

Restaurants Project

September 5, 2022

SIMPLILEARN CAPSTONE PROJECT
MARKETING DOMAIN
IDENTIFYING AND RECOMMENDING BEST RESTAURANTS

```
[1]: #import all libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[2]: data=pd.read_excel('data.xlsx')
```

```
[3]: data
```

```
[3]:
```

	Restaurant ID	Restaurant Name	Country Code	City \
0	7402935	Skye	94	Jakarta
1	7410290	Satoo - Hotel Shangri-La	94	Jakarta
2	7420899	Sushi Masa	94	Jakarta
3	7421967	3 Wise Monkeys	94	Jakarta
4	7422489	Avec Moi Restaurant and Bar	94	Jakarta
...
9546	18279289	BMG - All Day Dining	1	Dehradun
9547	2300497	Atmosphere Grill Cafe Sheesha	1	Kanpur
9548	18312106	UrbanCrave	1	Kanpur
9549	3900245	Deena Chat Bhandar	1	Varanasi
9550	18246202	VNS Live Studio	1	Varanasi

```
Address \
```

0	Menara BCA, Lantai 56, Jl. MH. Thamrin, Thamri...
1	Hotel Shangri-La, Jl. Jend. Sudirman
2	Jl. Tuna Raya No. 5, Penjaringan
3	Jl. Suryo No. 26, Senopati, Jakarta
4	Gedung PIC, Jl. Teluk Betung 43, Thamrin, Jakarta
...	...
9546	140 A, Rajpur Road, Jakhan, Dehradun
9547	8th Floor, J.S. Tower, 16/106 - Mall Road, Kan...
9548	14/125, The Mall, Mall Road, Colonelganj, Para...
9549	D-47/184, Luxa Road, Dashaswmedh Road, Varanasi
9550	Hotel Varuna Ground Floor, 22 Gulab Bagh, Sigr...

	Locality	Locality Verbose \
0	Grand Indonesia Mall, Thamrin	Grand Indonesia Mall, Thamrin, Jakarta
1	Hotel Shangri-La, Sudirman	Hotel Shangri-La, Sudirman, Jakarta
2	Penjaringan	Penjaringan, Jakarta
3	Senopati	Senopati, Jakarta
4	Thamrin	Thamrin, Jakarta
...
9546	Jakhan	Jakhan, Dehradun
9547	Mall Road	Mall Road, Kanpur
9548	Parade	Parade, Kanpur
9549	Dashaswmedh Road	Dashaswmedh Road, Varanasi
9550	Sigra	Sigra, Varanasi

	Longitude	Latitude \
0	106.821999	-6.196778
1	106.818961	-6.203292
2	106.800144	-6.101298
3	106.813400	-6.235241
4	106.821023	-6.196270
...
9546	78.068890	30.362686
9547	80.354002	26.472001
9548	80.342796	26.474986
9549	0.000000	0.000000
9550	82.991694	25.318345

	Cuisines	Average Cost for two \
0	Italian, Continental	800000
1	Asian, Indonesian, Western	800000
2	Sushi, Japanese	500000
3	Japanese	450000
4	French, Western	350000
...
9546	Chinese, North Indian, Fast Food	0
9547	Indian, Chinese, Continental	0
9548	Cafe, Continental, Desserts, Ice Cream, Italia...	0
9549	Street Food	0
9550	Chinese, North Indian	0

	Currency	Has Table booking	Has Online delivery \
0	Indonesian Rupiah(IDR)	No	No
1	Indonesian Rupiah(IDR)	No	No
2	Indonesian Rupiah(IDR)	No	No
3	Indonesian Rupiah(IDR)	No	No
4	Indonesian Rupiah(IDR)	No	No
...

9546	Indian Rupees(Rs.)	No	No
9547	Indian Rupees(Rs.)	No	No
9548	Indian Rupees(Rs.)	No	No
9549	Indian Rupees(Rs.)	No	No
9550	Indian Rupees(Rs.)	No	No

	Price range	Aggregate rating	Rating color	Rating text	Votes
0	3	4.1	Green	Very Good	1498
1	3	4.6	Dark Green	Excellent	873
2	3	4.9	Dark Green	Excellent	605
3	3	4.2	Green	Very Good	395
4	3	4.3	Green	Very Good	243
...
9546	1	4.3	Green	Very Good	63
9547	1	3.6	Yellow	Good	34
9548	1	3.9	Yellow	Good	127
9549	1	3.8	Yellow	Good	78
9550	1	3.5	Yellow	Good	109

[9551 rows x 19 columns]

```
[4]: cc=pd.read_excel('Country-Code.xlsx')
cc
```

```
[4]:
```

	Country Code	Country
0	1	India
1	14	Australia
2	30	Brazil
3	37	Canada
4	94	Indonesia
5	148	New Zealand
6	162	Phillipines
7	166	Qatar
8	184	Singapore
9	189	South Africa
10	191	Sri Lanka
11	208	Turkey
12	214	UAE
13	215	United Kingdom
14	216	United States

```
[5]: data1=pd.merge(data,cc,on='Country Code', how='left')
data1
```

```
[5]:
```

	Restaurant ID	Restaurant Name	Country Code	City \
0	7402935	Skye	94	Jakarta
1	7410290	Satoo - Hotel Shangri-La	94	Jakarta

2	7420899	Sushi Masa	94	Jakarta
3	7421967	3 Wise Monkeys	94	Jakarta
4	7422489	Avec Moi Restaurant and Bar	94	Jakarta
...
9546	18279289	BMG - All Day Dining	1	Dehradun
9547	2300497	Atmosphere Grill Cafe Sheesha	1	Kanpur
9548	18312106	UrbanCrave	1	Kanpur
9549	3900245	Deena Chat Bhandar	1	Varanasi
9550	18246202	VNS Live Studio	1	Varanasi

	Address \
0	Menara BCA, Lantai 56, Jl. MH. Thamrin, Thamrin...
1	Hotel Shangri-La, Jl. Jend. Sudirman
2	Jl. Tuna Raya No. 5, Penjaringan
3	Jl. Suryo No. 26, Senopati, Jakarta
4	Gedung PIC, Jl. Teluk Betung 43, Thamrin, Jakarta
...	...
9546	140 A, Rajpur Road, Jakhan, Dehradun
9547	8th Floor, J.S. Tower, 16/106 - Mall Road, Kan...
9548	14/125, The Mall, Mall Road, Colonelganj, Para...
9549	D-47/184, Luxa Road, Dashaswmedh Road, Varanasi
9550	Hotel Varuna Ground Floor, 22 Gulab Bagh, Sigr...

	Locality	Locality Verbose \
0	Grand Indonesia Mall, Thamrin	Grand Indonesia Mall, Thamrin, Jakarta
1	Hotel Shangri-La, Sudirman	Hotel Shangri-La, Sudirman, Jakarta
2	Penjaringan	Penjaringan, Jakarta
3	Senopati	Senopati, Jakarta
4	Thamrin	Thamrin, Jakarta
...
9546	Jakhan	Jakhan, Dehradun
9547	Mall Road	Mall Road, Kanpur
9548	Parade	Parade, Kanpur
9549	Dashaswmedh Road	Dashaswmedh Road, Varanasi
9550	Sigra	Sigra, Varanasi

	Longitude	Latitude \
0	106.821999	-6.196778
1	106.818961	-6.203292
2	106.800144	-6.101298
3	106.813400	-6.235241
4	106.821023	-6.196270
...
9546	78.068890	30.362686
9547	80.354002	26.472001
9548	80.342796	26.474986
9549	0.000000	0.000000

9550 82.991694 25.318345

	Cuisines	Average Cost for two \
0	Italian, Continental	800000
1	Asian, Indonesian, Western	800000
2	Sushi, Japanese	500000
3	Japanese	450000
4	French, Western	350000
...
9546	Chinese, North Indian, Fast Food	0
9547	Indian, Chinese, Continental	0
9548	Cafe, Continental, Desserts, Ice Cream, Italia...	0
9549	Street Food	0
9550	Chinese, North Indian	0

	Currency	Has Table booking	Has Online delivery \
0	Indonesian Rupiah(IDR)	No	No
1	Indonesian Rupiah(IDR)	No	No
2	Indonesian Rupiah(IDR)	No	No
3	Indonesian Rupiah(IDR)	No	No
4	Indonesian Rupiah(IDR)	No	No
...
9546	Indian Rupees(Rs.)	No	No
9547	Indian Rupees(Rs.)	No	No
9548	Indian Rupees(Rs.)	No	No
9549	Indian Rupees(Rs.)	No	No
9550	Indian Rupees(Rs.)	No	No

	Price range	Aggregate rating	Rating	color	Rating text	Votes	Country
0	3	4.1	Green	Very Good	1498	Indonesia	
1	3	4.6	Dark Green	Excellent	873	Indonesia	
2	3	4.9	Dark Green	Excellent	605	Indonesia	
3	3	4.2	Green	Very Good	395	Indonesia	
4	3	4.3	Green	Very Good	243	Indonesia	
...	
9546	1	4.3	Green	Very Good	63	India	
9547	1	3.6	Yellow	Good	34	India	
9548	1	3.9	Yellow	Good	127	India	
9549	1	3.8	Yellow	Good	78	India	
9550	1	3.5	Yellow	Good	109	India	

[9551 rows x 20 columns]

```
[6]: data1.head()
```

[6]:	Restaurant ID	Restaurant Name	Country Code	City \
0	7402935	Skye	94	Jakarta

1	7410290	Satoo - Hotel Shangri-La	94	Jakarta
2	7420899	Sushi Masa	94	Jakarta
3	7421967	3 Wise Monkeys	94	Jakarta
4	7422489	Avec Moi Restaurant and Bar	94	Jakarta

	Address \
0	Menara BCA, Lantai 56, Jl. MH. Thamrin, Thamri...
1	Hotel Shangri-La, Jl. Jend. Sudirman
2	Jl. Tuna Raya No. 5, Penjaringan
3	Jl. Suryo No. 26, Senopati, Jakarta
4	Gedung PIC, Jl. Teluk Betung 43, Thamrin, Jakarta

