zomato-analysis

August 3, 2024

[9]:
```
import pandas as  pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as  sns
```

[10]:
```
zom = pd.read_csv('zomato.csv', encoding='latin-1')
zom.head()
```

[10]:

	Restaurant ID	Restaurant Name	Country Code	City	Address	Locality	Locality Verbose	Longitude	Latitude	Cuisines	…	Currency	Has Table booking
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	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				
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				
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				

```
0        French, Japanese, Desserts  …  Botswana Pula(P)               Yes
1                          Japanese  …  Botswana Pula(P)               Yes
2  Seafood, Asian, Filipino, Indian  …  Botswana Pula(P)               Yes
3                   Japanese, Sushi  …  Botswana Pula(P)                No
4                  Japanese, Korean  …  Botswana Pula(P)               Yes

  Has Online delivery Is delivering now Switch to order menu Price range  \
0                  No                No                    No           3
1                  No                No                    No           3
2                  No                No                    No           4
3                  No                No                    No           4
4                  No                No                    No           4

   Aggregate rating  Rating color Rating text Votes
0               4.8    Dark Green   Excellent   314
1               4.5    Dark Green   Excellent   591
2               4.4         Green   Very Good   270
3               4.9    Dark Green   Excellent   365
4               4.8    Dark Green   Excellent   229

[5 rows x 21 columns]
```

[11]:
```
zom.columns
```

[11]:
```
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')
```

[12]:
```
zom.describe()
```

[12]:
```
       Restaurant ID  Country Code    Longitude     Latitude  \
count   9.551000e+03   9551.000000  9551.000000  9551.000000
mean    9.051128e+06     18.365616    64.126574    25.854381
std     8.791521e+06     56.750546    41.467058    11.007935
min     5.300000e+01      1.000000  -157.948486   -41.330428
25%     3.019625e+05      1.000000    77.081343    28.478713
50%     6.004089e+06      1.000000    77.191964    28.570469
75%     1.835229e+07      1.000000    77.282006    28.642758
max     1.850065e+07    216.000000   174.832089    55.976980

       Average Cost for two  Price range  Aggregate rating         Votes
count           9551.000000  9551.000000       9551.000000   9551.000000
mean            1199.210763     1.804837          2.666370    156.909748
```

```
std            16121.183073     0.905609          1.516378    430.169145
min                0.000000     1.000000          0.000000      0.000000
25%              250.000000     1.000000          2.500000      5.000000
50%              400.000000     2.000000          3.200000     31.000000
75%              700.000000     2.000000          3.700000    131.000000
max           800000.000000     4.000000          4.900000  10934.000000
```

[13]:
```
zom.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
```

0.1 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

[14]:
```
zom.isnull().sum()
```

[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
```

[15]:
```
[features for features in zom.columns if zom[features].isnull().sum()>0]
```

[15]:
```
['Cuisines']
```

[16]:
```
coun = pd.read_excel("Country-Code.xlsx")
coun.head()
```

[16]:

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

[17]:
```
final_df = pd.merge(zom,coun,on='Country Code', how = 'left')
```

[18]:
```
final_df.head()
```

[18]:

	Restaurant ID	Restaurant Name	Country Code	City	\
0	6317637	Le Petit Souffle	162	Makati City	
1	6304287	Izakaya Kikufuji	162	Makati City	
2	6300002	Heat - Edsa Shangri-La	162	Mandaluyong City	
3	6318506	Ooma	162	Mandaluyong City	
4	6314302	Sambo Kojin	162	Mandaluyong City	

```
                                              Address  \
0  Third Floor, Century City Mall, Kalayaan Avenu…
1  Little Tokyo, 2277 Chino Roces Avenue, Legaspi…
2  Edsa Shangri-La, 1 Garden Way, Ortigas, Mandal…
3  Third Floor, Mega Fashion Hall, SM Megamall, O…
4  Third Floor, Mega Atrium, SM Megamall, Ortigas…

                                    Locality  \
0   Century City Mall, Poblacion, Makati City
1  Little Tokyo, Legaspi Village, Makati City
2  Edsa Shangri-La, Ortigas, Mandaluyong City
3      SM Megamall, Ortigas, Mandaluyong City
4      SM Megamall, Ortigas, Mandaluyong City

                                  Locality Verbose   Longitude   Latitude  \
0  Century City Mall, Poblacion, Makati City, Mak…  121.027535  14.565443
1  Little Tokyo, Legaspi Village, Makati City, Ma…  121.014101  14.553708
2  Edsa Shangri-La, Ortigas, Mandaluyong City, Ma…  121.056831  14.581404
3  SM Megamall, Ortigas, Mandaluyong City, Mandal…  121.056475  14.585318
4  SM Megamall, Ortigas, Mandaluyong City, Mandal…  121.057508  14.584450

                          Cuisines  …  Has Table booking  \
0       French, Japanese, Desserts  …                Yes
1                         Japanese  …                Yes
2  Seafood, Asian, Filipino, Indian  …                Yes
3                  Japanese, Sushi  …                 No
4                 Japanese, Korean  …                Yes

  Has Online delivery Is delivering now Switch to order menu Price range  \
0                  No                No                    No           3
1                  No                No                    No           3
2                  No                No                    No           4
3                  No                No                    No           4
4                  No                No                    No           4

  Aggregate rating  Rating color  Rating text Votes      Country
0              4.8    Dark Green    Excellent   314  Phillipines
1              4.5    Dark Green    Excellent   591  Phillipines
2              4.4         Green    Very Good   270  Phillipines
3              4.9    Dark Green    Excellent   365  Phillipines
4              4.8    Dark Green    Excellent   229  Phillipines

[5 rows x 22 columns]
```

[19]:
```
final_df.dtypes
```

[19]:
```
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
```

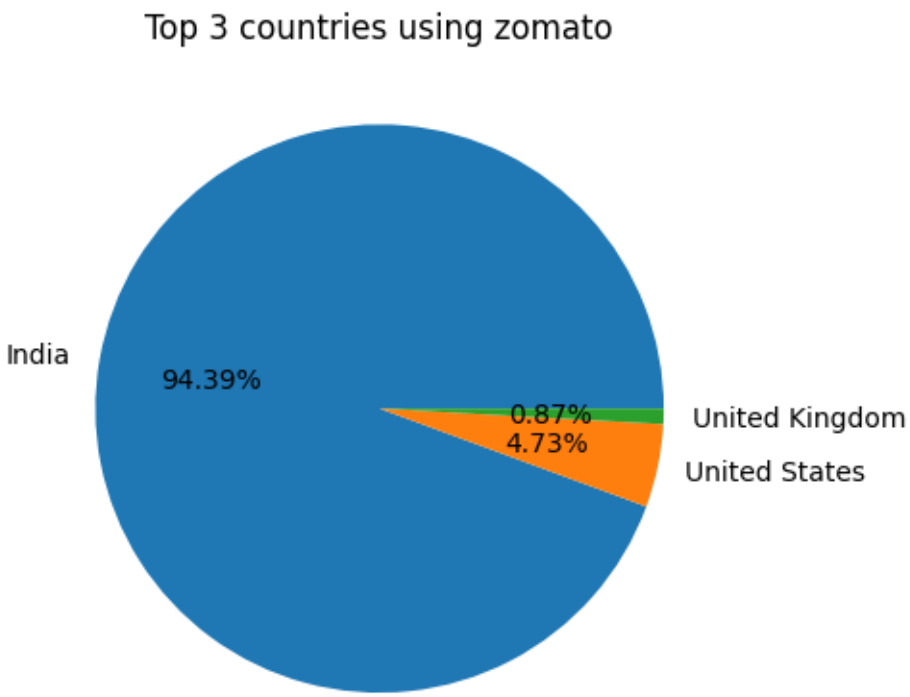
[20]:
```
final_df['Country'].value_counts()
```

[20]:
```
Country
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: count, dtype: int64
```

[21]:
```
country_names = final_df.Country.value_counts().index
```

[22]:
```
country_val = final_df.Country.value_counts().values
```

[23]:
```
## pie plot for the top 3 countries using zomato
plt.pie(country_val[:3], labels=country_names[:3], autopct= "%1.2f%%")
plt.title("Top 3 countries using zomato")
plt.savefig('top_3.jpg')
```



Observation: Zomato maximum records or transactions are from India After that USA and then United Kingdom.

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

[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')
```

[25]:
```
Ratings = final_df.groupby(['Aggregate rating','Rating color','Rating text']).
  ↪size().reset_index().rename(columns={0:"Rating Count"})
Ratings
```

[25]:

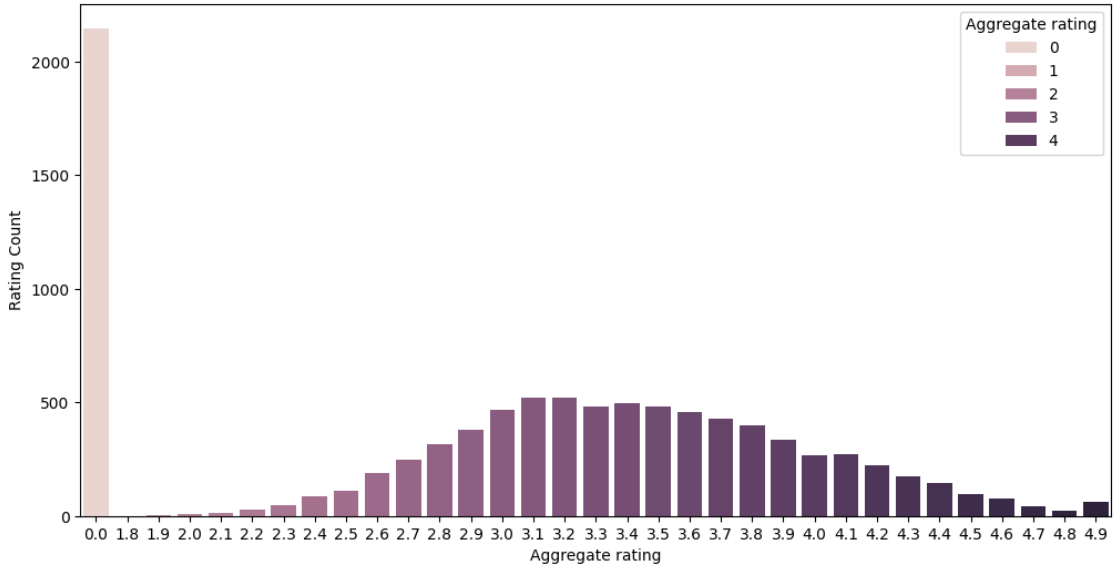
	Aggregate rating	Rating color	Rating text	Rating Count
0	0.0	White	Not rated	2148
1	1.8	Red	Poor	1
2	1.9	Red	Poor	2
3	2.0	Red	Poor	7
4	2.1	Red	Poor	15
5	2.2	Red	Poor	27
6	2.3	Red	Poor	47
7	2.4	Red	Poor	87
8	2.5	Orange	Average	110
9	2.6	Orange	Average	191
10	2.7	Orange	Average	250
11	2.8	Orange	Average	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

0.2 Observations

1. When Rating is between 4.5 to 4.9 —-> Excellent
2. When Rating are 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

[26]:
```
plt.figure(figsize=[12,6])
sns.barplot(x="Aggregate rating", y = "Rating Count", hue = "Aggregate rating",␣
 ↪data = Ratings)
```

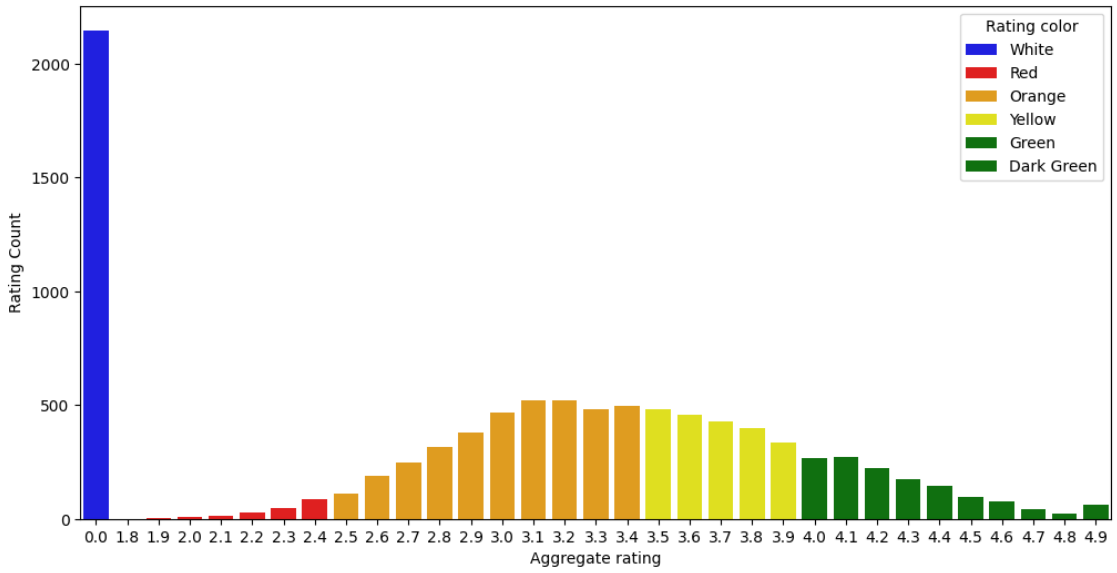
```
plt.savefig("Ag_rc.jpg")
```



[27]:
```
Ratings['Rating color'].unique()
```

[27]:
```
array(['White', 'Red', 'Orange', 'Yellow', 'Green', 'Dark Green'],
      dtype=object)
```

[28]:
```
plt.figure(figsize=[12,6])
sns.barplot(x="Aggregate rating", y = "Rating Count", hue = "Rating color",␣
 ↪data = Ratings, palette=['blue', 'Red', 'Orange', 'Yellow', 'Green',␣
 ↪'Green'])
plt.savefig('Rc_RC.jpg')
```



1 Observation:

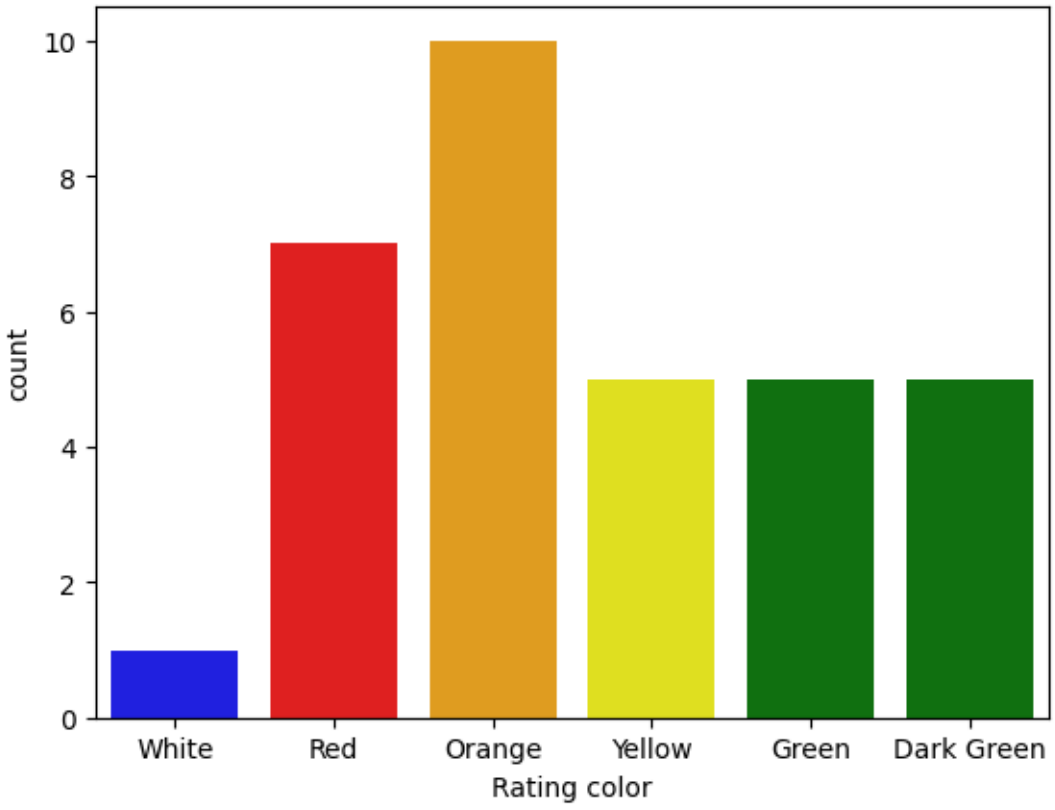
1. Highest number of people has given 0 rating.
2. The average rating is in between 2.5-3.4.

[29]:
```
## count plot
sns.countplot(x = 'Rating color', data = Ratings, palette=['blue', 'Red',␣
 ↪'Orange', 'Yellow', 'Green', 'Green'])
plt.savefig('RC_cp.jpg')
```

```
C:\Users\nagap\AppData\Local\Temp\ipykernel_21508\1676193572.py:2:
FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in
v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same
effect.

  sns.countplot(x = 'Rating color', data = Ratings, palette=['blue', 'Red',
'Orange', 'Yellow', 'Green', 'Green'])
```



[30]:
```
## find the countries names which has given 0 rating
final_df[final_df['Rating color'] == "White"].groupby(['Aggregate␣
↪rating','Country']).size().reset_index()
```

[30]:

	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

1.1 Observation

india holds the highest 0 ratings

[31]:
```
final_df.columns
```

[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')
```

[32]:
```
## find which currency is used by which country
final_df[['Country','Currency']].groupby(['Country','Currency']).size().
 ↪reset_index()
```

[32]:

	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

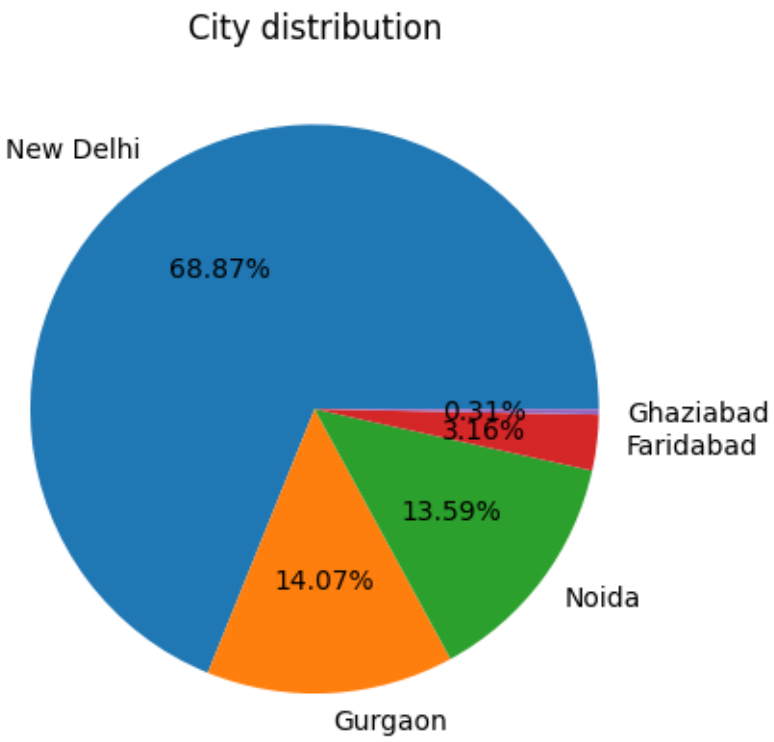
[33]:
```
final_df[final_df['Has Online delivery'] == "Yes"].Country.value_counts()
```

[33]:
```
Country
India    2423
UAE        28
Name: count, dtype: int64
```

1.2 Observation

Online deliveries were avaliable in the India and UAE.

[35]:
```
## Create the city distribution
plt.pie(final_df.City.value_counts().values[:5],
        labels = final_df.City.value_counts().index[:5],autopct="%1.2f%%")
plt.title("City distribution")
plt.show()
plt.savefig('city_dis.png')
```




```
<Figure size 640x480 with 0 Axes>
```

1.3 Observations

1. Top 5 cities are from india
2. Maximum nuber of transactions were taking places in New Delhi.
3. Online deliveries were avaliable in the India and UAE.
4. india holds the highest 0 ratings
5. Highest number of people has given 0 rating.
6. The average rating is in between 2.5-3.4.
    1. When Rating is between 4.5 to 4.9 —-> Excellent
    2. When Rating are 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
7. Observation: Zomato maximum records or transactions are from India After that USA and then United Kingdom.