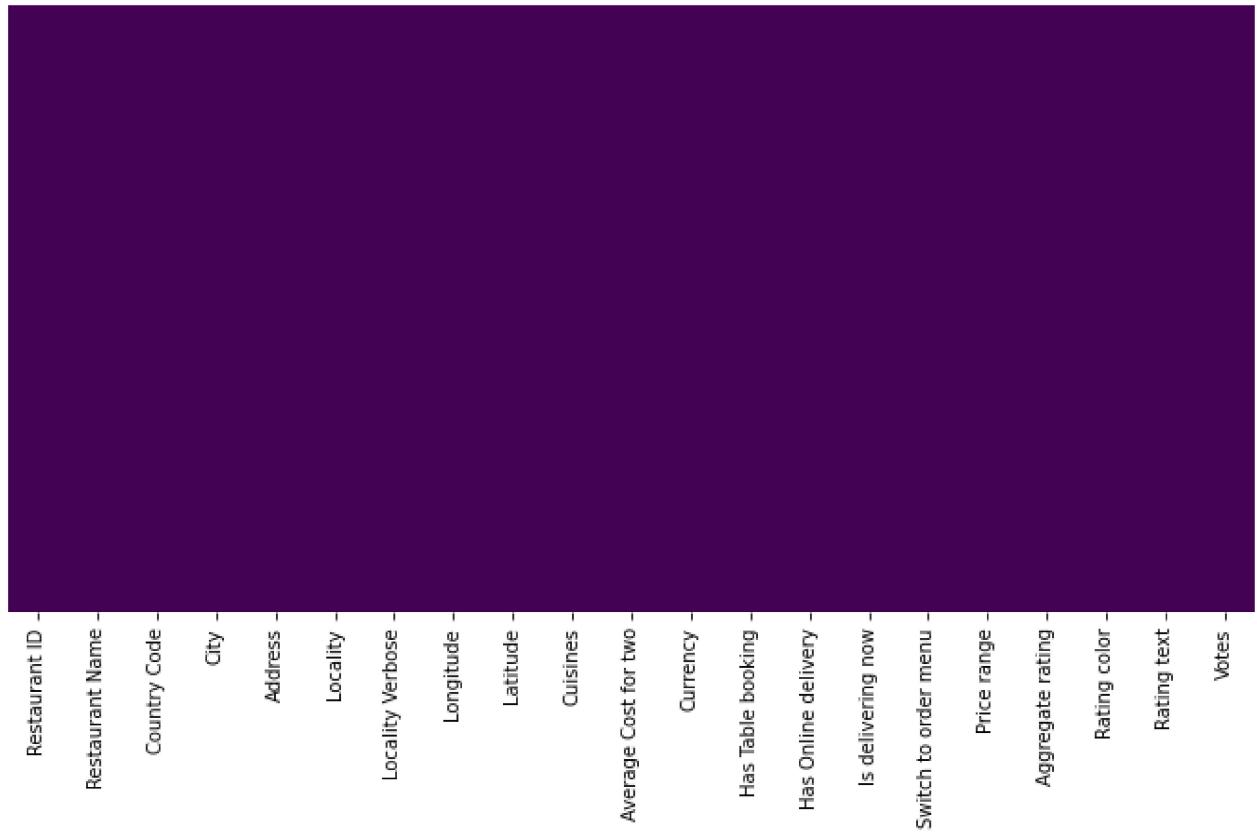
```
Votes          0
dtype: int64
```

```
In [18]: [features for features in df.columns if df[features].isnull().sum()>0]
#List comprehension
```

```
Out[18]: ['Cuisines']
```

```
In [73]: #Plotting for Heat Map
sns.heatmap(df.isnull(), yticklabels=False, cbar=False, cmap='viridis')
```

```
Out[73]: <AxesSubplot:>
```



```
In [24]: df_country=pd.read_excel('Country-code.xlsx')
df_country.head()
```

```
Out[24]:
```

| | 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 [25]: df.columns
```

```
Out[25]: 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')
```

Combining datasets zomato.csv and Country-code.xlsx because they have a common column named Country Code. joining both data sets using 'merge' and putting it in df_final dataset.

