

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import matplotlib inline

In [2]: df=pd.read_csv('zomato.csv',encoding='latin-1')
df.head()
```

Out[2]:

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Locality Verbose	Longitude	Latitude	Cuisines	...	Currency	Has Table booking	Has Online delivery	Is delivering now	Switch to order menu	Price range	Aggregate rating	Rating color	Rating text	Votes
0	6317637	Le Petit South	162	Makati City	Third Floor, Century City Mall, Avenue...	Century City	Century City Mall, Makati City, Ma...	121.027335	14.565443	French, Japanese, Desserts	...	Botswana Pula(P)	Yes	No	No	No	3	4.8	Dark Green	Excellent	314
1	6304287	Isakaya Kikuyu	162	Makati City	Unit 277, Chico Village, Legaspi Ave...	Little Tokyo	Legaspi Village, Makati City, Ma...	121.014101	14.553708	Japanese	...	Botswana Pula(P)	Yes	No	No	No	3	4.5	Dark Green	Excellent	591
2	6300002	Heat - Edsa Shangri-La	162	Mandaluyong City	Shangri-La, Edsa Shangri-La, Mandaluyong City	Edsa Shangri-La	Edsa Shangri-La, Mandaluyong City, Ma...	121.056831	14.581404	Seafood, Filipino, Indian	...	Botswana Pula(P)	Yes	No	No	No	4	4.4	Green	Very Good	270
3	6318906	Ooma	162	Mandaluyong City	Third Floor, Megamall, Fashion Hub, SM Megamall, Mandaluyong City	SM Megamall	SM Megamall, Mandaluyong City, Ma...	121.056475	14.589318	Japanese, Sushi	...	Botswana Pula(P)	No	No	No	No	4	4.9	Dark Green	Excellent	365
4	6314302	Sambo Ragni	162	Mandaluyong City	Third Floor, Megamall, Arirang, SM Megamall, Mandaluyong City	SM Megamall	SM Megamall, Mandaluyong City, Ma...	121.057908	14.598448	Japanese, Korean	...	Botswana Pula(P)	Yes	No	No	No	4	4.8	Dark Green	Excellent	229

5 rows x 21 columns

In [3]:

df.columns

Out[3]:

Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address', 'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines', '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 [4]:

df.info()

Out[4]:

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9551 entries, 0 to 9550
Data columns (total 21 columns):
Column Dtype
-- --
0 Restaurant ID non-null: 9551, dtype: int64
1 Restaurant Name non-null: 9551, dtype: object
2 Country Code non-null: 9551, dtype: int64
3 City non-null: 9551, dtype: object
4 Address non-null: 9551, dtype: object
5 Locality non-null: 9551, dtype: object
6 Locality Verbose non-null: 9551, dtype: object
7 Longitude non-null: 9551, dtype: float64
8 Latitude non-null: 9551, dtype: float64
9 Cuisines non-null: 9551, dtype: object
10 Average Cost for two non-null: 9551, dtype: object
11 Currency non-null: 9551, dtype: object
12 Has Table booking non-null: 9551, dtype: object
13 Has Online delivery non-null: 9551, dtype: object
14 Is delivering now non-null: 9551, dtype: object
15 Switch to order menu non-null: 9551, dtype: object
16 Price range non-null: 9551, dtype: int64
17 Aggregate rating non-null: 9551, dtype: float64
18 Rating color non-null: 9551, dtype: object
19 Rating text non-null: 9551, dtype: object
20 Votes non-null: 9551, dtype: int64
dtypes: float64(3), int64(5), object(13)
memory usage: 1.5+ MB

In [5]:

df.describe()

Out[5]:

	Restaurant ID	Country Code	Longitude	Latitude	Average Cost for two	Price range	Aggregate rating	Votes
count	9.551000e+03	9551.000000	9551.000000	9551.000000	9551.000000	9551.000000	9551.000000	9551.000000
mean	9.051129e+06	18.385616	64.126574	25.854381	1199.210763	1.884837	2.666370	156.909748
std	8.791522e+06	56.750546	41.467058	11.007935	16121.183073	0.905609	1.516378	430.169145
min	5.300000e+01	1.000000	-157.948486	-41.330428	0.000000	1.000000	2.500000	0.000000
25%	3.016629e+05	1.000000	77.081343	28.478713	250.000000	1.000000	2.500000	0.000000
50%	6.004089e+06	1.000000	77.191964	28.570489	400.000000	2.000000	3.200000	31.000000
75%	1.000000e+07	1.000000	78.000000	28.607750	700.000000	2.000000	3.700000	151.000000
max	1.850059e+07	216.000000	174.832089	55.976980	800000.000000	4.000000	4.900000	10934.000000

In [6]:

check for missing values

In [7]:

df.isnull().sum()

Out[7]:

Restaurant ID 0
Restaurant Name 0
Country Code 0
City 0
Address 0
Locality 0
Locality Verbose 0
Longitude 0
Latitude 0
Cuisines 0
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 [8]:

[features for features in df.columns if df[features].isnull().sum()==0]

Out[8]:

['Cuisines']

In [9]:

sns.heatmap(df.isnull(),yticklabels=False,cbar=False,cmap='viridis')

Out[9]:

<AxesSubplot>

In [10]:

df.shape

Out[10]:

(9551, 21)

In [11]:

df.country=pd.read_excel('Country-Code.xlsx')
df.country.head()

Out[11]:

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

In [12]:

df.columns

Out[12]:

Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address', 'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines', '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 [13]:

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

In [14]:

final_df.head(2)

Out[14]:

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Locality Verbose	Longitude	Latitude	Cuisines	...	Has Table booking	Has Online delivery	Is delivering now	Switch to order menu	Price range	Aggregate rating	Rating color	Rating text	Votes	Country
0	6317637	Le Petit South	162	Makati City	Third Floor, Century City Mall, Avenue...	Century City	Century City Mall, Makati City, Ma...	121.027335	14.565443	French, Japanese, Desserts	...	Yes	No	No	No	3	4.8	Dark Green	Excellent	314	Philippines
1	6304287	Isakaya Kikuyu	162	Makati City	Unit 277, Chico Village, Legaspi Ave...	Little Tokyo	Legaspi Village, Makati City, Ma...	121.014101	14.553708	Japanese	...	Yes	No	No	No	3	4.5	Dark Green	Excellent	591	Philippines

2 rows x 22 columns

In [15]:

#checking data types
final_df.dtypes

Out[15]:

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 int64
Votes object
dtype: object

In [16]:

final_df.columns

Out[16]:

Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address', 'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines', '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 [18]:

final_df.Country.value_counts()

Out[18]:

Country	count
India	8652
United States	424
United Kingdom	88
Brazil	68
UAE	68
South Africa	68
New Zealand	48
Turkey	34
Australia	24
Philippines	22
Indonesia	21
Singapore	20
Qatar	28
Sri Lanka	28
Canada	4

Name: Country, dtype: int64

In [33]:

country_names=final_df.Country.value_counts().index

In [34]:

country_val=final_df.Country.value_counts().values

In [38]:

#to see which country has maximum transactions

In [35]:

#pie chart
plt.pie(country_val,labels=country_names)

Out[35]:

([<matplotlib.patches.Wedge at 8x1bdcf5329a8>,
<matplotlib.patches.Wedge at 8x1bdcf532468>,
<matplotlib.patches.Wedge at 8x1bdcf532948>,
<matplotlib.patches.Wedge at 8x1bdcf532048>,
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<matplotlib.patches.Wedge at 8x1bdcf588098>,
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<matplotlib.patches.Wedge at 8x1bdcf58a708>,
<matplotlib.patches.Wedge at 8x1bdcf58a0e8>,
<matplotlib.patches.Wedge at 8x1bdcf73248>,
<matplotlib.patches.Wedge at 8x1bdcf584558>,
<matplotlib.patches.Wedge at 8x1bdcf58438>,
<matplotlib.patches.Wedge at 8x1bdcf584738>,
<matplotlib.patches.Wedge at 8x1bdcf6d9f38>],
[Text(1.052256163793291, 0.328557275757986, 'India'),
Text(0.99132983284435, -0.4713248941823, 'United States'),
Text(1.82738859613743, -0.283526707227465, 'United Kingdom'),
Text(1.878138816916819, -0.254584163112621, 'Brazil'),
Text(1.875556854419759, -0.3121388926848624, 'UAE'),
Text(1.886881147244973, -0.15897937230799818, 'South Africa'),
Text(1.8916639511823935, -0.312349329752956, 'New Zealand'),
Text(1.894796381461446, -0.1660299887838384, 'Turkey'),
Text(1.899631023943382, -0.868625628774326, 'Australia'),
Text(1.897807787764525, -0.868423588273521, 'Philippines'),
Text(1.8980791444015289, -0.8538894768943233, 'Indonesia'),
Text(1.8995895848742366, -0.8389058265131035, 'Singapore'),
Text(1.89472498282123, -0.8246018794173628, 'Qatar'),
Text(1.8992343402179836, -0.810139485275446, 'Sri Lanka'),
Text(1.899990477553414, -0.8914473888376707639, 'Canada')])

In [36]:

Pie Chart: Top 3 countries that uses zomato
plt.pie(country_val[:3],labels=country_names[:3],autopct='%1.2f%%')

Out[36]:

([<matplotlib.patches.Wedge at 8x1bdcf1c848>,
<matplotlib.patches.Wedge at 8x1bdcf1c8948>,
<matplotlib.patches.Wedge at 8x1bdcf1c8278>],
[Text(1.082374708952185, 0.18278674827378075, 'India'),
Text(1.877281715838356, -0.2224827134123297, 'United States'),
Text(1.8995895848742366, -0.8389058265131035, 'United Kingdom'),
Text(1.886881147244973, -0.15897937230799818, 'South Africa'),
Text(1.8916639511823935, -0.312349329752956, 'New Zealand'),
Text(1.89472498282123, -0.8246018794173628, 'Qatar'),
Text(1.8992343402179836, -0.810139485275446, 'Sri Lanka'),
Text(1.899990477553414, -0.8914473888376707639, 'Canada')])

In [37]:

India 94.3%
United States 4.4%
United Kingdom 1.2%

observation:most of the transactions are done in india followed by US And UK

In [38]:

final_df.columns

Out[38]:

Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address', 'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines', '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 [39]:

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

Out[41]:

	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	105
9	2.6	Orange	Average	135
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

observations

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 3.0 to 3.4 ---->average
5.when rating is between 2.5 to 2.9 ---->average
6.when rating is between 2.0 to 2.4 ---->poor

Out [43]:

ratings.head()

Out[43]:

	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 [45]:

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

Out[45]:

<AxesSubplot>

In [49]:

sns.barplot(x='Aggregate rating',y='Rating Count',hue='Rating color',data=ratings)

Out[49]:

<AxesSubplot>

In [56]:

sns.barplot(x='Aggregate rating',y='Rating Count',hue='Rating color',data=ratings,palette=['blue','red','orange','yellow','green','green'])

Out[56]:

<AxesSubplot>

In [58]:

sns.barplot(x='Aggregate rating',y='Rating Count',hue='Rating color',data=ratings,palette=['blue','red','orange','yellow','green','green'])

Out[58]:

<AxesSubplot>

In [59]:

observation:
1.not rated count is very high
2.maximum number of ratings are between 2.5 to 3.4

In [51]:

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

Out[51]:

<AxesSubplot>

In [52]:

find the countries name that has high grouping
final_df.groupby(['Rating color']).size().reset_index()

Out[52]:

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

In [53]:

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

Out[53]:

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

In [54]:

Observations Maximum number of ratings are from indian customers
#find out which currency is used by which country
final_df.columns

Out[54]:

Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address', 'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines', '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 [55]:

final_df[['Country','Currency']].groupby(['Country','Currency']).size().reset_index()

Out[55]:

	Country	Currency	0
0	Australia	Dollar(AU)	24
1	Brazil	Brazilian Real(R\$)	60
2	Canada	Dollar(CA)	852
3	India	Indian Rupee(Rs)	4054
4	Indonesia	Indonesian Rupee(Rp)	26
5	New Zealand	New Zealand\$	40
6	Philippines	Botswana Pula(P)	22
7	Qatar	Qatari Rial(QR)	20
8	Singapore	Dollar(SG)	20
9	South Africa	Rand(R)	60
10	Sri Lanka	Sri Lankan Rupee(LKR)	20
11	Turkey	Turkish Lira(TL)	34
12	United Kingdom	English Pound(GBP)	60
13	United Kingdom	Pound(GB)	80
14	United States	Dollar(US)	434

In [56]:

#w which Countries do have online deliveries option

Out[57]:

final_df[final_df['Has Online delivery']==Yes].Country.value_counts()

Out[57]:

	Country	0
0	India	2423
1	UAE	28

Name: Country, dtype: int64

In [58]:

final_df[['Has Online delivery','Country']].groupby(['Has Online delivery','Country']).size().reset_index()

Out[58]:

	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	Philippines	22
7	No	Qatar	20
8	No	Singapore	20
9	No	South Africa	60
10	No	South Africa	60
11	No	Sri Lanka	20
12	No	Turkey	34
13	No	UAE	32
14	No	United Kingdom	80
15	No	United States	434
16	Yes	India	2423
17	Yes	UAE	28

Observations

1. Online Deliveries are available in India and UAE

In [59]:

final_df.columns

Out[59]:

Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address', 'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines', '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 [60]:

Create a pie chart for top 5 cities distribution

In [61]:

final_df.City.value_counts().index

Out[61]:

Index(['New Delhi', 'Gurgaon', 'Noida', 'Faridabad', 'Ghaziabad', 'Bhubaneswar', 'Mumbai', 'Ahmedabad', 'Lucknow', 'Gwalhati', 'Panchajitra', 'Mc Millan', 'Mayfield', 'Macedon', 'Vineland Station'], dtype='object', length=14)

In [62]:

city_values=final_df.City.value_counts().values
city_labels=final_df.City.value_counts().index

Out[62]:

City	count
New Delhi	48.67%
Gurgaon	23.19%
Noida	13.31%
Faridabad	6.87%
Ghaziabad	6.87%

In []: