Project-Zomato API 2

June 25, 2020

- 0.0.1 Question 1: The dataset is highly skewed toward the cities included in Delhi-NCR. So, we will summarise all the other cities in Rest of India while those in New Delhi, Ghaziabad, Noida, Gurgaon, Faridabad to Delhi-NCR. Doing this would make our analysis turn toward Delhi-NCR v Rest of India.
- 1.1: Plot the bar graph of number of restaurants present in Delhi NCR vs Rest of India. To do so first we will import dataset and do data cleaning which is required.

[1]:		Restaurant ID	Restaurant Name Coun	try Code	\
	624	3400025	Jahanpanah	1	
	625	3400341	Rangrezz Restaurant	1	
	626	3400005	Time2Eat - Mama Chicken	1	
	627	3400021	Chokho Jeeman Marwari Jain Bhojanalya	1	
	628	3400017	Pinch Of Spice	1	

	9271	2800100	D Cabana	1	
	9272	2800418	Kaloreez	1	
	9273	2800881	Plot 17	1	
	9274	2800042	Vista - The Park	1	
	9275	2800019	Flying Spaghetti Monster	1	
		City	Address	\	
	624	Agra E 23, S	Shopping Arcade, Sadar Bazaar, Agra Cant		
	625	Agra E-20, S	Shopping Arcade, Sadar Bazaar, Agra Cant		
	626	Agra 1	Main Market, Sadar Bazaar, Agra Cantt, Agra		

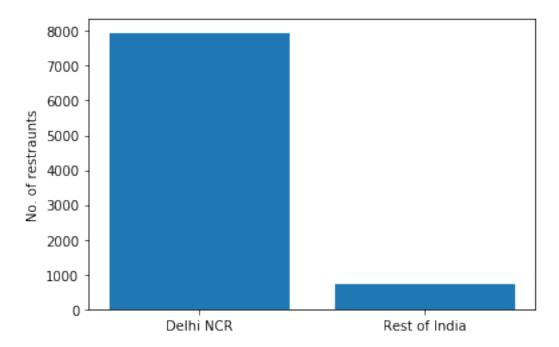
```
627
       Agra 1/48, Delhi Gate, Station Road, Raja Mandi, Ci...
628
             23/453, Opposite Sanjay Cinema, Wazipura Road,...
       Agra
             Beach Road, Near Bus Stop, Sagar Nagar, Visakh...
9271
      Vizag
9272
             Plot 95, Opposite St. Lukes Nursing School, Da...
      Vizag
9273
                     Plot 17, Gangapur Layout, Siripuram, Vizag
      Vizag
9274
             The Park, Beach Road, Pedda Waltair, Lawsons B...
      Vizag
9275
      Vizag
             10-50-12/F2, Sai Dakshata Complex, Beside Leno...
                                           Locality Verbose
                    Locality
                                                             Longitude
624
                                           Agra Cantt, Agra
                                                             78.011544
                 Agra Cantt
625
                 Agra Cantt
                                           Agra Cantt, Agra
                                                               0.00000
626
                 Agra Cantt
                                           Agra Cantt, Agra
                                                             78.011608
627
                Civil Lines
                                         Civil Lines, Agra
                                                             77.998092
628
                Civil Lines
                                         Civil Lines, Agra
                                                             78.007553
9271
                                         Sagar Nagar, Vizag
                                                             83.361377
                Sagar Nagar
9272
                                           Siripuram, Vizag
                  Siripuram
                                                               0.000000
9273
                  Siripuram
                                           Siripuram, Vizag
                                                             83.315281
9274
                              The Park, Lawsons Bay, Vizag
      The Park, Lawsons Bay
                                                              83.336840
9275
            Waltair Uplands
                                    Waltair Uplands, Vizag
                                                             83.314942
       Latitude
                                                             Cuisines
624
      27.161661
                                               North Indian, Mughlai
625
       0.00000
                                               North Indian, Mughlai ...
626
      27.160832
                                                        North Indian ...
                                                          Rajasthani
627
      27.195928
628
      27.201725
                                     North Indian, Chinese, Mughlai ...
                  Continental, Seafood, Chinese, North Indian, B...
9271
      17.764287
9272
       0.000000
                                         Cafe, North Indian, Chinese ...
9273
      17.719539
                                              Burger, Pizza, Biryani
9274
      17.721182
                          American, North Indian, Thai, Continental
9275
      17.721119
                                                              Italian ...
                Currency Has Table booking Has Online delivery
624
      Indian Rupees(Rs.)
                                         No
                                                               No
625
      Indian Rupees(Rs.)
                                         No
                                                               No
626
      Indian Rupees(Rs.)
                                         No
                                                               No
627
      Indian Rupees(Rs.)
                                         No
                                                               No
628
      Indian Rupees (Rs.)
                                         No
                                                               No
9271
      Indian Rupees(Rs.)
                                         No
                                                               No
9272 Indian Rupees(Rs.)
                                         No
                                                               No
9273
      Indian Rupees(Rs.)
                                         No
                                                               No
9274
      Indian Rupees(Rs.)
                                         No
                                                               No
9275
      Indian Rupees(Rs.)
                                         No
                                                               No
```

```
Is delivering now Switch to order menu Price range Aggregate rating \
624
                     No
                                            No
                                                          3
                                                                            3.9
                                                          2
625
                                                                            3.5
                     No
                                            No
626
                     No
                                            No
                                                          2
                                                                            3.6
627
                                                          2
                                                                            4.0
                     No
                                            No
628
                                            No
                                                          3
                                                                            4.2
                     No
                                                          2
9271
                                                                            3.6
                                            No
                     No
9272
                     No
                                            No
                                                          2
                                                                            3.7
                                                          2
                                                                            4.3
9273
                     No
                                            No
9274
                     No
                                            No
                                                          4
                                                                            3.8
9275
                     No
                                            No
                                                          3
                                                                            4.4
      Rating color Rating text Votes
624
            Yellow
                            Good
                                    140
```

625 Yellow Good 71 626 Yellow Good 94 627 Green Very Good 87 628 Green Very Good 177 ••• ... 193 9271 Yellow Good 9272 Yellow Good 85 9273 Green Very Good 172 Good 9274 Yellow 74 9275 Green Very Good 316

[8652 rows x 21 columns]

Total Restaurants 8652
Total restaurants in Delhi NCR 7947
Total restaurants in Rest of India 705



1.2: Find the cuisines which are not present in restaurant of Delhi NCR but present in rest of India. Check using Zomato API whether this cuisines are actually not served in restaurants of Delhi-NCR or just it due to incomplete dataset.

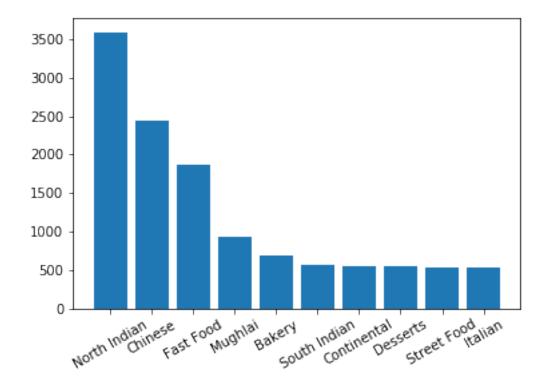
```
Set of cuisines served in rest of India but not in delhi NCR {'Cajun', 'BBQ', 'Malwani', 'German'}
Cajun False
BBQ True
Malwani True
German False
```

1.3: Find the top 10 cuisines served by maximum number of restaurants in Delhi NCR and rest of India. For Delhi NCR

```
[20]: # A dictionary is created for cuisines served in Delhi NCR.
      # Key = cuisine, Value = No. of restaurants that serve that cuisine
      df1 = pd.DataFrame(d1) # d1 is defined in previous block
      dic1 = \{\}
      for index, row in df1.iterrows():
          for i in row['Cuisines']:
              dic1[i] = dic1.get(i,0) + 1
      # arrays x and y are populated to plot graph
      count = 0
      x = \prod
      y = []
      for w in sorted(dic1, key = dic1.get, reverse = True):
          x.append(w)
          y.append(dic1[w])
          print (w, dic1[w])
          count += 1
          if count == 10:
              break
      # To plot required graph
      plt.bar(x, y, align = 'center')
```

```
plt.xticks(range(len(x)), list(x), rotation = 30)
plt.show()
```

North Indian 3597 Chinese 2448 Fast Food 1866 Mughlai 933 Bakery 697 South Indian 569 Continental 547 Desserts 542 Street Food 538 Italian 535



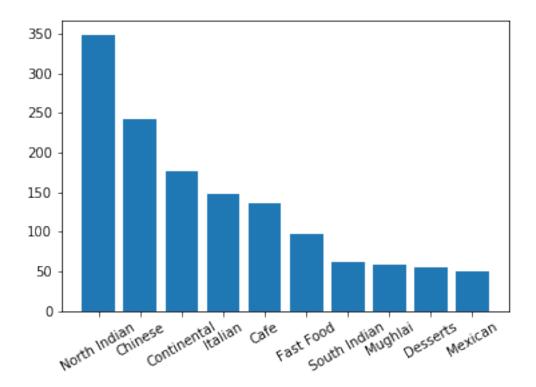
For Rest of India

```
[21]: # Same as before
    df2 = pd.DataFrame(d2)
    dic2 = {}
    for index, row in df2.iterrows():
        for i in row['Cuisines']:
            dic2[i] = dic2.get(i,0) + 1
    count = 0
    x = []
```

```
y = []
for w in sorted(dic2, key = dic2.get, reverse=True):
    x.append(w)
    y.append(dic2[w])
    print (w, dic2[w])
    count += 1
    if count == 10:
        break

plt.bar(x, y, align = 'center')
plt.xticks(range(len(x)), list(x), rotation = 30)
plt.show()
```

North Indian 349 Chinese 242 Continental 177 Italian 147 Cafe 136 Fast Food 97 South Indian 62 Mughlai 59 Desserts 55 Mexican 50



- 1.4: Write a short detailed analysis of how cuisine served is different from Delhi NCR to Rest of India. Plot the suitable graph to explain your inference.
 - 1. North Indian and Chinese cusines are served the most, both in Delhi NCR and Rest of India
 - 2. After that, fast food and mughlai is more popular in Delhi NCR while continental and Italian restaurants in Rest of the India
 - 3. Overall number of restaurants registered with Zomato is much more in Delhi NCR than in Rest of India.
 - 4. Previous 2 graphs are depicting these trends well
- 0.0.2 Question 2: User Rating of a restaurant plays a crucial role in selecting a restaurant or ordering the food from the restaurant.
- 2.1: Write a short detail analysis of how the rating is affected by restaurant due following features: Plot a suitable graph to explain your inference.

2.1.1: Number of Votes given Restaurant

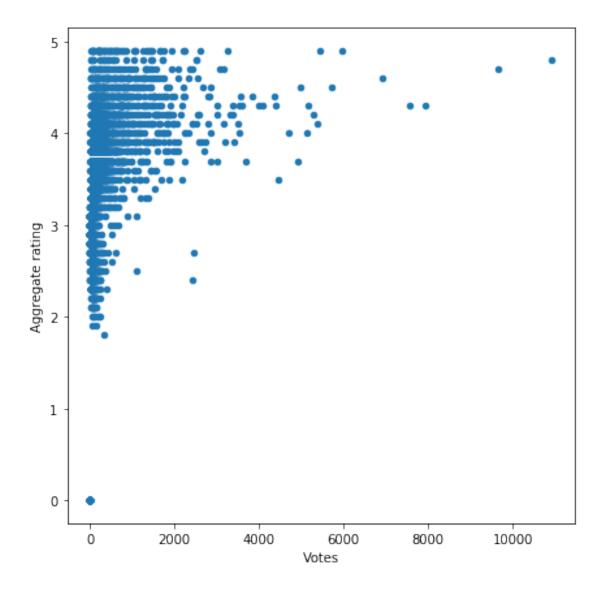
```
[5]: import numpy as np

df = pd.read_csv('zomato.csv', encoding = 'ISO-8859-1')
    df1 = df[['Aggregate rating','Votes']]
    df1 = df1.sort_values(['Votes'], ascending=[True])
    df1

import matplotlib.pyplot as plt
    %matplotlib inline

df1.plot(x='Votes', y='Aggregate rating', kind = 'scatter', figsize = (7,7))
```

[5]: <matplotlib.axes._subplots.AxesSubplot at 0x7f983c820e90>

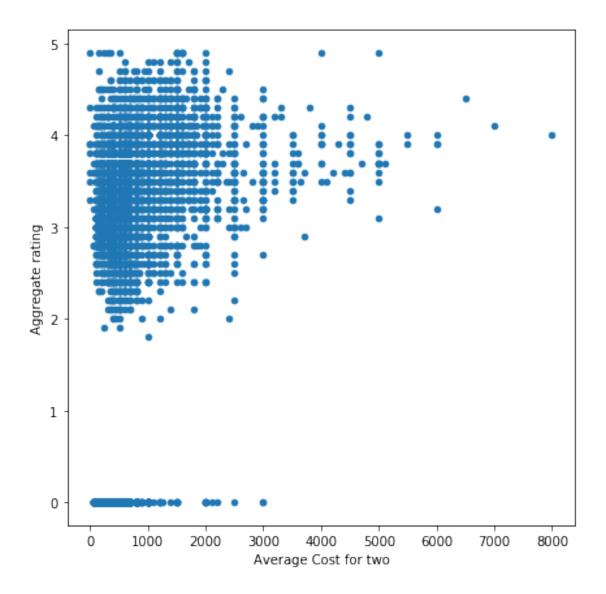


For votes < 2000 No particular trend can be seen.

For votes > 4000 Ratings go up as number of votes increase (maybe because restaurant is good that's why more people are visiting the restaurant)

2.1.2: Restaurant serving more number of cuisines.

[6]: <matplotlib.axes._subplots.AxesSubplot at 0x7f983c8fee10>



Only Indian restaurants are considered because different currencies in different countries.

No trend can be seen for $\cos t < 3000$

For cost > 3000, rating is > 3

2.1.3: Average Cost of Restaurant

```
[10]: df3 = df[['Cuisines', 'Aggregate rating']]
df3 = df3[df3['Cuisines'].notnull()]

def f(a):
    if a is np.nan :
        return 0
    s = a.split(',')
```

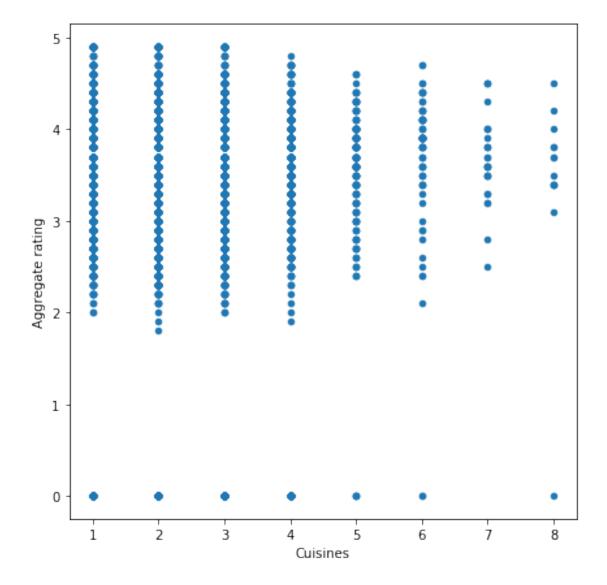
```
return len(s)

df3['Cuisines'] = df3['Cuisines'].apply(lambda x: f(x))

df3.plot(x = 'Cuisines', y = 'Aggregate rating', kind = 'scatter', figsize =

→(7, 7))
```

[10]: <matplotlib.axes._subplots.AxesSubplot at 0x7f983e037b90>



No trend whatsoever

2.1.4: Restaurant serving some specific cuisines.

```
[11]: df4 = df[['Cuisines', 'Aggregate rating']]
df4 = df4[df4['Cuisines'].notnull()]
```

```
[11]:
                  cuisine count
                                      mean
      209
            North Indian
                            2992 2.399699
                  Chinese
                            1880 2.755319
      33
      46
                Fast Food
                           1314 2.671537
      92
            North Indian
                            968 2.852789
      157
                  Chinese
                             855 2.325731
      . .
      221
                               1 4.600000
                    Ramen
      162 Cuisine Varies
                               1 0.000000
      161
                    Cuban
                               1 4.400000
      83
                  Mineira
                               1 0.000000
      44
                  Dí ner
                               1 4.400000
```

[249 rows x 3 columns]

Number of cuisines are too many to plot, hence only dataframe is shown

Each cuisine has very different number of restaurants which serve that cuisine therefore, more popular cuisines actually have very less average rating hence no conclusions can be made

2.2: Find the weighted restaurant rating of each locality and find out the top 10 localities with more weighted restaurant rating?

```
Weighted Restaurant Rating=\Sigma (number of votes * rating) / \Sigma (number of votes).
```

```
[12]: # A new column 'prod' is created which is the product of rating into votes df['prod'] = df['Aggregate rating'] * df['Votes']
```

[12]:		Votes	prod	$w_{\mathtt{rating}}$
	Locality			
	Penjaringan	605	2964.5	4.9
	Aminabad	1057	5179.3	4.9
	Cengkareng	1662	8143.8	4.9
	Barwa Towers, Al Sadd	182	891.8	4.9
	Kenwood	1424	6977.6	4.9
	Hotel Clarks Amer, Malviya Nagar	322	1577.8	4.9
	Venetian Village, Al Maqtaa	216	1058.4	4.9
	Setor De Clubes Esportivos Sul	30	147.0	4.9
	Caddebostan	522	2557.8	4.9
	New Tampa	1413	6923.7	4.9

0.0.3 Question 3: Visualization

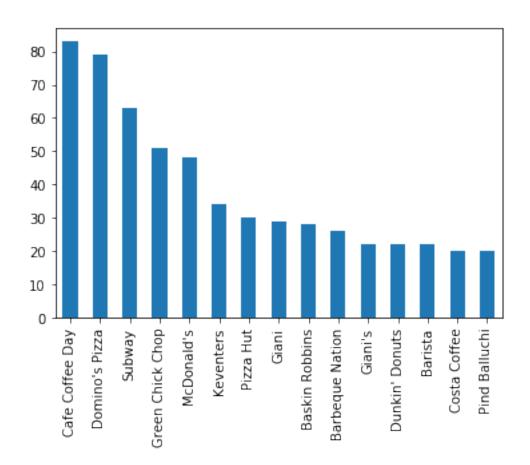
3.1: Plot the bar graph for top 15 restaurants have a maximum number of outlets.

```
[26]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline

df = pd.read_csv('zomato.csv', encoding = 'ISO-8859-1')

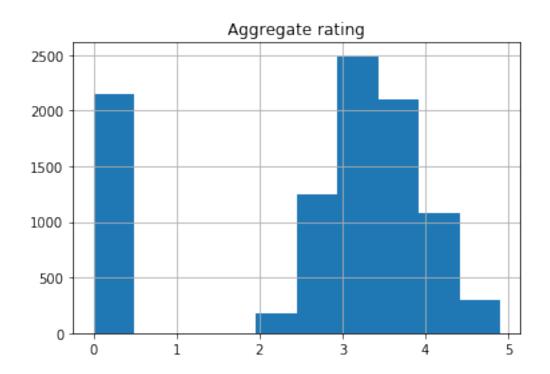
df['Restaurant Name'].value_counts().head(15).plot(kind='bar')
```

[26]: <matplotlib.axes._subplots.AxesSubplot at 0x7f7f0f92af90>



3.2: Plot the histogram of aggregate rating of restaurant (drop the unrated restaurant).

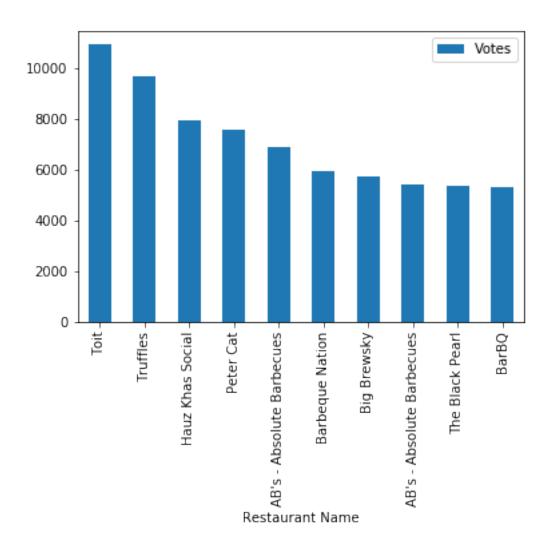
```
[27]: df2 = df[df['Aggregate rating'].notnull()]
df2.hist(column = 'Aggregate rating')
```



3.3: Plot the bar graph top 10 restaurants in the data with the highest number of votes.

```
[28]: df3 = df.sort_values(['Votes'], ascending=[False]).head(10)
df3.plot(x='Restaurant Name', y='Votes', kind = 'bar')
```

[28]: <matplotlib.axes._subplots.AxesSubplot at 0x7f7f0cd45b50>



3.4: Plot the pie graph of top 10 cuisines present in restaurants in the USA. In the following code I first split the cuisines and made separate row for each cuisine. Then I count the number of each cuisines present in datafram and sorted and took top 10

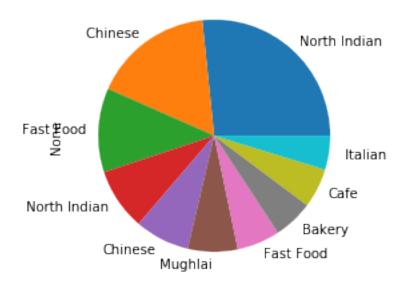
```
[30]: df4 = df[df['Country Code'] == 216]
df4 = df[['Cuisines']]
df4 = df4[df4['Cuisines'].notnull()]

def f(a):
    s = a.split(',')
    return s

df4['Cuisines'] = df4['Cuisines'].apply(lambda x: f(x))

df4.reset_index(inplace = True)
```

[30]: <matplotlib.axes._subplots.AxesSubplot at 0x7f7f0db58290>



3.5: Plot the bubble graph of a number of Restaurants present in the city of India and keeping the weighted restaurant rating of the city in a bubble.

```
[36]: df5 = df[df['Country Code'] == 1].copy(deep = True)
    pd.options.mode.chained_assignment = None
    df5['prod'] = df5['Aggregate rating'] * df5['Votes']

a = df5.groupby("City").agg(['sum', 'count'])
a = a[['Votes', 'prod']]
a['w'] = a['prod']['sum'] / a['Votes']['sum']
a = a[['w', 'Votes']]
a.columns = a.columns.droplevel()
del a['sum']

a = a.reset_index(drop = False)
a.columns = ['City', 'w', 'count']
```

```
xi = np.array(a['City'])
yi = np.array(a['count'])
si = np.array(a['w'])

plt.figure(figsize = (20, 10))
plt.scatter(x = xi, y = yi, s = si * 10, marker = 'o')
plt.xticks(rotation = 90)
plt.xlabel('City')
plt.ylabel('Number of restaurantsr')
plt.legend(['weighted restaurant rating'])

plt.show()
```