```
In [27]: df_final=pd.merge(df,df_country, on='Country Code', how='left')
```

```
In [29]: df_final.head(2)
```

Out[29]:

| | Restaurant ID | Restaurant Name | Country Code | City | Address | Locality | Locality Verbose | Longitude | Latitude | Cuis |
|---|---------------|------------------|--------------|-------------|--|--|---|------------|-----------|---------|
| 0 | 6317637 | Le Petit Souffle | 162 | Makati City | Third Floor, Century City Mall, Kalayaan Avenue... | Century City Mall, Poblacion, Makati City | Century City Mall, Poblacion, Makati City, Mak... | 121.027535 | 14.565443 | Fre Des |
| 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 | Japa |

2 rows × 22 columns

```
In [30]: # To check data types
df_final.dtypes
```

```
Out[30]: 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 [31]: df_final.columns

Out[31]: 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 [36]: df_final.Country.value_counts()

Out[36]:

| Country | Count |
|----------------|-------|
| India | 8652 |
| United States | 434 |
| United Kingdom | 80 |
| Brazil | 60 |
| South Africa | 60 |
| UAE | 60 |
| New Zealand | 40 |
| Turkey | 34 |
| Australia | 24 |
| Phillipines | 22 |
| Indonesia | 21 |
| Sri Lanka | 20 |
| Singapore | 20 |
| Qatar | 20 |
| Canada | 4 |

Name: Country, dtype: int64

In [37]: Country_names=df_final.Country.value_counts().index #y-axis for pie chart

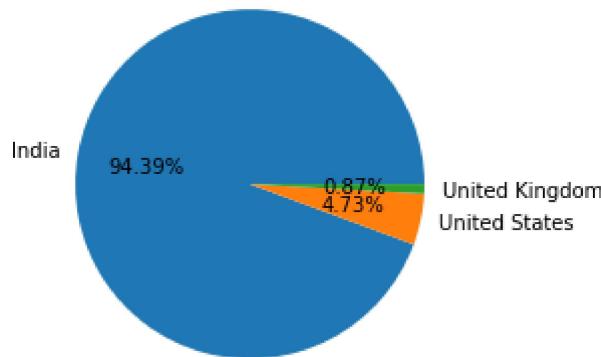
In [40]: Country_val=df_final.Country.value_counts().values #x-axis for pie chart

In [51]:

```
# Pie Chart
plt.pie(Country_val[:3], labels=Country_names[:3], autopct='%1.2f%%') #Top 3 countries th
```

Out[51]: ([<matplotlib.patches.Wedge at 0x1778d295b80>,
 <matplotlib.patches.Wedge at 0x1778d2a3340>,
 <matplotlib.patches.Wedge at 0x1778d2a3a00>],
 [Text(-1.0829742700952103, 0.19278674827836725, 'India'),
 Text(1.077281715838356, -0.22240527134123297, 'United States'),

```
Text(1.0995865153823035, -0.03015783794312073, 'United Kingdom')],  
[Text(-0.590713238233751, 0.10515640815183668, '94.39%'),  
Text(0.5876082086391032, -0.12131196618612707, '4.73%'),  
Text(0.5997744629358018, -0.01644972978715676, '0.87%')])
```



Observation: The country with maximum % records of transactions are from India, then United States and 3rd is United Kingdom.

In [53]: `df_final.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 [65]: `ratings=df_final.groupby(['Aggregate rating', 'Rating color', 'Rating text']).size().re
#to change it to a data frame
#and doing indexing and where column is 0 change it to Rating Count.`

In [66]: `ratings`

| | 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 |

| | Aggregate rating | Rating color | Rating text | Rating Count |
|----|------------------|--------------|-------------|--------------|
| 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 4.4....> very good
3. When rating is between 3.5 to 3.9....> good
4. When rating is between 2.5 to 3.4....> average
5. When rating is between 1.8 to 2.4.....> poor

In [67]:

```
ratings.head()
```

Out[67]:

| | Aggregate rating | Rating color | Rating text | Rating Count |
|---|------------------|--------------|-------------|--------------|
| 0 | 0.0 | White | Not rated | 2148 |
| 1 | 1.8 | Red | Poor | 1 |

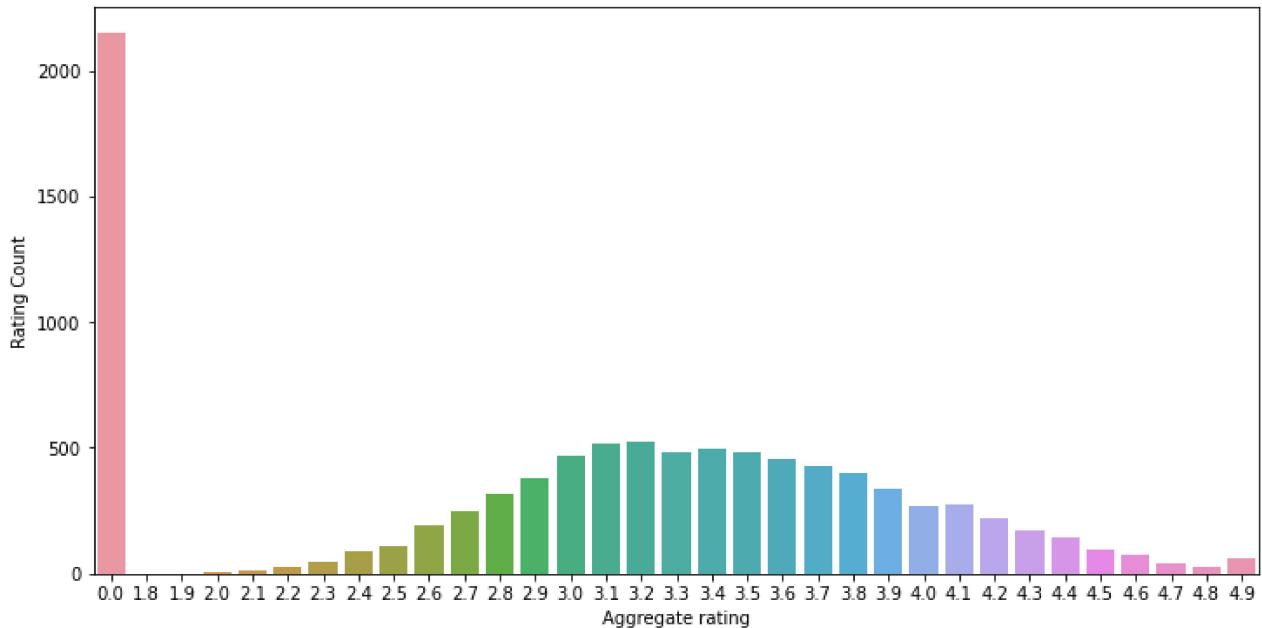
| Aggregate rating | Rating color | Rating text | Rating Count |
|------------------|--------------|-------------|--------------|
|------------------|--------------|-------------|--------------|

| | | | | |
|---|-----|-----|------|----|
| 2 | 1.9 | Red | Poor | 2 |
| 3 | 2.0 | Red | Poor | 7 |
| 4 | 2.1 | Red | Poor | 15 |

In [75]:

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

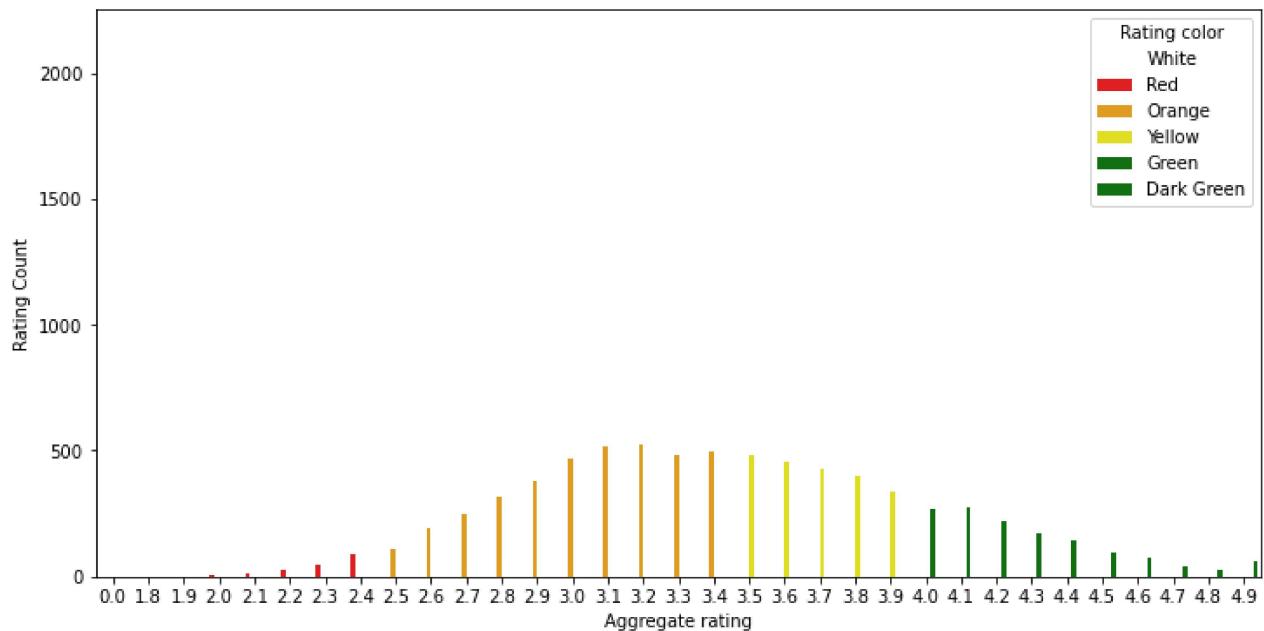
Out[75]: <AxesSubplot:xlabel='Aggregate rating', ylabel='Rating Count'>



In [79]:

```
# Mapping color according to the rating
sns.barplot(x='Aggregate rating', y='Rating Count', hue='Rating color', data=ratings, pal
```

Out[79]: <AxesSubplot:xlabel='Aggregate rating', ylabel='Rating Count'>



In []:

Observation

1. Not rated count **is** very high
2. Maximum rating **is** between **2.5** to **3.4**

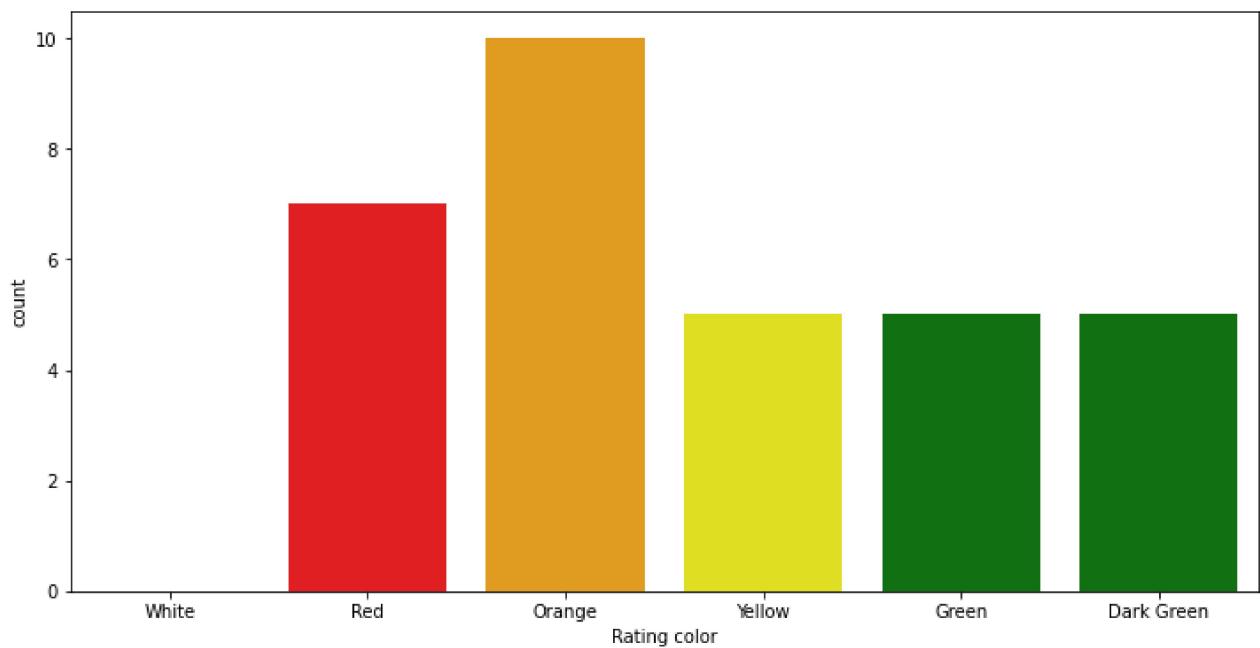
In [140...]

Count plot

sns.countplot(x='Rating color', data=ratings, palette=['white', 'red', 'orange', 'yellow', 'green', 'darkgreen'])

Out[140...]

<AxesSubplot:xlabel='Rating color', ylabel='count'>



In [81]:

ratings

Out[81]:

Aggregate rating Rating color Rating text Rating Count

| Aggregate rating | Rating color | Rating text | Rating Count |
|------------------|--------------|-------------|--------------|
| 0 | White | Not rated | 2148 |

| | Aggregate rating | Rating color | Rating text | Rating Count |
|----|------------------|--------------|-------------|--------------|
| 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 []:

Countries that has given 0 ratings

```
In [85]: df_final[df_final['Rating color']=='White'].groupby('Country').size().reset_index()
```

Out[85]:

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

```
In [91]: df_final.groupby(['Aggregate rating', 'Country']).size().reset_index().head(5)
```

Out[91]:

| | 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 |

Observation Maximum number of 0 ratings is from India

```
In [93]: # Which currency is used by which country?
df_final.columns
```

Out[93]: 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 [96]: df_final.groupby(['Country', 'Currency']).size().reset_index()
```

Out[96]:

| | 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 |

| | Country | Currency | 0 |
|----|----------------|-----------------------|-----|
| 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 [101...]

```
# Another method to see the country and currency is:
df_final[['Country', 'Currency']].groupby(['Country', 'Currency']).size().reset_index()
```

Out[101...]

| | 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 [109...]

```
# Which Country do have online delivery options
df_final[df_final['Has Online delivery']=='Yes'].Country.value_counts()
```

Out[109...]

```
India    2423
UAE     28
Name: Country, dtype: int64
```

In [110...]

```
# Another method to see the countries which has online delivery
df_final.groupby(['Has Online delivery', 'Country']).size().reset_index()
```

Out[110...]

| | 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 |

Observation

1. Online deliveries are available in India and UAE

In [112...]

```
# Create a pie chart for cities distribution
df_final.columns
```

Out[112...]

```
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 [115...]

```
df_final.City.value_counts().reset_index()
```

Out[115...]

| | index | City |
|---|-----------|------|
| 0 | New Delhi | 5473 |
| 1 | Gurgaon | 1118 |

| index | City | |
|-------|---------------|------|
| 2 | Noida | 1080 |
| 3 | Faridabad | 251 |
| 4 | Ghaziabad | 25 |
| ... | ... | ... |
| 136 | Yorkton | 1 |
| 137 | Paynesville | 1 |
| 138 | Victor Harbor | 1 |
| 139 | Armidale | 1 |
| 140 | Montville | 1 |

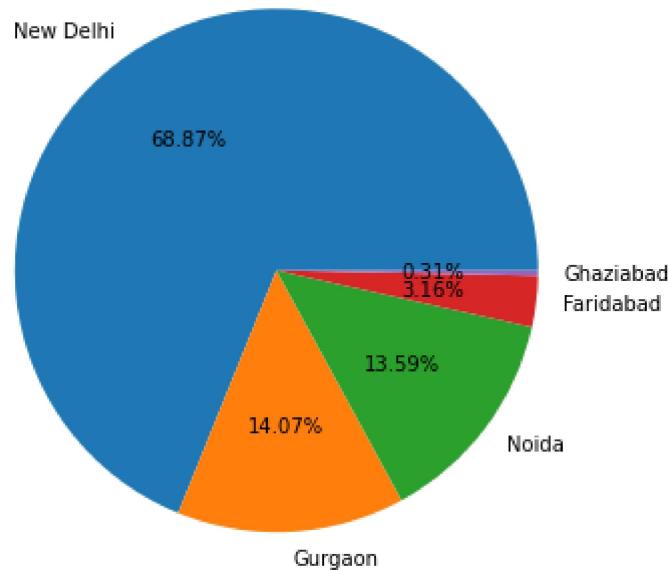
141 rows × 2 columns

```
In [122... City_labels=df_final.City.value_counts().index
```

```
In [123... City_val=df_final.City.value_counts().values
```

```
In [127... # Top 5 City distribution
plt.pie(City_val[:5],labels=City_labels[:5],autopct='%1.2f%')
```

```
Out[127... ([<matplotlib.patches.Wedge at 0x17791ef5dc0>,
 <matplotlib.patches.Wedge at 0x17791f044f0>,
 <matplotlib.patches.Wedge at 0x17791f04af0>,
 <matplotlib.patches.Wedge at 0x17791f13130>,
 <matplotlib.patches.Wedge at 0x17791f13730>],
 [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 [128...]

```
# Find Top 10 cuisines
df_final.columns
```

Out[128...]

```
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 [139...]

```
df_final.Cuisines.value_counts().reset_index().head(10)
```

Out[139...]

| | index | Cuisines |
|----------|--------------------------------|----------|
| 0 | North Indian | 936 |
| 1 | North Indian, Chinese | 511 |
| 2 | Chinese | 354 |
| 3 | Fast Food | 354 |
| 4 | North Indian, Mughlai | 334 |
| 5 | Cafe | 299 |
| 6 | Bakery | 218 |
| 7 | North Indian, Mughlai, Chinese | 197 |
| 8 | Bakery, Desserts | 170 |
| 9 | Street Food | 149 |

In []: