## **ZOMATO DATASET EXPLORATORY DATA ANALYSIS**

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

Matplotlib is building the font cache; this may take a moment.

In [4]: df = pd.read\_csv("C:\\Users\\Sonali Thakur.DESKTOP-T4FPVTD.000\\Downloads\\zomatc
df.head()

#### Out[4]:

City	Address	Locality	Locality Verbose	Longitude	Latitude	Cuisines	 Currency	Hab Tab bookii
cati 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)	Y
cati 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)	Y
aluyong 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)	Y
aluyong 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)	١
aluyong 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)	Y

#### In [5]: df.columns

#### 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			
dtype	es: float64(3), int64(	5), object(13)				
momony usage 1 FL MD						

memory usage: 1.5+ MB

- here int64 basically means that for integer variable
- · objects in pandas means strings and it can also be categorical variable or a text variable or may be integer variable

## In [7]: df.describe()

#### Out[7]:

	Restaurant ID	Country Code	Longitude	Latitude	Average Cost for two	Price range	Aggreç ral
count	9.551000e+03	9551.000000	9551.000000	9551.000000	9551.000000	9551.000000	9551.000
mean	9.051128e+06	18.365616	64.126574	25.854381	1199.210763	1.804837	2.666
std	8.791521e+06	56.750546	41.467058	11.007935	16121.183073	0.905609	1.516
min	5.300000e+01	1.000000	-157.948486	-41.330428	0.000000	1.000000	0.000
25%	3.019625e+05	1.000000	77.081343	28.478713	250.000000	1.000000	2.500
50%	6.004089e+06	1.000000	77.191964	28.570469	400.000000	2.000000	3.200
75%	1.835229e+07	1.000000	77.282006	28.642758	700.000000	2.000000	3.700
max	1.850065e+07	216.000000	174.832089	55.976980	800000.000000	4.000000	4.900

• the features which are taken inside the describe function are integer features

### IN DATA ANALYSIS THE THINGS WE KNOW

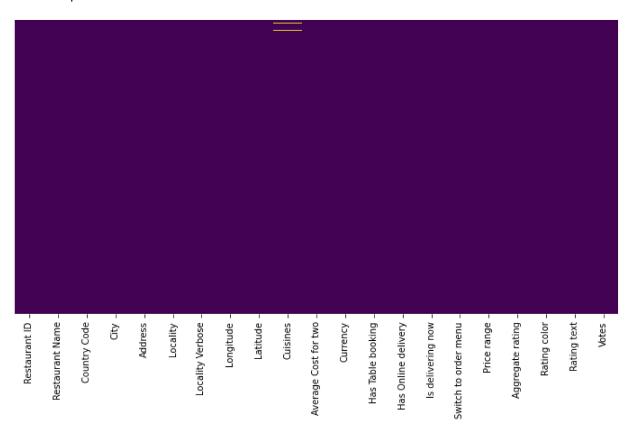
- 1. Missing Values
- 2. Explore about numerical variables
- 3. Explore about categorical variables
- 4. Finding relationship between features.
- IN ORDER TO FIND MISSING VALUES WE DO

```
In [8]: df.isnull().sum()
Out[8]: Restaurant ID
                                 0
        Restaurant Name
                                 0
        Country Code
                                 0
        City
                                 0
        Address
                                 0
        Locality
                                 0
        Locality Verbose
                                 0
        Longitude
        Latitude
                                 0
        Cuisines
                                 9
        Average Cost for two
                                 0
        Currency
                                 0
        Has Table booking
        Has Online delivery
        Is delivering now
        Switch to order menu
                                 0
        Price range
                                 0
        Aggregate rating
                                 0
        Rating color
                                 0
        Rating text
                                 0
                                 0
        Votes
        dtype: int64
In [9]: [features for features in df.columns if df[features].isnull().sum()>0]
```

Out[9]: ['Cuisines']

In [47]: sns.heatmap(df.isnull(),yticklabels=False,cbar=False,cmap='viridis')

#### Out[47]: <AxesSubplot:>



In [12]: df\_country = pd.read\_excel('C:\\Users\\Sonali Thakur.DESKTOP-T4FPVTD.000\\Downloa
df\_country.head()

#### Out[12]:

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

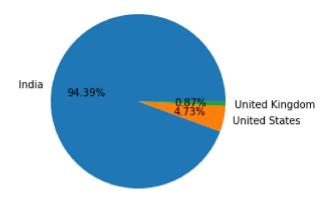
• COMBINING df\_country ALONG WITH df WITH THIS COUNTRYCODE

```
In [18]: final_df = pd.merge(df,df_country,on='Country Code',how='left')
final_df.head(2)
```

#### Out[18]:

Address	Locality	Locality Verbose	Longitude	Latitude	Cuisines	 Has Table booking	Has Online delivery	ls delivering now	•
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	
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	

```
In [19]: ## TO CHECK DATA TYPES
         final_df.dtypes
Out[19]: Restaurant ID
                                    int64
                                   object
         Restaurant Name
         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
                                  float64
         Aggregate rating
         Rating color
                                   object
                                   object
         Rating text
         Votes
                                    int64
         Country
                                   object
         dtype: object
In [20]: |final df.columns
Out[20]: 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 [26]: country names = final df.Country.value counts().index
         country_names
Out[26]: Index(['India', 'United States', 'United Kingdom', 'Brazil', 'UAE',
                 'South Africa', 'New Zealand', 'Turkey', 'Australia', 'Phillipines',
                 'Indonesia', 'Singapore', 'Qatar', 'Sri Lanka', 'Canada'],
               dtype='object')
         country val=final df.Country.value counts().values
In [28]:
         country_val
Out[28]: array([8652, 434,
                                     60,
                                           60,
                                                              34,
                                                                    24,
                                                                          22,
                               80,
                                                 60,
                                                       40,
                                                                                21,
                                   4], dtype=int64)
                         20,
                               20,
                   20,
```



- ZOMATO'S MAXIMUM TRANSCATIONS ARE FROM INDIA. AFTER THAT U.S.A.
- 2. THE COMPANY IS GETTING MORE PROFIT FROM INDIA
- 3. MAJOR BUSINESS IS HAPPENING IN INDIA

In [44]: ratings

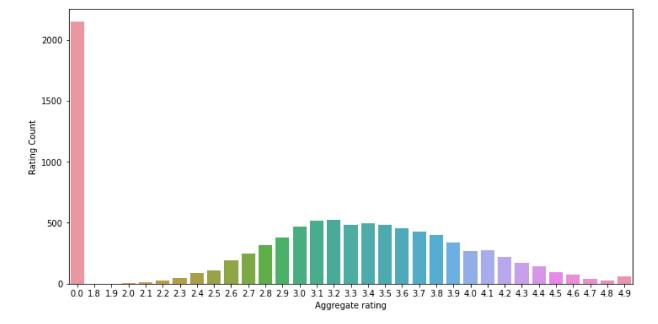
#### Out[44]:

	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

- 1. WHEN RATING IS BETWEEN FROM 4.5 TO 4.9 THE RATING WERE EXCELLENT
- 2. WHEN RATING IS BETWEEN FROM 4.0 TO 4.4 THE RATING WERE VERY GOOD
- 3. WHEN RATING IS BETWEEN FROM 3.5 TO 3.9 THE RATING WERE GOOD
- 4. WHEN RATING IS BETWEEN FROM 2.5 TO 3.4 THE RATING WERE AVERAGE
- 5. WHEN RATING IS BETWEEN FROM 1.8 TO 2.4 THE RATING WERE POOR
- 6. WHEN RATING IS 0 IT MEANS PEOPLE HAS NOT GIVEN ANY RATING