	Locality	Locality Verbose \
0	Grand Indonesia Mall, Thamrin	Grand Indonesia Mall, Thamrin, Jakarta
1	Hotel Shangri-La, Sudirman	Hotel Shangri-La, Sudirman, Jakarta
2	Penjaringan	Penjaringan, Jakarta
3	Senopati	Senopati, Jakarta
4	Thamrin	Thamrin, Jakarta

	Longitude	Latitude	Cuisines	Average Cost for two \
0	106.821999	-6.196778	Italian, Continental	800000
1	106.818961	-6.203292	Asian, Indonesian, Western	800000
2	106.800144	-6.101298	Sushi, Japanese	500000
3	106.813400	-6.235241	Japanese	450000
4	106.821023	-6.196270	French, Western	350000

	Currency	Has Table booking	Has Online delivery	Price range \
0	Indonesian Rupiah(IDR)	No	No	3
1	Indonesian Rupiah(IDR)	No	No	3
2	Indonesian Rupiah(IDR)	No	No	3
3	Indonesian Rupiah(IDR)	No	No	3
4	Indonesian Rupiah(IDR)	No	No	3

	Aggregate rating	Rating color	Rating text	Votes	Country
0	4.1	Green	Very Good	1498	Indonesia
1	4.6	Dark Green	Excellent	873	Indonesia
2	4.9	Dark Green	Excellent	605	Indonesia
3	4.2	Green	Very Good	395	Indonesia
4	4.3	Green	Very Good	243	Indonesia

```
[7]: data1.columns=data1.columns.str.replace(' ','_')
data1.columns
```

```
[7]: 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', 'Price_range', 'Aggregate_rating',
```

```
    'Rating_color', 'Rating_text', 'Votes', 'Country'],
    dtype='object')
```

```
[8]: data1.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 9551 entries, 0 to 9550
Data columns (total 20 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Restaurant_ID          9551 non-null   int64
1   Restaurant_Name         9550 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  Price_range            9551 non-null   int64
15  Aggregate_rating       9551 non-null   float64
16  Rating_color           9551 non-null   object
17  Rating_text            9551 non-null   object
18  Votes                  9551 non-null   int64
19  Country                9551 non-null   object
dtypes: float64(3), int64(5), object(12)
memory usage: 1.5+ MB
```

```
[9]: data1.describe
```

```
[9]: <bound method NDFrame.describe of Restaurant_ID
Restaurant_Name Country_Code City \
0          7402935                Skye          94  Jakarta
1          7410290      Satoo - Hotel Shangri-La    94  Jakarta
2          7420899                Sushi Masa    94  Jakarta
3          7421967          3 Wise Monkeys    94  Jakarta
4          7422489  Avec Moi Restaurant and Bar    94  Jakarta
...
9546      18279289      BMG - All Day Dining          1  Dehradun
9547      2300497  Atmosphere Grill Cafe Sheesha    1    Kanpur
9548      18312106                UrbanCrave    1    Kanpur
9549      3900245      Deena Chat Bhandar    1  Varanasi
```

9550	18246202	VNS Live Studio	1	Varanasi
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	Address \
0	Menara BCA, Lantai 56, Jl. MH. Thamrin, Thamri...
1	Hotel Shangri-La, Jl. Jend. Sudirman
2	Jl. Tuna Raya No. 5, Penjaringan
3	Jl. Suryo No. 26, Senopati, Jakarta
4	Gedung PIC, Jl. Teluk Betung 43, Thamrin, Jakarta
...	...
9546	140 A, Rajpur Road, Jakhan, Dehradun
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9548	14/125, The Mall, Mall Road, Colonelganj, Para...
9549	D-47/184, Luxa Road, Dashaswmedh Road, Varanasi
9550	Hotel Varuna Ground Floor, 22 Gulab Bagh, Sigr...

	Locality	Locality_Verbose \
0	Grand Indonesia Mall, Thamrin	Grand Indonesia Mall, Thamrin, Jakarta
1	Hotel Shangri-La, Sudirman	Hotel Shangri-La, Sudirman, Jakarta
2	Penjaringan	Penjaringan, Jakarta
3	Senopati	Senopati, Jakarta
4	Thamrin	Thamrin, Jakarta
...
9546	Jakhan	Jakhan, Dehradun
9547	Mall Road	Mall Road, Kanpur
9548	Parade	Parade, Kanpur
9549	Dashaswmedh Road	Dashaswmedh Road, Varanasi
9550	Sigra	Sigra, Varanasi

	Longitude	Latitude \
0	106.821999	-6.196778
1	106.818961	-6.203292
2	106.800144	-6.101298
3	106.813400	-6.235241
4	106.821023	-6.196270
...
9546	78.068890	30.362686
9547	80.354002	26.472001
9548	80.342796	26.474986
9549	0.000000	0.000000
9550	82.991694	25.318345

	Cuisines	Average_Cost_for_two \
0	Italian, Continental	800000
1	Asian, Indonesian, Western	800000
2	Sushi, Japanese	500000
3	Japanese	450000
4	French, Western	350000


```

...
9546          Chinese, North Indian, Fast Food          0
9547          Indian, Chinese, Continental          0
9548 Cafe, Continental, Desserts, Ice Cream, Italia... 0
9549          Street Food          0
9550          Chinese, North Indian          0

```

```

          Currency Has_Table_booking Has_Online_delivery \
0   Indonesian Rupiah(IDR)          No          No
1   Indonesian Rupiah(IDR)          No          No
2   Indonesian Rupiah(IDR)          No          No
3   Indonesian Rupiah(IDR)          No          No
4   Indonesian Rupiah(IDR)          No          No
...
9546   Indian Rupees(Rs.)          No          No
9547   Indian Rupees(Rs.)          No          No
9548   Indian Rupees(Rs.)          No          No
9549   Indian Rupees(Rs.)          No          No
9550   Indian Rupees(Rs.)          No          No

```

```

          Price_range Aggregate_rating Rating_color Rating_text Votes Country
0          3          4.1      Green  Very Good  1498  Indonesia
1          3          4.6  Dark Green  Excellent   873  Indonesia
2          3          4.9  Dark Green  Excellent   605  Indonesia
3          3          4.2      Green  Very Good   395  Indonesia
4          3          4.3      Green  Very Good   243  Indonesia
...
9546          1          4.3      Green  Very Good    63    India
9547          1          3.6    Yellow    Good    34    India
9548          1          3.9    Yellow    Good   127    India
9549          1          3.8    Yellow    Good    78    India
9550          1          3.5    Yellow    Good   109    India

```

[9551 rows x 20 columns]>

```
[10]: data1.shape
```

```
[10]: (9551, 20)
```

```
[11]: data1.dtypes
```

```

[11]: 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
Price_range           int64
Aggregate_rating      float64
Rating_color          object
Rating_text           object
Votes                int64
Country               object
dtype: object

```

```

[12]: #Data Cleaning
      #Removing null values
      data1.isnull().sum()

```

```

[12]: Restaurant_ID      0
      Restaurant_Name    1
      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
      Price_range        0
      Aggregate_rating    0
      Rating_color       0
      Rating_text        0
      Votes              0
      Country            0
      dtype: int64

```

```

[13]: data.isnull()

```

```

[13]:
      Restaurant ID  Restaurant Name  Country Code  City  Address  Locality \
0                False             False         False  False  False  False
1                False             False         False  False  False  False

```

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

	Locality	Verbose	Longitude	Latitude	Cuisines	Average Cost for two	\
0		False	False	False	False		False
1		False	False	False	False		False
2		False	False	False	False		False
3		False	False	False	False		False
4		False	False	False	False		False
...	
9546		False	False	False	False		False
9547		False	False	False	False		False
9548		False	False	False	False		False
9549		False	False	False	False		False
9550		False	False	False	False		False

	Currency	Has Table booking	Has Online delivery	Price range	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	
...
9546	False	False	False	False	False
9547	False	False	False	False	False
9548	False	False	False	False	False
9549	False	False	False	False	False
9550	False	False	False	False	False

	Aggregate rating	Rating color	Rating text	Votes
0	False	False	False	False
1	False	False	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False
...
9546	False	False	False	False
9547	False	False	False	False
9548	False	False	False	False
9549	False	False	False	False

9550 False False False False

[9551 rows x 19 columns]

```
[14]: #Restaurant name is missing, drop the name and reset the index
data1.dropna(axis=0, subset=['Restaurant_Name'], inplace=True)
data1.reset_index(drop=True, inplace=True)
```

```
[15]: data1[data1['Cuisines'].isnull()]
```

```
[15]:
```

	Restaurant_ID	Restaurant_Name	Country_Code	\
9082	17374552	Corkscrew Cafe	216	
9085	17501439	Dovetail	216	
9093	17059060	Hillstone	216	
9405	17284158	Jimmie's Hot Dogs	216	
9493	17142698	Leonard's Bakery	216	
9503	17616465	Tybee Island Social Club	216	
9532	17284105	Cookie Shoppe	216	
9534	17284211	Pearly's Famous Country Cookng	216	
9538	17606621	HI Lite Bar & Lounge	216	

	City	Address	\
9082	Gainesville	51 W Main St, Dahlonge, GA 30533	
9085	Macon	543 Cherry St, Macon, GA 31201	
9093	Orlando	215 South Orlando Avenue, Winter Park, FL 32789	
9405	Albany	204 S Jackson St, Albany, GA 31701	
9493	Rest of Hawaii	933 Kapahulu Ave, Honolulu, HI 96816	
9503	Savannah	1311 Butler Ave, Tybee Island, GA 31328	
9532	Albany	115 N Jackson St, Albany, GA 31701	
9534	Albany	814 N Slappey Blvd, Albany, GA 31701	
9538	Miller	109 N Broadway Ave, Miller, SD 57362	

	Locality	Locality_Verbose	Longitude	Latitude	Cuisines	\
9082	Dahlonge	Dahlonge, Gainesville	-83.985800	34.531800	NaN	
9085	Macon	Macon, Macon	-83.627979	32.836410	NaN	
9093	Winter Park	Winter Park, Orlando	-81.365260	28.596682	NaN	
9405	Albany	Albany, Albany	-84.153400	31.575100	NaN	
9493	Kaimuki	Kaimuki, Rest of Hawaii	-157.813432	21.284586	NaN	
9503	Tybee Island	Tybee Island, Savannah	-80.848297	31.995810	NaN	
9532	Albany	Albany, Albany	-84.154000	31.577200	NaN	
9534	Albany	Albany, Albany	-84.175900	31.588200	NaN	
9538	Miller	Miller, Miller	-98.989100	44.515800	NaN	

	Average_Cost_for_two	Currency	Has_Table_booking	Has_Online_delivery	\
9082	40	Dollar(\$)	No	No	
9085	40	Dollar(\$)	No	No	
9093	40	Dollar(\$)	No	No	

9405	10	Dollar(\$)	No	No
9493	10	Dollar(\$)	No	No
9503	10	Dollar(\$)	No	No
9532	0	Dollar(\$)	No	No
9534	0	Dollar(\$)	No	No
9538	0	Dollar(\$)	No	No

	Price_range	Aggregate_rating	Rating_color	Rating_text	Votes	\
9082	3	3.9	Yellow	Good	209	
9085	3	3.8	Yellow	Good	102	
9093	3	4.4	Green	Very Good	1158	
9405	1	3.9	Yellow	Good	160	
9493	1	4.7	Dark Green	Excellent	707	
9503	1	3.9	Yellow	Good	309	
9532	1	3.4	Orange	Average	34	
9534	1	3.4	Orange	Average	36	
9538	1	3.4	Orange	Average	11	

	Country
9082	United States
9085	United States
9093	United States
9405	United States
9493	United States
9503	United States
9532	United States
9534	United States
9538	United States

```
[16]: #Only 9 recordds without Cuisines, so replace the null values with Others
data1['Cuisines'].fillna('Others',inplace=True)
```

```
[17]: data1.isnull().sum()
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9551 entries, 0 to 9550
Data columns (total 19 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Restaurant ID          9551 non-null   int64
1   Restaurant Name        9550 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  Price range          9551 non-null   int64
15  Aggregate rating     9551 non-null   float64
16  Rating color         9551 non-null   object
17  Rating text          9551 non-null   object
18  Votes               9551 non-null   int64
dtypes: float64(3), int64(5), object(11)
memory usage: 1.4+ MB

```

Performing Exploratory Data Analysis

```

[18]: #Explore the geographical distribution of the restaurants
      #Identify the cities with the maximum and minimum number of restaurants
      #using groupby we can further proceed
      cntry_dist=data1.groupby(['Country_Code','Country']).
      ↪agg(Count=('Restaurant_ID','count'))
      cntry_dist.sort_values(by='Count',ascending=False)

```

```

[18]:
Country_Code Country      Count
1           India      8651
216         United States    434
215         United Kingdom    80
30          Brazil         60
189         South Africa    60
214          UAE          60
148         New Zealand    40
208          Turkey        34
14          Australia       24
162         Phillipines     22
94          Indonesia       21
166          Qatar         20
184          Singapore      20
191          Sri Lanka       20
37          Canada         4

```

The above result shows that India has maximum number of restaurants whereas Canada has the minimum.

```

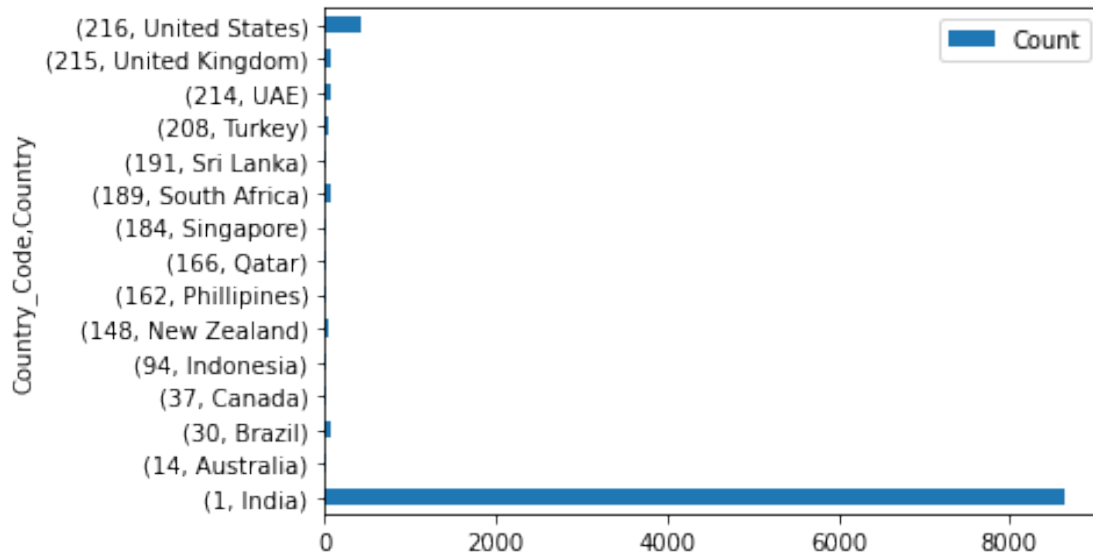
[19]: cntry_dist.plot(kind='barh')

```

```

[19]: <AxesSubplot:ylabel='Country_Code,Country'>

```



```
[20]: #City with max and min restaurant count
city_dist=data1.groupby(['Country','City']).agg(Count=('Restaurant_ID','count'))
city_dist.describe()
```

```
[20]:          Count
count    141.000000
mean      67.730496
std       476.723952
min        1.000000
25%        1.000000
50%       20.000000
75%       20.000000
max      5473.000000
```

The above analysis show that the max value of restaurant in a city is 5473 Whereas the minimum value is 1

```
[21]: #Sorting the Cities with most number of restaurants
city_dist.sort_values(by='Count',ascending=False)
```

```
[21]:          Country      City      Count
India      New Delhi    5473
           Gurgaon     1118
           Noida       1080
           Faridabad    251
           Ghaziabad     25
...
           Panchkula      1
```

Australia	Balingup	1
Indonesia	Bandung	1
Phillipines	Quezon City	1
United States	Winchester Bay	1

[141 rows x 1 columns]

The above analysis shows that New Delhi has maximum number of restaurants

[22]: *#Number of countries with minimum restaurants*

```
min_rest=city_dist[city_dist['Count']==1]
min_rest.info()
min_rest
```

```
<class 'pandas.core.frame.DataFrame'>
MultiIndex: 46 entries, ('Australia', 'Armidale') to ('United States',
'Winchester Bay')
Data columns (total 1 columns):
#   Column  Non-Null Count  Dtype
---  -
0    Count   46 non-null       int64
dtypes: int64(1)
memory usage: 1.8+ KB
```

[22]:

Country	City	Count
Australia	Armidale	1
	Balingup	1
	Beechworth	1
	Dicky Beach	1
	East Ballina	1
	Flaxton	1
	Forrest	1
	Huskisson	1
	Inverloch	1
	Lakes Entrance	1
	Lorn	1
	Macedon	1
	Mayfield	1
	Middleton Beach	1
	Montville	1
	Palm Cove	1
	Paynesville	1
	Penola	1
	Phillip Island	1
	Tanunda	1
	Trentham East	1

	Victor Harbor	1
Canada	Chatham-Kent	1
	Consort	1
	Vineland Station	1
	Yorkton	1
India	Mohali	1
	Panchkula	1
Indonesia	Bandung	1
Phillipines	Quezon City	1
	Tagaytay City	1
South Africa	Randburg	1
United States	Clatskanie	1
	Cochrane	1
	Fernley	1
	Lakeview	1
	Lincoln	1
	Mc Millan	1
	Miller	1
	Monroe	1
	Ojo Caliente	1
	Potrero	1
	Princeton	1
	Vernonia	1
	Weirton	1
	Winchester Bay	1

The above data shows that there are 46 cities in 7 different countries with only 1 restaurant

```
[23]: #How ratings distributed overall and highest and lowest rated restaurants
max_rate=data1.sort_values(by='Aggregate_rating',ascending=False).
    ↳groupby(['Country','City'],as_index=False).first()
#highest rating restaurants
min_rate=data1.sort_values(by='Aggregate_rating',ascending=False).
    ↳groupby(['Country','City'],as_index=False).last()
#lowest rating restaurants
```

```
[24]: df_max=max_rate[['Country','City','Restaurant_Name','Aggregate_rating']]#dataframe
    ↳created for highrated restaurants
df_min=min_rate[['Country','City','Restaurant_Name','Aggregate_rating']]#dataframe
    ↳created for lowrated restaurants
rest_rating=df_max.
    ↳merge(df_min,left_on='City',right_on='City',how='inner')#Merging of both
    ↳dataframes
```

```
[25]: rest_rating
```

```
[25]:
```

	Country_x	City	Restaurant_Name_x \
0	Australia	Armidale	Whitebull Hotel
1	Australia	Balingup	Taste of Balingup
2	Australia	Beechworth	Bridge Road Brewers
3	Australia	Dicky Beach	The Giggling Goat
4	Australia	East Ballina	The Belle General
..
136	United States	Valdosta	Smok'n Pig B-B-Q
137	United States	Vernonia	Blue House Cafe
138	United States	Waterloo	Four Queens Dairy Cream
139	United States	Weirton	Theo Yianni's Authentic Greek Restaurant
140	United States	Winchester Bay	Fishpatrick's Crabby Cafe

	Aggregate_rating_x	Country_y \
0	3.5	Australia
1	3.2	Australia
2	4.6	Australia
3	3.6	Australia
4	4.1	Australia
..
136	4.1	United States
137	4.3	United States
138	3.9	United States
139	3.9	United States
140	3.2	United States

	Restaurant_Name_y	Aggregate_rating_y
0	Whitebull Hotel	3.5
1	Taste of Balingup	3.2
2	Bridge Road Brewers	4.6
3	The Giggling Goat	3.6
4	The Belle General	4.1
..
136	El Tereo Mexican Restaurant	3.1
137	Blue House Cafe	4.3
138	Masala Grill & Coffee House	3.2
139	Theo Yianni's Authentic Greek Restaurant	3.9
140	Fishpatrick's Crabby Cafe	3.2

[141 rows x 7 columns]

```
[26]: rest_rating.drop(columns='Country_y',axis=1,inplace=True)
rest_rating.columns=['Country','City','Highest Rated Restaurant','Rating_
↳Max','Lowest Rated Restaurant','Rating Min']
rest_rating
```

```
[26]:
```

	Country	City	Highest Rated Restaurant \
0	Australia	Armidale	Whitebull Hotel
1	Australia	Balingup	Taste of Balingup
2	Australia	Beechworth	Bridge Road Brewers
3	Australia	Dicky Beach	The Giggling Goat
4	Australia	East Ballina	The Belle General
..
136	United States	Valdosta	Smok'n Pig B-B-Q
137	United States	Vernonia	Blue House Cafe
138	United States	Waterloo	Four Queens Dairy Cream
139	United States	Weirton	Theo Yianni's Authentic Greek Restaurant
140	United States	Winchester Bay	Fishpatrick's Crabby Cafe

	Rating Max	Lowest Rated Restaurant	Rating Min
0	3.5	Whitebull Hotel	3.5
1	3.2	Taste of Balingup	3.2
2	4.6	Bridge Road Brewers	4.6
3	3.6	The Giggling Goat	3.6
4	4.1	The Belle General	4.1
..
136	4.1	El Toreo Mexican Restaurant	3.1
137	4.3	Blue House Cafe	4.3
138	3.9	Masala Grill & Coffee House	3.2
139	3.9	Theo Yianni's Authentic Greek Restaurant	3.9
140	3.2	Fishpatrick's Crabby Cafe	3.2

[141 rows x 6 columns]

```
[27]: data_rest=data1.copy()
data_rest.columns
```

```
[27]: 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', 'Price_range', 'Aggregate_rating',
        'Rating_color', 'Rating_text', 'Votes', 'Country'],
        dtype='object')
```

```
[28]: #The ratio between restaurants that allow table booking vs that do not allow
        ↳table booking?
        #What is the percentage of restaurants providing online delivery?
        #Is there a difference in no. of votes for the restaurants that deliver and the
        ↳restaurant that don't?

dummy=['Has_Table_booking', 'Has_Online_delivery']
```

```
[29]: data_rest=pd.get_dummies(data_rest,columns=dummy,drop_first=True)
data_rest.head()
```

```
[29]: Restaurant_ID      Restaurant_Name  Country_Code    City \
0      7402935                Skye                94  Jakarta
1      7410290      Satoo - Hotel Shangri-La        94  Jakarta
2      7420899                Sushi Masa        94  Jakarta
3      7421967          3 Wise Monkeys        94  Jakarta
4      7422489  Avec Moi Restaurant and Bar        94  Jakarta

                Address \
0  Menara BCA, Lantai 56, Jl. MH. Thamrin, Thamri...
1      Hotel Shangri-La, Jl. Jend. Sudirman
2      Jl. Tuna Raya No. 5, Penjaringan
3      Jl. Suryo No. 26, Senopati, Jakarta
4  Gedung PIC, Jl. Teluk Betung 43, Thamrin, Jakarta

                Locality      Locality_Verbose \
0  Grand Indonesia Mall, Thamrin  Grand Indonesia Mall, Thamrin, Jakarta
1      Hotel Shangri-La, Sudirman  Hotel Shangri-La, Sudirman, Jakarta
2      Penjaringan                Penjaringan, Jakarta
3      Senopati                  Senopati, Jakarta
4      Thamrin                  Thamrin, Jakarta

    Longitude  Latitude      Cuisines  Average_Cost_for_two \
0  106.821999 -6.196778  Italian, Continental      800000
1  106.818961 -6.203292  Asian, Indonesian, Western      800000
2  106.800144 -6.101298      Sushi, Japanese      500000
3  106.813400 -6.235241      Japanese      450000
4  106.821023 -6.196270  French, Western      350000

                Currency  Price_range  Aggregate_rating  Rating_color \
0  Indonesian Rupiah(IDR)          3          4.1      Green
1  Indonesian Rupiah(IDR)          3          4.6  Dark Green
2  Indonesian Rupiah(IDR)          3          4.9  Dark Green
3  Indonesian Rupiah(IDR)          3          4.2      Green
4  Indonesian Rupiah(IDR)          3          4.3      Green

    Rating_text  Votes  Country  Has_Table_booking_Yes \
0   Very Good   1498  Indonesia          0
1   Excellent    873  Indonesia          0
2   Excellent    605  Indonesia          0
3   Very Good    395  Indonesia          0
4   Very Good    243  Indonesia          0

    Has_Online_delivery_Yes
0                          0
```

1	0
2	0
3	0
4	0

In the above table 0 indicates No and 1 indicates Yes

```
[30]: #Ratio between restaurants that allow table booking vs that do not allow table
      ↳ booking.
      tab_book=data_rest[data_rest['Has_Table_booking_Yes']==1]['Restaurant_ID'].
      ↳ count()
      notab_book =data_rest[data_rest['Has_Table_booking_Yes']==0]['Restaurant_ID'].
      ↳ count()
      print('Ratio between restaurants that allow table booking vs. those that do not
      ↳ allow table booking: ',
            round((tab_book/notab_book),2))
```

Ratio between restaurants that allow table booking vs. those that do not allow table booking: 0.14

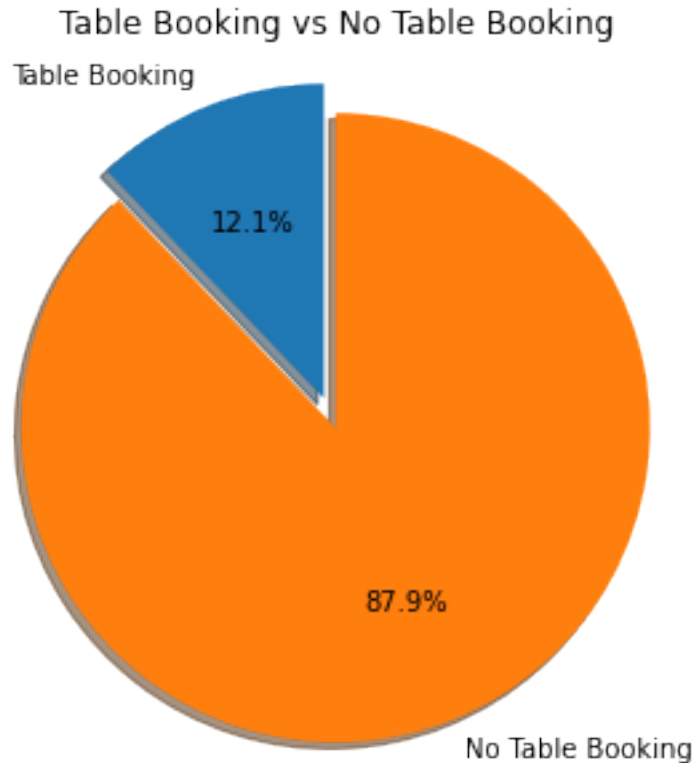
```
[31]: print(tab_book,notab_book)
```

1158 8392

```
[32]: #Pie chart to show percentage of restaurants which allow table booking and
      ↳ those which don't
      labels = 'Table Booking', 'No Table Booking'
      sizes = [tab_book,notab_book]
      explode = (0.1, 0) # only "explode" the 2nd slice (i.e. 'Hogs')

      fig1, ax1 = plt.subplots(figsize=(5,5))
      ax1.pie(sizes, explode=explode, labels=labels, autopct='%1.1f%%',shadow=True,
      ↳ startangle=90)
      ax1.set_title("Table Booking vs No Table Booking")
      ax1.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle.

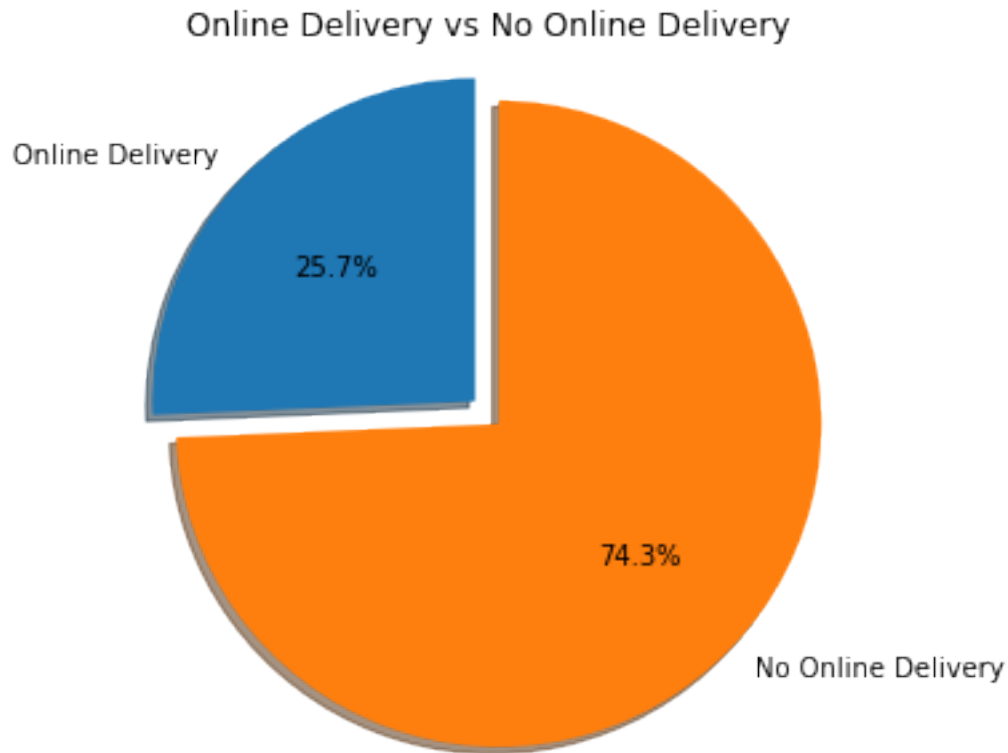
      plt.show()
```



```
[33]: #Percentage of restaurant that has online delivery
ol_dlvry = data_rest[data_rest['Has_Online_delivery_Yes'] == 1]['Restaurant_ID'].count()
no_ol_dlvry = data_rest[data_rest['Has_Online_delivery_Yes'] == 0]['Restaurant_ID'].count()
print('Percentage of restaurants providing online delivery : {} %'.
      format((round(ol_dlvry/len(data_rest),3)*100)))
```

Percentage of restaurants providing online delivery : 25.7 %

```
[34]: #pie chart to show percentages of restaurants allowing online delivery vs those
      which do not have online delivery
labels = 'Online Delivery','No Online Delivery'
size = [ol_dlvry,no_ol_dlvry]
explode = (0.1,0)
fig1,ax1 = plt.subplots(figsize=(5,5))
ax1.pie(size,explode=explode,labels=labels,autopct='%1.1f%%',shadow=True,startangle=90)
ax1.set_title("Online Delivery vs No Online Delivery")
ax1.axis('equal')
plt.show()
```

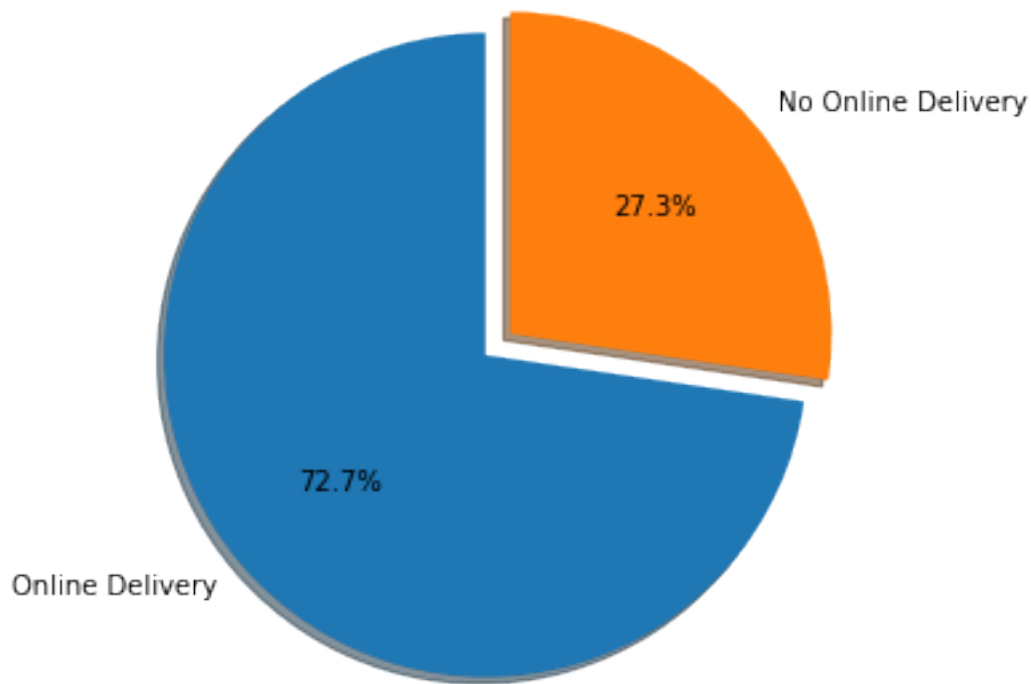


```
[35]: #Difference in no. of votes for the restaurants that deliver and the restaurant
      ↳ that don't?
rest_del = data_rest[data_rest['Has_Table_booking_Yes'] == 1]['Votes'].sum()
rest_ndel = data_rest[data_rest['Has_Table_booking_Yes'] == 0]['Votes'].sum()
print('Difference in number of votes for restaurants that deliver and dont_
      ↳ deliver: ',abs((rest_del - rest_ndel)))
```

Difference in number of votes for restaurants that deliver and dont deliver:
680082

```
[36]: labels = 'Online Delivery','No Online Delivery'
size = [rest_ndel,rest_del]
explode = (0,0.1)
fig1,ax1 = plt.subplots(figsize=(5,5))
ax1.pie(size,explode=explode,labels=labels,autopct='%1.
      ↳ 1f%%',shadow=True,startangle=90)
ax1.set_title("Votes: Online Delivery vs Votes:No Online Delivery")
ax1.axis('equal')
plt.show()
```

Votes: Online Delivery vs Votes:No Online Delivery



Out of the total votes about 27.3% votes were given to restaurants that dont have online delivery option Out of the total votes about 72.7% votes were given to restaurants that dont have online delivery option This clearly shows that restaurants that have online delivery are more likely to get a vote(feedback)

```
[37]: #What are the top 10 cuisines served across cities?
      #What is the maximum and minimum no. of cuisines that a restaurant serves?
```

```
[38]: data1.columns
      cuisines = data1['Cuisines'].apply(lambda x: pd.Series(x.split(',')))
```

```
[39]: cuisines.columns =
      ↳ ['Cuisine_1', 'Cuisine_2', 'Cuisine_3', 'Cuisine_4', 'Cuisine_5', 'Cuisine_6', 'Cuisine_7', 'Cuisine_8']
      cuisines.tail()
```

```
[39]:
```

	Cuisine_1	Cuisine_2	Cuisine_3	Cuisine_4	Cuisine_5	\
9545	Chinese	North Indian	Fast Food	NaN	NaN	
9546	Indian	Chinese	Continental	NaN	NaN	
9547	Cafe	Continental	Desserts	Ice Cream	Italian	
9548	Street Food	NaN	NaN	NaN	NaN	
9549	Chinese	North Indian	NaN	NaN	NaN	

	Cuisine_6	Cuisine_7	Cuisine_8
9545	NaN	NaN	NaN
9546	NaN	NaN	NaN
9547	Beverages	NaN	NaN
9548	NaN	NaN	NaN
9549	NaN	NaN	NaN

```
[40]: data_cuisines = pd.concat([data1,cuisines],axis=1)
data_cuisines.head()
```

```
[40]: Restaurant_ID      Restaurant_Name  Country_Code  City \
0      7402935                Skye           94  Jakarta
1      7410290      Satoo - Hotel Shangri-La       94  Jakarta
2      7420899                Sushi Masa       94  Jakarta
3      7421967            3 Wise Monkeys       94  Jakarta
4      7422489  Avec Moi Restaurant and Bar       94  Jakarta
```

	Address \
0	Menara BCA, Lantai 56, Jl. MH. Thamrin, Thamri...
1	Hotel Shangri-La, Jl. Jend. Sudirman
2	Jl. Tuna Raya No. 5, Penjaringan
3	Jl. Suryo No. 26, Senopati, Jakarta
4	Gedung PIC, Jl. Teluk Betung 43, Thamrin, Jakarta

	Locality	Locality_Verbose \
0	Grand Indonesia Mall, Thamrin	Grand Indonesia Mall, Thamrin, Jakarta
1	Hotel Shangri-La, Sudirman	Hotel Shangri-La, Sudirman, Jakarta
2	Penjaringan	Penjaringan, Jakarta
3	Senopati	Senopati, Jakarta
4	Thamrin	Thamrin, Jakarta

	Longitude	Latitude	Cuisines ...	Votes	Country \
0	106.821999	-6.196778	Italian, Continental ...	1498	Indonesia
1	106.818961	-6.203292	Asian, Indonesian, Western ...	873	Indonesia
2	106.800144	-6.101298	Sushi, Japanese ...	605	Indonesia
3	106.813400	-6.235241	Japanese ...	395	Indonesia
4	106.821023	-6.196270	French, Western ...	243	Indonesia

	Cuisine_1	Cuisine_2	Cuisine_3	Cuisine_4	Cuisine_5	Cuisine_6 \
0	Italian	Continental	NaN	NaN	NaN	NaN
1	Asian	Indonesian	Western	NaN	NaN	NaN
2	Sushi	Japanese	NaN	NaN	NaN	NaN
3	Japanese	NaN	NaN	NaN	NaN	NaN
4	French	Western	NaN	NaN	NaN	NaN

	Cuisine_7	Cuisine_8
0	NaN	NaN

```

1      NaN      NaN
2      NaN      NaN
3      NaN      NaN
4      NaN      NaN

```

[5 rows x 28 columns]

```

[41]: cuisine_loc = pd.
      ↪DataFrame(data_cuisines[['Country','City','Locality_Verbose','Cuisine_1','Cuisine_2','Cuisine_3',
      ↪                               ↵
      ↪'Cuisine_4','Cuisine_5','Cuisine_6','Cuisine_7','Cuisine_8']])

```

```

[42]: cuisine_loc_stack=pd.DataFrame(cuisine_loc.stack()) #stacking the columns
      cuisine_loc.head()

```

```

[42]:      Country      City      Locality_Verbose Cuisine_1 \
0  Indonesia  Jakarta  Grand Indonesia Mall, Thamrin, Jakarta  Italian
1  Indonesia  Jakarta      Hotel Shangri-La, Sudirman, Jakarta  Asian
2  Indonesia  Jakarta      Penjaringan, Jakarta  Sushi
3  Indonesia  Jakarta      Senopati, Jakarta  Japanese
4  Indonesia  Jakarta      Thamrin, Jakarta  French

      Cuisine_2 Cuisine_3 Cuisine_4 Cuisine_5 Cuisine_6 Cuisine_7 Cuisine_8
0  Continental      NaN      NaN      NaN      NaN      NaN      NaN
1  Indonesian  Western      NaN      NaN      NaN      NaN      NaN
2  Japanese      NaN      NaN      NaN      NaN      NaN      NaN
3      NaN      NaN      NaN      NaN      NaN      NaN      NaN
4  Western      NaN      NaN      NaN      NaN      NaN      NaN

```

```

[43]: keys = [c for c in cuisine_loc if c.startswith('Cuisine')]
      a=pd.melt(cuisine_loc, id_vars='Locality_Verbose', value_vars=keys,
      ↪value_name='Cuisines')
      #melting the stack into one row
      max_rate=pd.DataFrame(a.groupby(by=['Locality_Verbose','variable','Cuisines']).
      ↪size().reset_index())
      #find the highest restaurant in the city
      max_rate
      del max_rate['variable']
      max_rate.columns=['Locality_Verbose','Cuisines','Count']
      max_rate.head()

```

```

[43]:      Locality_Verbose      Cuisines  Count
0  ILD Trade Centre Mall, Sohna Road, Gurgaon      Cafe      1
1  ILD Trade Centre Mall, Sohna Road, Gurgaon  North Indian      1
2  ILD Trade Centre Mall, Sohna Road, Gurgaon  Beverages      1
3  ILD Trade Centre Mall, Sohna Road, Gurgaon      Mughlai      1
4  12th Square Building, Banjara Hills, Hyderabad      Mughlai      1

