

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

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

```
In [3]: df.head()
```

Out[3]:

	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 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	No	No	No	3	4.8	Dark Green	Excellent	314
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)	Yes	No	No	No	3	4.5	Dark Green	Excellent	591
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	...	Botswana Pula(P)	Yes	No	No	No	4	4.4	Green	Very Good	270
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)	No	No	No	No	4	4.9	Dark Green	Excellent	365
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)	Yes	No	No	No	4	4.8	Dark Green	Excellent	229

5 rows × 21 columns

```
In [4]: df.columns
```

```
Out[4]: 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 [5]: df.shape
```

```
Out[5]: (9551, 21)
```

```
In [6]: 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 [7]: df.describe()
```

Out[7]:

	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	9.051128e+06	18.365616	64.126574	25.854381	1199.210763	1.804837	2.666370	156.909748
std	8.791521e+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	0.000000	0.000000
25%	3.019625e+05	1.000000	77.081343	28.478713	250.000000	1.000000	2.500000	5.000000
50%	6.004089e+06	1.000000	77.191964	28.570469	400.000000	2.000000	3.200000	31.000000
75%	1.835229e+07	1.000000	77.282006	28.642758	700.000000	2.000000	3.700000	131.000000
max	1.850065e+07	216.000000	174.832089	55.976980	800000.000000	4.000000	4.900000	10934.000000

In Data Analysis What All Things We Do

1. Missing Values
2. Explore About the Numerical Variables
3. Explore About categorical Variables
4. Finding Relationship between features

```
In [8]: df.isnull()
```

```
Out[8]:
```

	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	False	False	False	False	False	False	False	False	False	...	False	False	False	False	False	False	False	False	False	False
	1	False	False	False	False	False	False	False	False	False	...	False	False	False	False	False	False	False	False	False	False
	2	False	False	False	False	False	False	False	False	False	...	False	False	False	False	False	False	False	False	False	False
	3	False	False	False	False	False	False	False	False	False	...	False	False	False	False	False	False	False	False	False	False
	4	False	False	False	False	False	False	False	False	False	...	False	False	False	False	False	False	False	False	False	False
...
	9546	False	False	False	False	False	False	False	False	False	...	False	False	False	False	False	False	False	False	False	False
	9547	False	False	False	False	False	False	False	False	False	...	False	False	False	False	False	False	False	False	False	False
	9548	False	False	False	False	False	False	False	False	False	...	False	False	False	False	False	False	False	False	False	False
	9549	False	False	False	False	False	False	False	False	False	...	False	False	False	False	False	False	False	False	False	False
	9550	False	False	False	False	False	False	False	False	False	...	False	False	False	False	False	False	False	False	False	False

9551 rows × 21 columns

Ways to see null values

```
In [9]: df.isnull().sum()
```

```
Out[9]:
```

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 [10]: [features for features in df.columns if df[features].isnull().sum()>0]
#for every features in df.columns check the condition.
#random variable in features is iterated throughout the column
```

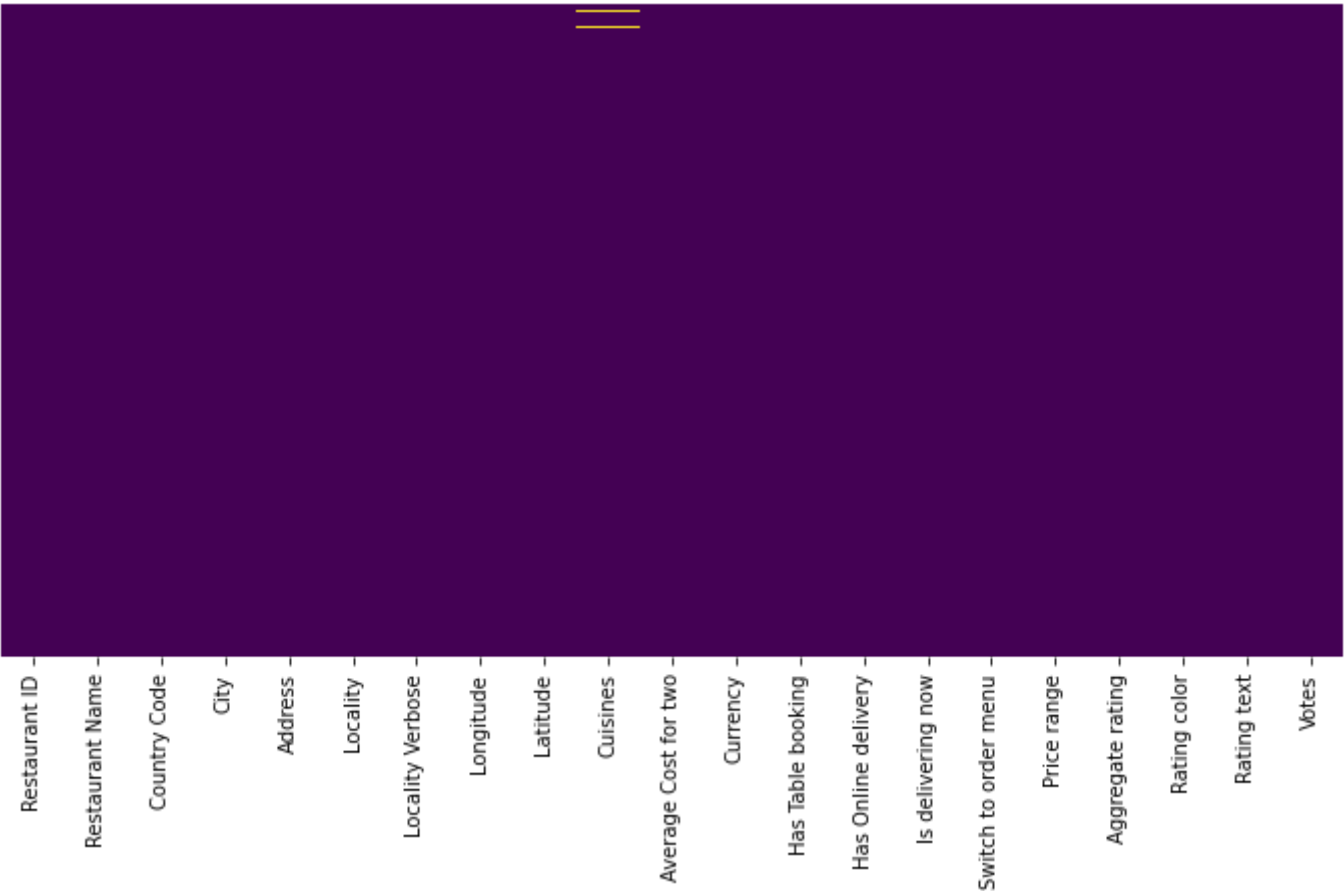
```
Out[10]:
```

['Cuisines']

```
In [11]: matplotlib.rcParams['figure.figsize']=(12,6)
sns.heatmap(df.isnull(),yticklabels=False,cbar=False,cmap='viridis') #small value so cant see
```

```
Out[11]:
```

<AxesSubplot:>



```
In [12]: df_country=pd.read_excel("Country-code.xlsx")
```

```
In [13]: df_country.head()
```

```
Out[13]:
```

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

```
In [14]: final_df=pd.merge(df,df_country,on='Country Code', how="left" )
#combining 2 dataframes
#on=On which feature we combine the 2 tables
#how=what type of join
```

In [15]: final_df.head()

Out[15]:

	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 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	...	Yes	No	No	No	3	4.8	Dark Green	Excellent	314	Phillipines
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	...	Yes	No	No	No	3	4.5	Dark Green	Excellent	591	Phillipines
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	...	Yes	No	No	No	4	4.4	Green	Very Good	270	Phillipines
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	...	No	No	No	No	4	4.9	Dark Green	Excellent	365	Phillipines
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	...	Yes	No	No	No	4	4.8	Dark Green	Excellent	229	Phillipines

5 rows × 22 columns

In [16]: final_df.dtypes

Out[16]:

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 [17]: final_df.Country.value_counts()

Out[17]:

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: Country, dtype: int64

In [18]: country_names=final_df.Country.value_counts().index
country_names

Out[18]:

Index(['India', 'United States', 'United Kingdom', 'Brazil', 'UAE', 'South Africa', 'New Zealand', 'Turkey', 'Australia', 'Phillipines', 'Indonesia', 'Singapore', 'Qatar', 'Sri Lanka', 'Canada'], dtype='object')

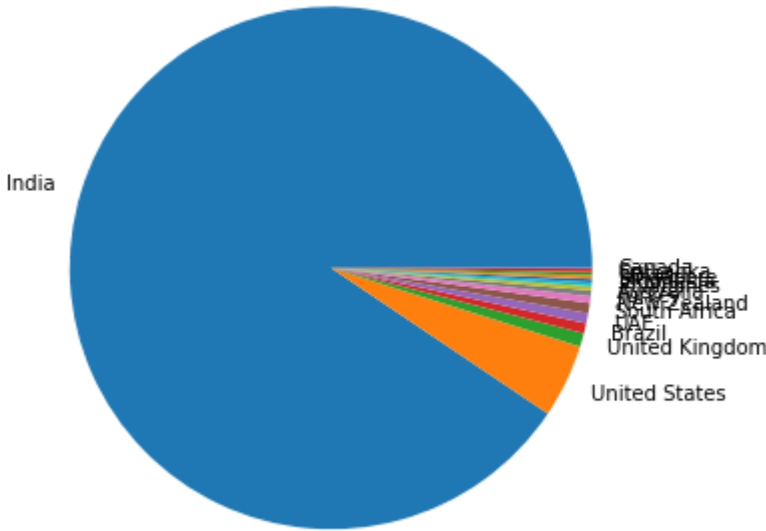
In [19]: country_val=final_df.Country.value_counts().values
country_val

Out[19]:

array([8652, 434, 80, 60, 60, 60, 40, 34, 24, 22, 21, 20, 20, 20, 4], dtype=int64)

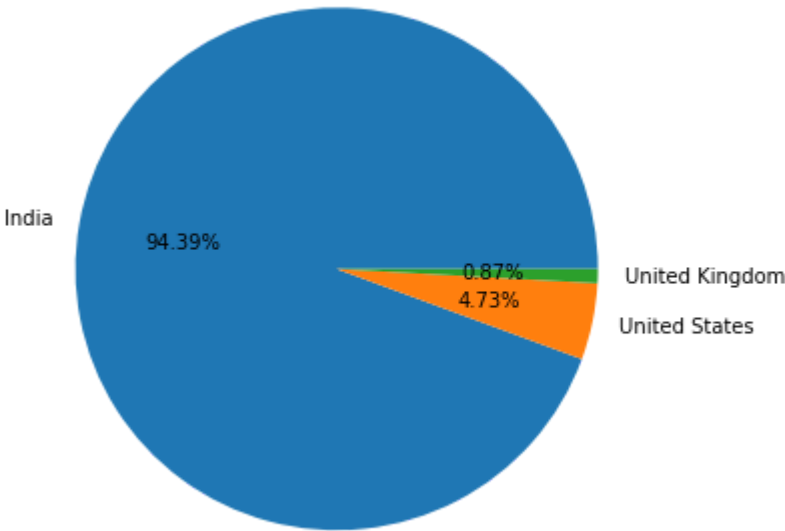
In [20]: plt.pie(country_val,labels=country_names)

Out[20]: ([<matplotlib.patches.Wedge at 0xe274faa10>, <matplotlib.patches.Wedge at 0xe274faf20>, <matplotlib.patches.Wedge at 0xe274fb400>, <matplotlib.patches.Wedge at 0xe274fb8e0>, <matplotlib.patches.Wedge at 0xe274fbd0>, <matplotlib.patches.Wedge at 0xe275242e0>, <matplotlib.patches.Wedge at 0xe275247c0>, <matplotlib.patches.Wedge at 0xe27524ca0>, <matplotlib.patches.Wedge at 0xe27525180>, <matplotlib.patches.Wedge at 0xe27525660>, <matplotlib.patches.Wedge at 0xe274fa9e0>, <matplotlib.patches.Wedge at 0xe27525ff0>, <matplotlib.patches.Wedge at 0xe275264d0>, <matplotlib.patches.Wedge at 0xe275269b0>, <matplotlib.patches.Wedge at 0xe27526e90>], [Text(-1.052256163793291, 0.3205572737577906, 'India'), Text(0.9911329812843455, -0.477132490415823, 'United States'), Text(1.0572858296119743, -0.3035567072257165, 'United Kingdom'), Text(1.070138816916019, -0.2545641619112621, 'Brazil'), Text(1.0793506814479759, -0.21213699926648824, 'UAE'), Text(1.086881147244973, -0.16937937230799818, 'South Africa'), Text(1.0918635911832035, -0.1335436192729486, 'New Zealand'), Text(1.0947903814016446, -0.10692998078388304, 'Turkey'), Text(1.096631023945382, -0.08602556201794338, 'Australia'), Text(1.0978070729776455, -0.06942355882735218, 'Phillipines'), Text(1.0986791544015209, -0.05388984768543213, 'Indonesia'), Text(1.0993059848742366, -0.039068550263413035, 'Singapore'), Text(1.0997248508282123, -0.02460187941736628, 'Qatar'), Text(1.0999533462179636, -0.010130949802716446, 'Sri Lanka'), Text(1.0999990477553414, -0.0014473898376707638, 'Canada')])



In [21]: *## Pie Chart- Top 3 countries that uses zomato*
plt.pie(country_val[:3],labels=country_names[:3],autopct='%1.2f%%')

Out[21]: ([<matplotlib.patches.Wedge at 0xe27591150>, <matplotlib.patches.Wedge at 0xe27591870>, <matplotlib.patches.Wedge at 0xe27591f90>], [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:Zomato maximum records or transaction are from India After that USA and then United Kingdoms

In [22]: final_df.columns

Out[22]: 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 [23]: final_df.groupby(['Aggregate rating','Rating color','Rating text'])
#This is dataframe groupby object

Out[23]: <pandas.core.groupby.generic.DataFrameGroupBy object at 0x0000000E2749BD30>

In [24]: final_df.groupby(['Aggregate rating','Rating color','Rating text']).size()

Out[24]:

Aggregate rating	Rating color	Rating text	
0.0	White	Not rated	2148
1.8	Red	Poor	1
1.9	Red	Poor	2
2.0	Red	Poor	7
2.1	Red	Poor	15
2.2	Red	Poor	27
2.3	Red	Poor	47
2.4	Red	Poor	87
2.5	Orange	Average	110
2.6	Orange	Average	191
2.7	Orange	Average	250
2.8	Orange	Average	315
2.9	Orange	Average	381
3.0	Orange	Average	468
3.1	Orange	Average	519
3.2	Orange	Average	522
3.3	Orange	Average	483
3.4	Orange	Average	498
3.5	Yellow	Good	480
3.6	Yellow	Good	458
3.7	Yellow	Good	427
3.8	Yellow	Good	400
3.9	Yellow	Good	335
4.0	Green	Very Good	266
4.1	Green	Very Good	274
4.2	Green	Very Good	221
4.3	Green	Very Good	174
4.4	Green	Very Good	144
4.5	Dark Green	Excellent	95
4.6	Dark Green	Excellent	78
4.7	Dark Green	Excellent	42
4.8	Dark Green	Excellent	25
4.9	Dark Green	Excellent	61

dtype: int64

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

Out[25]:

	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
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 [26]: ratings=final_df.groupby(['Aggregate rating', 'Rating color', 'Rating text']).size().reset_index().rename(columns={0: 'Rating_count'})
#reset_index=resetting index & converting to dataframe.
ratings
```

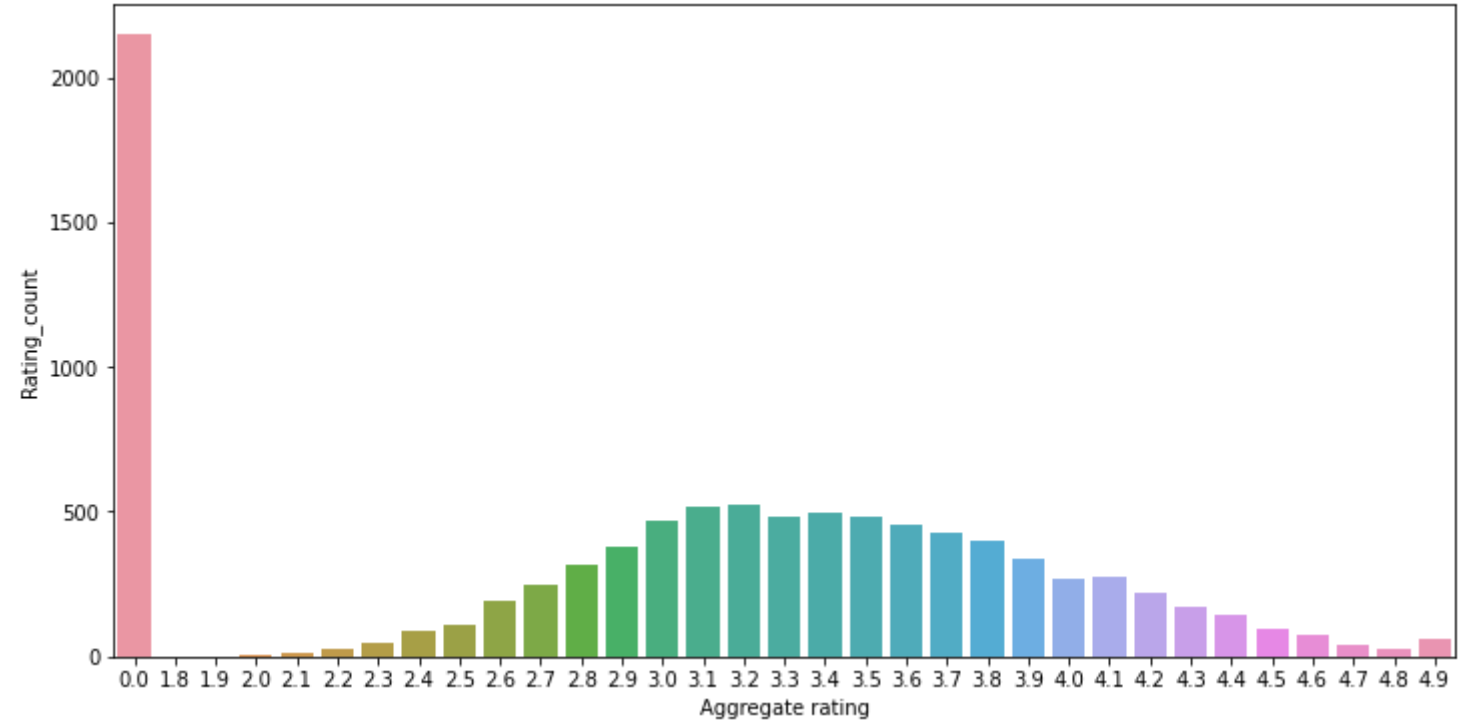
Out[26]:

	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
When Rating is between 4.5 to 4.9---> Excellent
When Rating are between 4.0 to 3.4--->very good
when Rating is between 3.5 to 3.9----> good
when Rating is between 3.0 to 3.4----> average
when Rating is between 2.5 to 2.9----> average
when Rating is between 2.0 to 2.4----> Poor
```

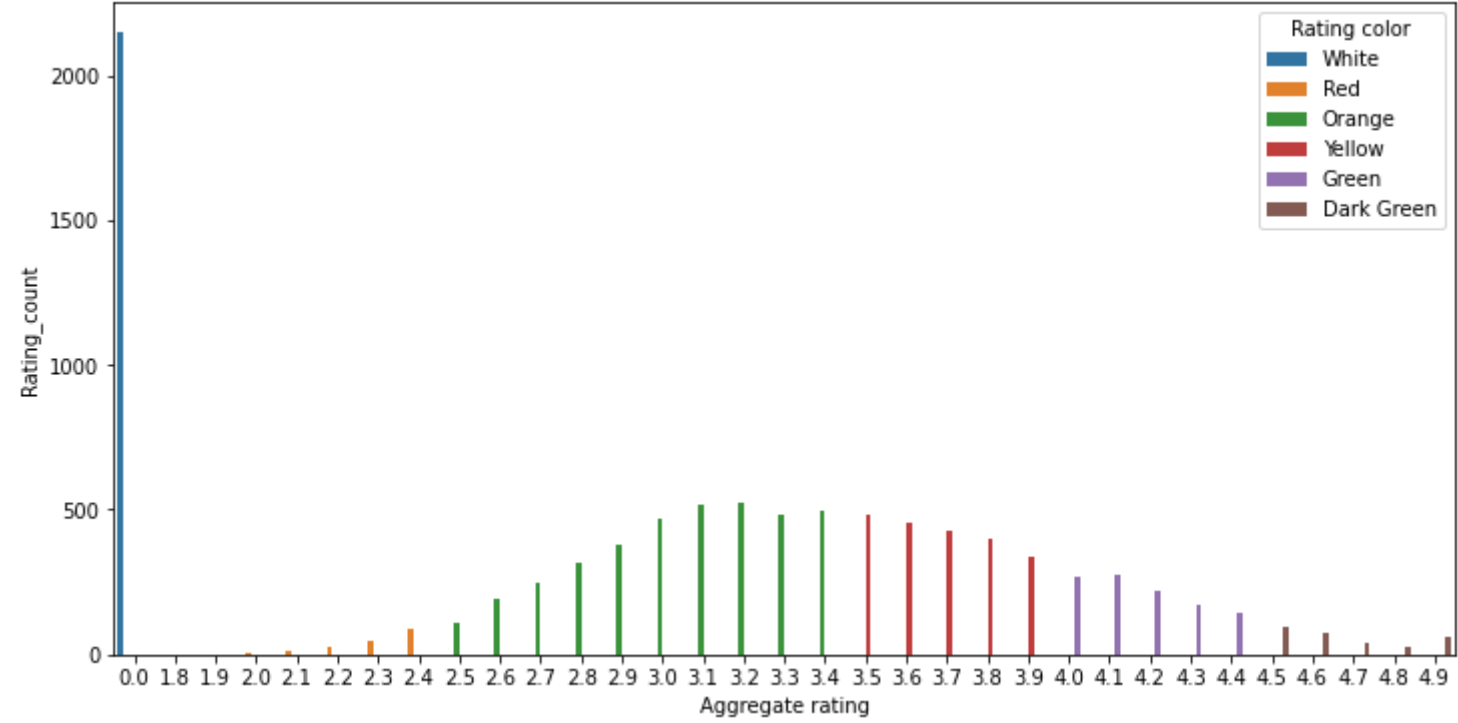
```
In [27]: matplotlib.rcParams['figure.figsize']=(12,6)
sns.barplot(x="Aggregate rating",y="Rating_count",data=ratings)
#It Looks Like gaussian curve.
```

Out[27]: <AxesSubplot:xlabel='Aggregate rating', ylabel='Rating_count'>



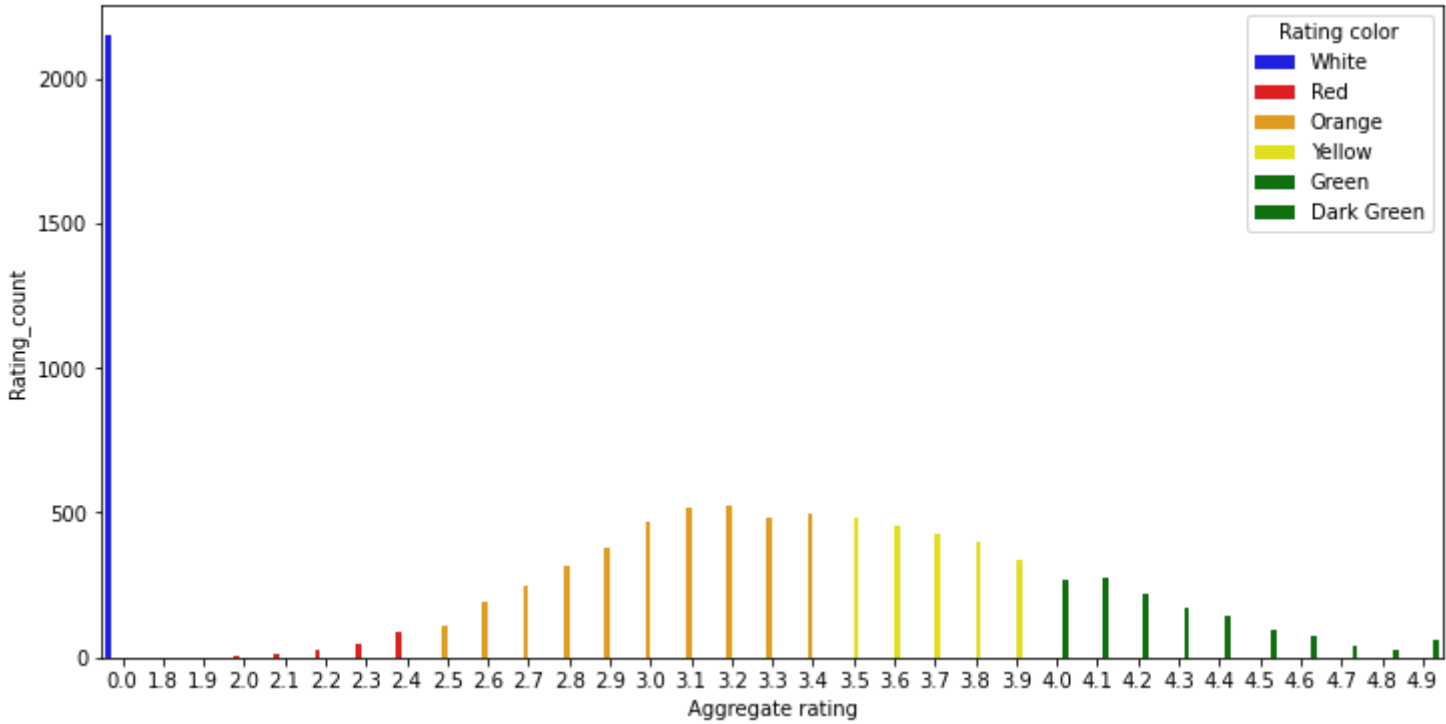
```
In [28]: sns.barplot(x="Aggregate rating",y="Rating_count",hue='Rating color',data=ratings)
#HUE ATTRIBUTE DOESN'T GIVE CORRECT COLOR
```

Out[28]: <AxesSubplot:xlabel='Aggregate rating', ylabel='Rating_count'>



In [29]: sns.barplot(x="Aggregate rating",y="Rating_count",hue='Rating color',data=ratings,palette=['blue','red','orange','yellow','green','green'])

Out[29]: <AxesSubplot:xlabel='Aggregate rating', ylabel='Rating_count'>



Observation:

Not Rated count is very high Maximum number of rating are between 2.5 to 3.4

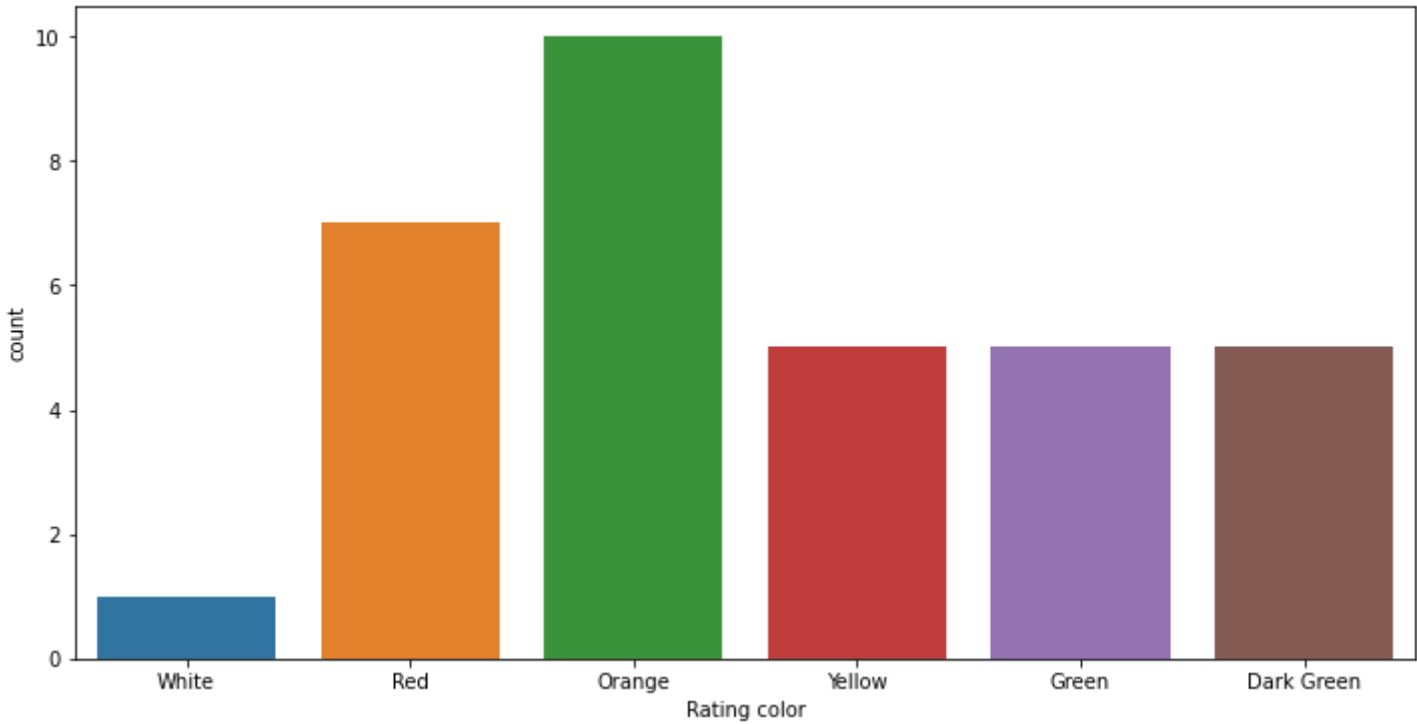
In [30]: ratings

Out[30]:

	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
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15	3.2	Orange	Average	522
16	3.3	Orange	Average	483
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18	3.5	Yellow	Good	480
19	3.6	Yellow	Good	458
20	3.7	Yellow	Good	427
21	3.8	Yellow	Good	400
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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 [31]: sns.countplot(x="Rating color",data=ratings)
#Countplot gives the frequency of this rating color

Out[31]: <AxesSubplot:xlabel='Rating color', ylabel='count'>



In [32]: `### Find the countries name that has given 0 rating`
`final_df[final_df['Rating color']=='White'].groupby('Country').size().reset_index()`

Out[32]:

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

Observations Maximum number of 0 ratings are from Indian customers

In [33]: `##find out which currency is used by which country?`
`final_df.groupby(["Country","Currency"]).size().reset_index()`

Out[33]:

	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

Which Countries do have online deliveries option

In [34]: `final_df[final_df['Has Online delivery']=='Yes'].groupby(['Has Online delivery','Country']).size().reset_index()`
`# final_df[final_df['Rating color']=='White'].groupby('Country').size().reset_index()`

Out[34]:

	Has Online delivery	Country	0
0	Yes	India	2423
1	Yes	UAE	28

Observations:

Online Deliveries are available in India and UAE

In [35]: `## Create a pie chart for top 5 cities distribution`

In [36]: `final_df.City.value_counts().index`

Out[36]: `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 [37]: `final_df.City.value_counts().values`

Out[37]: `array([5473, 1118, 1080, 251, 25, 21, 21, 21, 21, 21, 20,
20, 20, 20, 20, 20, 20, 20, 20, 20, 20, 20,
20, 20, 20, 20, 20, 20, 20, 20, 20, 20, 20,
20, 20, 20, 20, 20, 20, 20, 20, 20, 20, 20,
20, 20, 20, 20, 20, 20, 20, 20, 20, 20, 20,
18, 18, 16, 14, 11, 6, 4, 4, 3, 3, 2,
2, 2, 2, 2, 2, 2, 2, 1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1], dtype=int64)`

In [38]: `city_values=final_df.City.value_counts().values`
`city_labels=final_df.City.value_counts().index`

In [39]: `plt.pie(city_values[:5],labels=city_labels[:5],autopct='%1.2f%%')`

Out[39]: `([<matplotlib.patches.Wedge at 0xe29bb6e00>,
<matplotlib.patches.Wedge at 0xe29bb7550>,
<matplotlib.patches.Wedge at 0xe29bb7c70>,
<matplotlib.patches.Wedge at 0xe29bec3d0>,
<matplotlib.patches.Wedge at 0xe29becaf0>],
[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%')])`