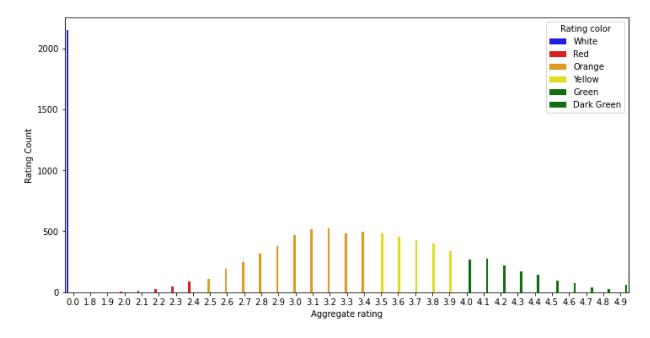
```
In [46]: import matplotlib
matplotlib.rcParams['figure.figsize']=(12,6)
sns.barplot(x="Aggregate rating",y="Rating Count",data=ratings)
```

Out[46]: <AxesSubplot:xlabel='Aggregate rating', ylabel='Rating Count'>



In [51]: sns.barplot(x="Aggregate rating",y="Rating Count",hue='Rating color',data=ratings
✓

Out[51]: <AxesSubplot:xlabel='Aggregate rating', ylabel='Rating Count'>

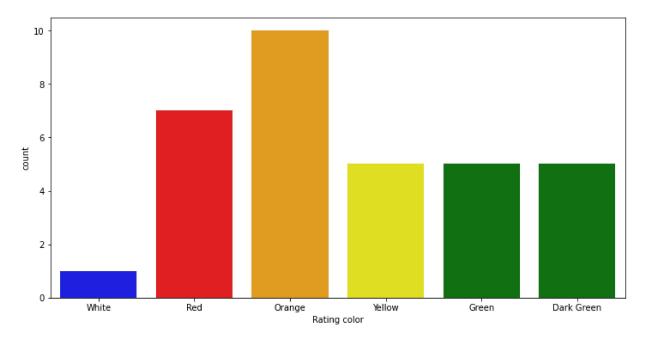


## **OBSERVATION**

- 1. NOT RATED COUNT IS VERY HIGH
- 2. MAXIMUM NUMBER OF RATINGS ARE BETWEEN 2.5 TO 3.4

In [52]: ## COUNT PLOT
sns.countplot(x="Rating color",data=ratings,palette=['blue','red','orange','yellong

Out[52]: <AxesSubplot:xlabel='Rating color', ylabel='count'>



# FIND THE COUNTRIES NAME THAT HAS GIVEN ZERO RATING

In [73]: final\_df.groupby(["Aggregate rating","Country"]).size().reset\_index().head(4)

#### Out[73]:

Δ	Aggregate rating	Country	0
	0.0	Brazil	5
	0.0	India	2139
	0.0	United Kingdom	1
	0.0	United States	3

MAXIMUM NUMBER OF ZERO RATING ARE FROM INDIA

# FIND OUT WHICH CURRENCY IS USED BY WHICH COUNTRY

In [76]: final\_df.groupby(["Country","Currency"]).size().reset\_index()

Out[76]:

	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 DELIEVERY OPTIONS

In [78]: final\_df.groupby(["Country","Has Online delivery"]).size().reset\_index()

Out[78]:

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

```
In [79]: final_df[final_df['Has Online delivery']=='Yes'].Country.value_counts()
```

Out[79]: India

India 2423 UAE 28

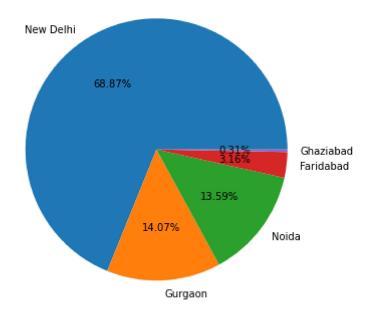
Name: Country, dtype: int64

## **OBSEVATIONS**

1. ONLINE DELIVERIES ARE AVAILABLE IN INDIA AND UAE

# CREATE A PIE CHART FOR TOP 5 CITIES DISTRIBUTION

```
In [83]: | final df.columns
Out[83]: 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 [88]:
         city_values = final_df.City.value_counts().values
         city_labels = final_df.City.value_counts().index
In [90]: |plt.pie(city_values[:5],labels=city_labels[:5],autopct='%1.2f%%')
Out[90]: ([<matplotlib.patches.Wedge at 0x26173713e80>,
           <matplotlib.patches.Wedge at 0x2617371b550>,
           <matplotlib.patches.Wedge at 0x2617371bc70>,
           <matplotlib.patches.Wedge at 0x261737273d0>,
           <matplotlib.patches.Wedge at 0x26173727af0>],
           [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%')])
```

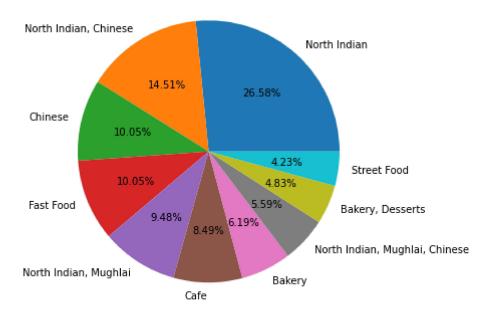


### FIND THE TOP 10 CUISINES

In [100]: cuisines\_values = final\_df.Cuisines.value\_counts().values
 cuisines\_labels = final\_df.Cuisines.value\_counts().index

```
In [102]: plt.pie(cuisines_values[:10],labels=cuisines_labels[:10],autopct='%1.2f%%')
```

```
Out[102]: ([<matplotlib.patches.Wedge at 0x26173946d60>,
            <matplotlib.patches.Wedge at 0x261739484c0>,
            <matplotlib.patches.Wedge at 0x26173948be0>,
            <matplotlib.patches.Wedge at 0x26173912340>,
            <matplotlib.patches.Wedge at 0x26173912a60>,
            <matplotlib.patches.Wedge at 0x2617396d1c0>,
            <matplotlib.patches.Wedge at 0x2617396d8e0>,
            <matplotlib.patches.Wedge at 0x26173966040>,
            <matplotlib.patches.Wedge at 0x26173966760>,
            <matplotlib.patches.Wedge at 0x26173966e80>],
            [Text(0.7383739846958008, 0.8153550507137645, 'North Indian'),
            Text(-0.5794679314239953, 0.9349956772366362, 'North Indian, Chinese'),
            Text(-1.067309479615702, 0.26617752482593154, 'Chinese'),
            Text(-1.0185984499802057, -0.4152796620326146, 'Fast Food'),
            Text(-0.5935788454809928, -0.9261015895664211, 'North Indian, Mughlai'),
            Text(-0.005887079599915552, -1.0999842463843672, 'Cafe'),
            Text(0.4842062514572988, -0.9876964645323336, 'Bakery'),
            Text(0.808736477166136, -0.7456174022251013, 'North Indian, Mughlai, Chines
          e'),
            Text(1.0055375294202338, -0.44597564611473206, 'Bakery, Desserts'),
            Text(1.090298995560443, -0.14576728123927227, 'Street Food')],
            [Text(0.4027494461977095, 0.4447391185711442, '26.58%'),
            Text(-0.316073417140361, 0.5099976421290743, '14.51%'),
            Text(-0.5821688070631101, 0.14518774081414446, '10.05%'),
            Text(-0.5555991545346576, -0.22651617929051704, '10.05%'),
            Text(-0.32377027935326874, -0.5051463215816842, '9.48%'),
            Text(-0.003211134327226664, -0.5999914071187457, '8.49%'),
            Text(0.26411250079489024, -0.5387435261085456, '6.19%'),
            Text(0.441128987545165, -0.40670040121369155, '5.59%'),
            Text(0.5484750160474001, -0.24325944333530836, '4.83%'),
            Text(0.5947085430329688, -0.07950942613051214, '4.23%')])
```



- 1. THE MOST DEMANDED CUISINES IS NORTH INDIAN.
- 2. THE SECOND MOST DEMANDED CUISINES IS NORTH INDIAN, CHINESE.