```

```
[44]: #find the highest restuarant in the city
loc=max_rate.sort_values('Count', ascending=False).
      ↳groupby(by=['Locality_Verbose'],as_index=False).first()
loc.head()
```

```
[44]:
```

	Locality_Verbose	Cuisines	Count
0	ILD Trade Centre Mall, Sohna Road, Gurgaon	Cafe	1
1	12th Square Building, Banjara Hills, Hyderabad	Mughlai	1
2	A Hotel, Gurdev Nagar, Ludhiana	Chinese	1
3	ARSS Mall, Paschim Vihar, New Delhi	North Indian	1
4	Aaya Nagar, New Delhi	Cuisine Varies	1

```
[45]: rating_res=loc.
      ↳merge(data1,left_on='Locality_Verbose',right_on='Locality_Verbose',how='inner')
      #inner join to merge the two dataframe
df=pd.
      ↳DataFrame(rating_res[['Country','City','Locality_Verbose','Cuisines_x','Count']])
      #making a dataframe of rating restaurant
country=rating_res.sort_values('Count', ascending=False).
      ↳groupby(by=['Country'],as_index=False).first()
      #grouping the data by country code
con=pd.DataFrame(country[['Country','City','Locality','Cuisines_x','Count']])
con.columns=['Country','City','Locality','Cuisines','Number of restaurants in_
      ↳the country']
      #renaming the columns
con1=con.sort_values('Number of restaurants in the country', ascending=False)
      #sorting the restaurants on the basis of the number of restaurants in the_
      ↳country
con1[:10]
final_con=con1.drop(con1.index[[7,10]])
```

```
[46]: final_con
```

```
[46]:
```

	Country	City \
3	India	New Delhi
14	United States	Dubuque
5	New Zealand	Wellington City
1	Brazil	Rio de Janeiro
6	Phillipines	Mandaluyong City
8	Singapore	Singapore
9	South Africa	Cape Town
11	Turkey	Ankara
12	UAE	Abu Dhabi
0	Australia	Victor Harbor
2	Canada	Vineland Station
4	Indonesia	Jakarta
7	Qatar	Doha

	Locality	Cuisines \
3	Connaught Place	North Indian
14	Dubuque	American
5	Te Aro	Cafe
1	Ipanema	Brazilian
6	SM Megamall, Ortigas, Mandaluyong City	Japanese
8	Marina Centre, Downtown Core	Seafood
9	Green Point	Grill
11	Gazi Osman PaÅŪa	World Cuisine
12	Abu Dhabi Mall, Tourist Club Area (Al Zahiyah)	American
0	Victor Harbor	Coffee and Tea
2	Vineland Station	Italian
4	Tebet	Western
7	The Westin Doha Hotel & Spa, Fereej Bin Mahmoud	Thai

	Number of restaurants in the country
3	48
14	9
5	5
1	3
6	2
8	2
9	2
11	2
12	2
0	1
2	1
4	1
7	1

```
[47]: loc_list=final_con['City'] #converting the series to dataframe
a_list=loc_list.tolist()

cui_list=final_con['Cuisines']# converting the series to dataframe
b_list=cui_list.tolist()

count_list=final_con['Number of restaurants in the country']# converting the
↪series to dataframe
c_list=count_list.tolist()
```

```
[48]: from plotly.offline import download_plotlyjs, init_notebook_mode, iplot
from plotly.graph_objs import *
init_notebook_mode()
import plotly.graph_objs as go #importing plotly or graphs
trace0 = go.Bar(# BarChart 1 (Popular cuisines of the country)
x=b_list, #x axis label
```

```

y=c_list, # y axis label
text=loc_list, # location of the cuisine
name='Popular Cuisine',
marker=dict( color=['rgb(255,69,0)',
                    'rgb(255,140,0)',
                    'rgb(165,42,42)',
                    'rgb(220,20,60)',
                    'rgb(255,0,0)',
                    'rgb(255,99,71)',
                    'rgb(255,127,80)',
                    'rgb(205,92,92)',
                    'rgb(240,128,128)',
                    'rgb(233,150,122)',
                    'rgb(250,128,114)',
                    'rgb(255,160,122)'],
            line=dict(
                color='rgb(255,0,0)',#color of the bar graph's line
                width=1.5, #width of the bar graph
            )
        ), opacity=1.0
)
data = [trace0]
layout = go.Layout(

    legend=dict( #the layout of the graph( beautification)
        x=0,
        y=1,
        traceorder='normal',
        font=dict(
            family='sans-serif',
            size=12,
            color='#000'
        ), bgcolor='#E2E2E2',
        bordercolor='#FFFFFF',
        borderwidth=20,
    ),
    autosize=False,
    width=1000, # size of the graph
    height=450,
    margin=Margin(r=20, l=300,
                  b=75, t=125),
    title="Graph 2.1 : Most popular cuisines in the World<br>\
    <i>hover with cursor to see location in the country where they are most_
    popular </i>", #title of the graph plot_bgcolor='rgba(245, 246, 249, 1)',
    xaxis=dict(tickangle=-45,title=
    'Cuisine',mirror=True,showticklabels=True),
    #making the graphs label inclined at 45 deg

```

```

    yaxis= {'title': 'Number of restaurants offering<br> cuisine in the_
↪location'},#label of y-axis
)
fig = go.Figure(data=data, layout=layout)#plotting the graph
iplot(fig)

```

/usr/local/lib/python3.7/site-packages/plotly/graph_objs/_deprecations.py:410:
DeprecationWarning:

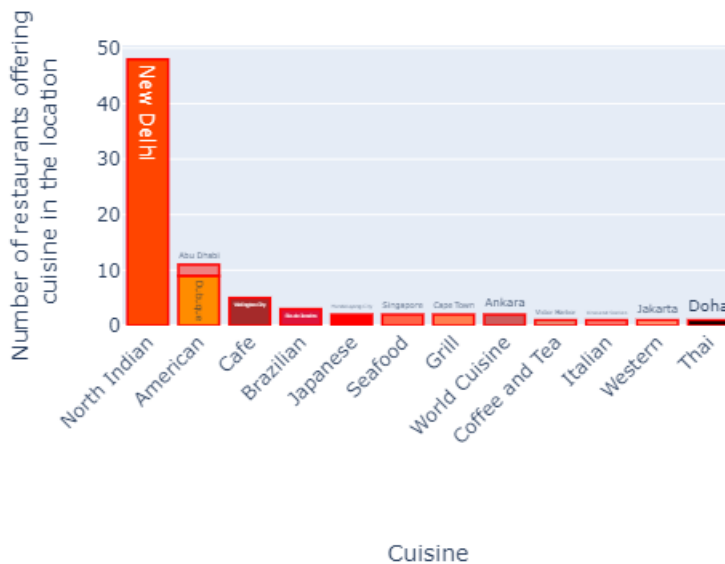
plotly.graph_objs.Margin is deprecated.

Please replace it with one of the following more specific types

- plotly.graph_objs.layout.Margin

Graph 2.1 : Most popular cuisines in the World

hover with cursor to see location in the country where they are most popul



```

[49]: rest_cuisine = pd.
↪DataFrame(data_cuisines[['Restaurant_Name','City','Cuisine_1','Cuisine_2','Cuisine_3','Cuis
↪
↪'Cuisine_5','Cuisine_6','Cuisine_7','Cuisine_8']])
rest_cuisine_stack=pd.DataFrame(rest_cuisine.stack()) #stacking the columns
rest_cuisine.head()

```

```

[49]:
0      Restaurant_Name  City Cuisine_1  Cuisine_2  Cuisine_3 \
      Skye  Jakarta  Italian  Continental  NaN

```

1	Satoo - Hotel Shangri-La	Jakarta	Asian	Indonesian	Western
2	Sushi Masa	Jakarta	Sushi	Japanese	NaN
3	3 Wise Monkeys	Jakarta	Japanese	NaN	NaN
4	Avec Moi Restaurant and Bar	Jakarta	French	Western	NaN

	Cuisine_4	Cuisine_5	Cuisine_6	Cuisine_7	Cuisine_8
0	NaN	NaN	NaN	NaN	NaN
1	NaN	NaN	NaN	NaN	NaN
2	NaN	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN	NaN
4	NaN	NaN	NaN	NaN	NaN

```
[50]: keys1 = [c for c in rest_cuisine if c.startswith('Cuisine')]
b=pd.melt(rest_cuisine, id_vars='Restaurant_Name', value_vars=keys1,
↳value_name='Cuisines')
#melting the stack into one row
max_rate1=pd.DataFrame(b.groupby(by=['Restaurant_Name','variable','Cuisines']).
↳size().reset_index())
#find the highest restuarant in the city
max_rate1
del max_rate1['variable']
max_rate1.columns=['Restaurant_Name','Cuisines','Count']
max_rate1.head(20)
```

```
[50]:
```

	Restaurant_Name	Cuisines	Count
0	12212	Fast Food	1
1	Let's Burrrip	Chinese	1
2	Let's Burrrip	North Indian	1
3	#45	Cafe	1
4	#Dilliwaala6	North Indian	1
5	#InstaFreeze	Ice Cream	1
6	#OFF Campus	Cafe	1
7	#OFF Campus	Continental	1
8	#OFF Campus	Italian	1
9	#OFF Campus	Fast Food	1
10	#Urban Cafè©	North Indian	1
11	#Urban Cafè©	Chinese	1
12	#Urban Cafè©	Italian	1
13	#hashtag	Cafe	1
14	'Ohana	Hawaiian	1
15	10 Downing Street	North Indian	2
16	10 Downing Street	Chinese	2
17	10 To 10 In Delhi	Indian	1
18	10 To 10 In Delhi	Cafe	1
19	11th Avenue Cafe Bistro	Cafe	1

```
[51]: max_rate1.sort_values('Count',ascending=False)
#Cafe Coffee Day has the max number of cuisines and The least number of
↳ cuisines in a restaurant is 1.
```

```
[51]:
```

	Restaurant_Name	Cuisines	Count
2479	Cafe Coffee Day	Cafe	83
4596	Domino's Pizza	Pizza	79
4597	Domino's Pizza	Fast Food	78
12984	Subway	Salad	63
12985	Subway	Healthy Food	63
...
5568	Gabbar's Bar & Kitchen	Chinese	1
5569	Gabbar's Bar & Kitchen	Mexican	1
5570	Gabbar's Bar & Kitchen	Italian	1
5571	Gaga Manjero	World Cuisine	1
15963	İlâukuraŦŦa SofrasŦ±	Izgara	1

[15964 rows x 3 columns]

```
[52]: rating =
↳ data_rest[['Restaurant_ID','Restaurant_Name','Country','City','Aggregate_rating','Average_C
```

```
[53]: rating = rating.
↳ merge(max_rate1,left_on='Restaurant_Name',right_on='Restaurant_Name',how='left')
rating
```

```
[53]:
```

	Restaurant_ID	Restaurant_Name	Country	City	\
0	7402935	Skye	Indonesia	Jakarta	
1	7402935	Skye	Indonesia	Jakarta	
2	7410290	Satoo - Hotel Shangri-La	Indonesia	Jakarta	
3	7410290	Satoo - Hotel Shangri-La	Indonesia	Jakarta	
4	7410290	Satoo - Hotel Shangri-La	Indonesia	Jakarta	
...	
23810	18312106	UrbanCrave	India	Kanpur	
23811	18312106	UrbanCrave	India	Kanpur	
23812	3900245	Deena Chat Bhandar	India	Varanasi	
23813	18246202	VNS Live Studio	India	Varanasi	
23814	18246202	VNS Live Studio	India	Varanasi	

	Aggregate_rating	Average_Cost_for_two	Votes	Price_range	\
0	4.1	800000	1498	3	
1	4.1	800000	1498	3	
2	4.6	800000	873	3	
3	4.6	800000	873	3	
4	4.6	800000	873	3	
...	
23810	3.9	0	127	1	

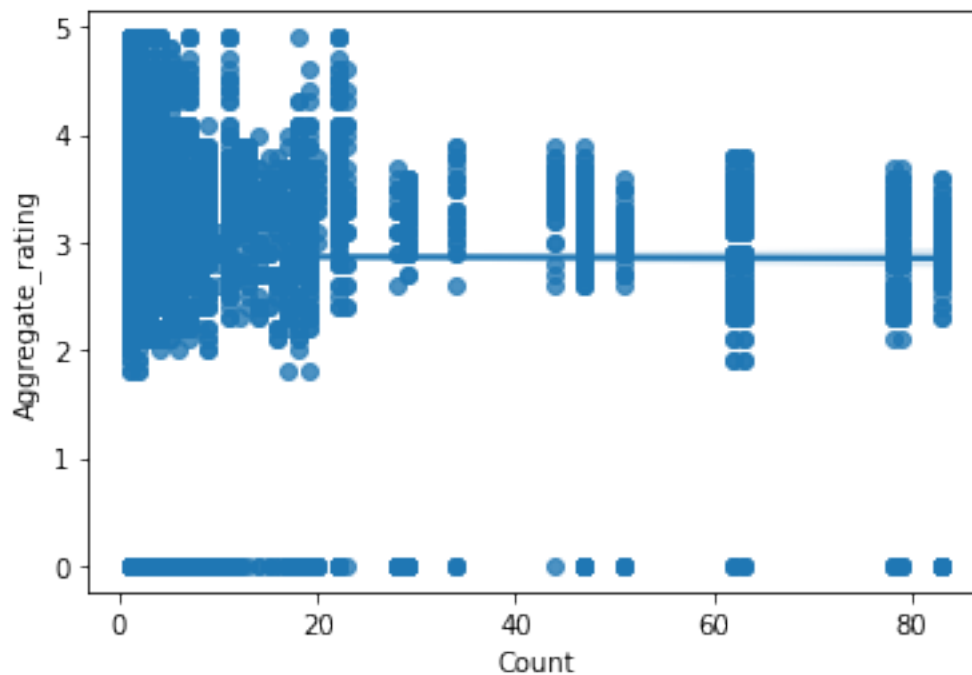
23811	3.9	0	127	1
23812	3.8	0	78	1
23813	3.5	0	109	1
23814	3.5	0	109	1

	Has_Table_booking_Yes	Has_Online_delivery_Yes	Cuisines	Count
0	0	0	Italian	1
1	0	0	Continental	1
2	0	0	Asian	1
3	0	0	Indonesian	1
4	0	0	Western	1
...
23810	0	0	Italian	1
23811	0	0	Beverages	1
23812	0	0	Street Food	1
23813	0	0	Chinese	1
23814	0	0	North Indian	1

[23815 rows x 12 columns]

```
[54]: sns.regplot(x='Count',y='Aggregate_rating',data=rating)
rating[["Count", "Aggregate_rating"]].corr()
#Number of cuisines is not a good factor to decide the rating of a restaurant
```

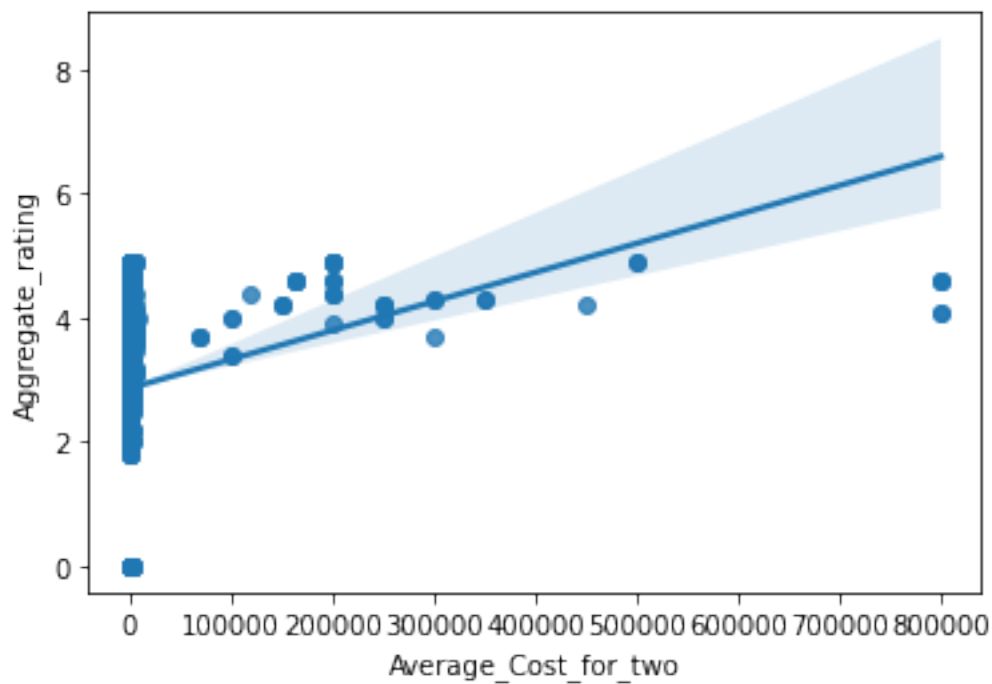
```
[54]:
          Count  Aggregate_rating
Count      1.000000      -0.001642
Aggregate_rating -0.001642      1.000000
```



```
[55]: sns.regplot(x='Average_Cost_for_two',y='Aggregate_rating',data=rating)
rating[["Average_Cost_for_two", "Aggregate_rating"]].corr()
#Average cost for two is a weak positive factor to decide the rating of a
↪restaurant
```

```
[55]:
```

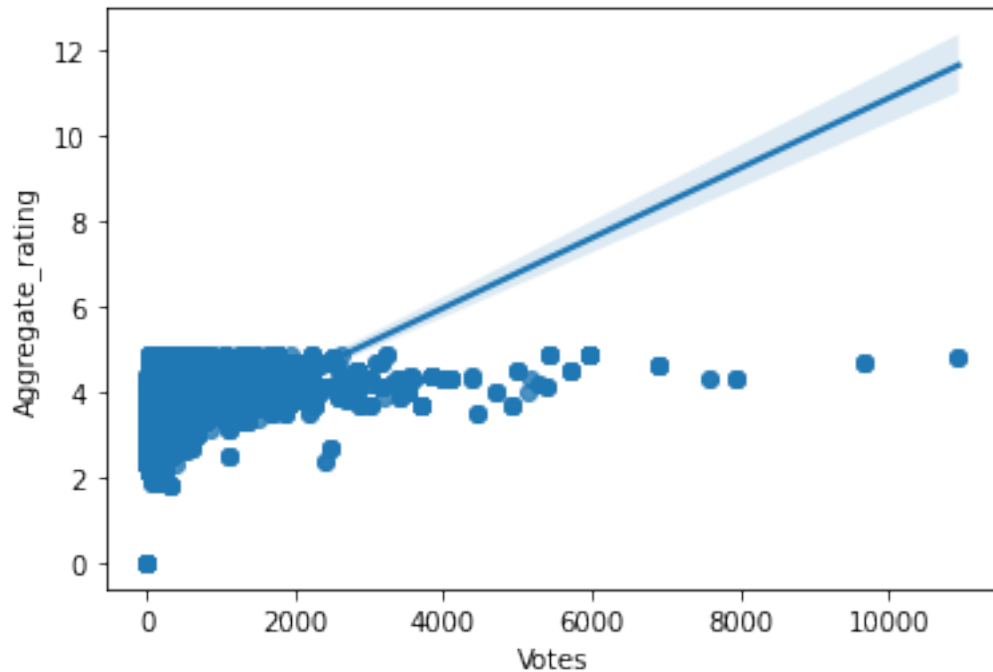
	Average_Cost_for_two	Aggregate_rating
Average_Cost_for_two	1.00000	0.05011
Aggregate_rating	0.05011	1.00000



```
[56]: sns.regplot(x='Votes',y='Aggregate_rating',data=rating)
rating[['Votes','Aggregate_rating']].corr()
##Average cost for two can be a factor to decide the rating of a restaurant
```

```
[56]:
```

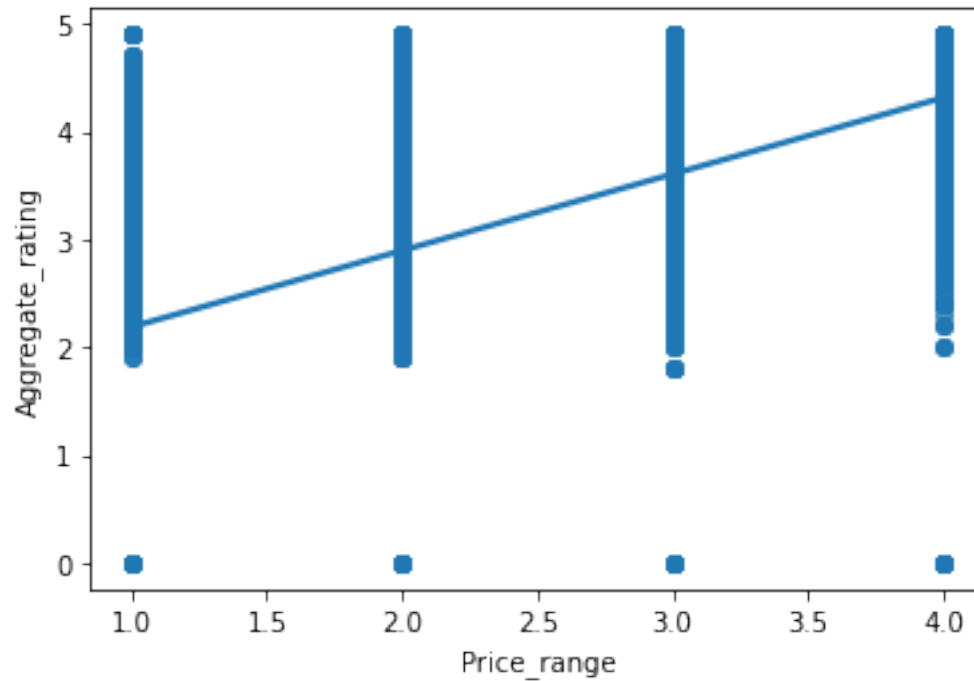
	Votes	Aggregate_rating
Votes	1.000000	0.318667
Aggregate_rating	0.318667	1.000000



```
[57]: abc = data_rest[data_rest['Has_Online_delivery_Yes'] == 1]['Aggregate_rating'].
      ↪mean()
      xyz = data_rest[data_rest['Has_Online_delivery_Yes'] == 0]['Aggregate_rating'].
      ↪mean()
      sns.regplot(x='Price_range',y='Aggregate_rating',data=rating)
      rating[['Price_range','Aggregate_rating']].corr()
      ##Price range can be a factor to decide the rating of a restaurant
```

```
[57]:
```

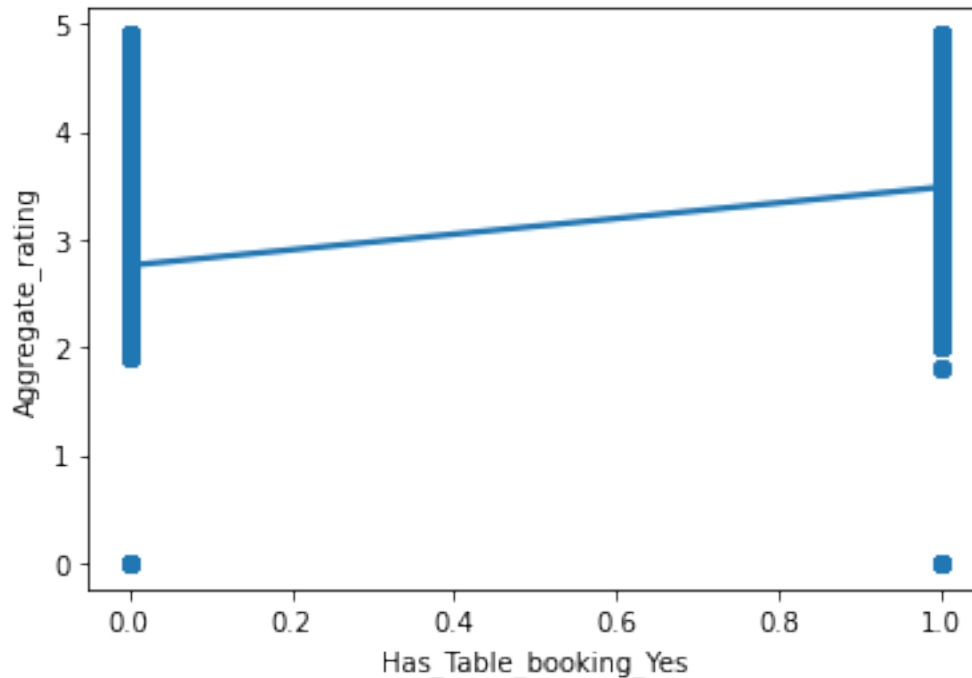
	Price_range	Aggregate_rating
Price_range	1.000000	0.462983
Aggregate_rating	0.462983	1.000000



```
[58]: sns.regplot(x='Has_Table_booking_Yes',y='Aggregate_rating',data=rating)
rating[['Has_Table_booking_Yes','Aggregate_rating']].corr()
##Table booking can be a factor to decide the rating of a restaurant
```

```
[58]:
```

	Has_Table_booking_Yes	Aggregate_rating
Has_Table_booking_Yes	1.000000	0.181843
Aggregate_rating	0.181843	1.000000



We see that there is no single variable that affects the rating strongly, however table booking,online delivery,avg price for two and price range, number of votes do play a part in affecting the rating of a restaurant.

```
[62]: data_rest.to_excel("Restaurant_Zomato_Data.xls",index=False)
```

```
[64]: new=pd.read_excel("Restaurant_Zomato_Data.xls")
```

```
[65]: new
```

```
[65]:
```

	Restaurant_ID	Restaurant_Name	Country_Code	City \
0	7402935	Skye	94	Jakarta
1	7410290	Satoo - Hotel Shangri-La	94	Jakarta
2	7420899	Sushi Masa	94	Jakarta
3	7421967	3 Wise Monkeys	94	Jakarta
4	7422489	Avec Moi Restaurant and Bar	94	Jakarta
...
9545	18279289	BMG - All Day Dining	1	Dehradun
9546	2300497	Atmosphere Grill Cafe Sheesha	1	Kanpur
9547	18312106	UrbanCrave	1	Kanpur
9548	3900245	Deena Chat Bhandar	1	Varanasi
9549	18246202	VNS Live Studio	1	Varanasi

```

                                Address \
0    Menara BCA, Lantai 56, Jl. MH. Thamrin, Thamri...
```

1 Hotel Shangri-La, Jl. Jend. Sudirman
 2 Jl. Tuna Raya No. 5, Penjaringan
 3 Jl. Suryo No. 26, Senopati, Jakarta
 4 Gedung PIC, Jl. Teluk Betung 43, Thamrin, Jakarta
 ...
 9545 140 A, Rajpur Road, Jakhan, Dehradun
 9546 8th Floor, J.S. Tower, 16/106 - Mall Road, Kan...
 9547 14/125, The Mall, Mall Road, Colonelganj, Para...
 9548 D-47/184, Luxa Road, Dashaswmedh Road, Varanasi
 9549 Hotel Varuna Ground Floor, 22 Gulab Bagh, Sigr...

	Locality	Locality_Verbose \
0	Grand Indonesia Mall, Thamrin	Grand Indonesia Mall, Thamrin, Jakarta
1	Hotel Shangri-La, Sudirman	Hotel Shangri-La, Sudirman, Jakarta
2	Penjaringan	Penjaringan, Jakarta
3	Senopati	Senopati, Jakarta
4	Thamrin	Thamrin, Jakarta
...
9545	Jakhan	Jakhan, Dehradun
9546	Mall Road	Mall Road, Kanpur
9547	Parade	Parade, Kanpur
9548	Dashaswmedh Road	Dashaswmedh Road, Varanasi
9549	Sigra	Sigra, Varanasi

	Longitude	Latitude \
0	106.821999	-6.196778
1	106.818961	-6.203292
2	106.800144	-6.101298
3	106.813400	-6.235241
4	106.821023	-6.196270
...
9545	78.068890	30.362686
9546	80.354002	26.472001
9547	80.342796	26.474986
9548	0.000000	0.000000
9549	82.991694	25.318345

	Cuisines	Average_Cost_for_two \
0	Italian, Continental	800000
1	Asian, Indonesian, Western	800000
2	Sushi, Japanese	500000
3	Japanese	450000
4	French, Western	350000
...
9545	Chinese, North Indian, Fast Food	0
9546	Indian, Chinese, Continental	0
9547	Cafe, Continental, Desserts, Ice Cream, Italia...	0

9548	Street Food	0
9549	Chinese, North Indian	0

	Currency	Price_range	Aggregate_rating	Rating_color	\
0	Indonesian Rupiah(IDR)	3	4.1	Green	
1	Indonesian Rupiah(IDR)	3	4.6	Dark Green	
2	Indonesian Rupiah(IDR)	3	4.9	Dark Green	
3	Indonesian Rupiah(IDR)	3	4.2	Green	
4	Indonesian Rupiah(IDR)	3	4.3	Green	
...	
9545	Indian Rupees(Rs.)	1	4.3	Green	
9546	Indian Rupees(Rs.)	1	3.6	Yellow	
9547	Indian Rupees(Rs.)	1	3.9	Yellow	
9548	Indian Rupees(Rs.)	1	3.8	Yellow	
9549	Indian Rupees(Rs.)	1	3.5	Yellow	

	Rating_text	Votes	Country	Has_Table_booking_Yes	\
0	Very Good	1498	Indonesia	0	
1	Excellent	873	Indonesia	0	
2	Excellent	605	Indonesia	0	
3	Very Good	395	Indonesia	0	
4	Very Good	243	Indonesia	0	
...	
9545	Very Good	63	India	0	
9546	Good	34	India	0	
9547	Good	127	India	0	
9548	Good	78	India	0	
9549	Good	109	India	0	

	Has_Online_delivery_Yes
0	0
1	0
2	0
3	0
4	0
...	...
9545	0
9546	0
9547	0
9548	0
9549	0

[9550 rows x 20 columns]

[]: