C	import matplotlib.pyplot as plt import seaborn as sns import matplotlib.pyplot as plt import matplotlib.pyplot as plt import matplotlib.pyplot as plt import matplotlib.pyplot as plt import seaborn as sns import matplotlib.pyplot as plt import seaborn as sns import matplotlib.pyplot as plt import matplotlib.pyplot in port matplotlib.pyplotlib.pyplot in port matplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.pyplotlib.p
3	Rikufuji Avenue, Legaspi Avenue, Legaspi Avenue, Legaspi Edsa Shangri-La, Shangri-La, Shangri-La Shangri-La Shangri-La Third Floor, Mandaluvong Mand
6]: [c 6]: 2 7]: [c	Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address',
S R	Scalars Pandas core frame DataFrame Stange Index
C	20 Votes 9551 non-null dtypes: float64(3), int64(5), object(13)
) : [0)] : [6 6 0	75% 1.835229e+07 1.00000 77.282006 28.642758 700.00000 2.00000 3.70000 131.000000 max 1.850065e+07 216.00000 174.832089 55.976980 80000.000000 4.000000 10934.000000 First find the missing values df.isnull().sum() Restaurant ID 0 0 Restaurant Name 0 0 Country Code 0 0 City 0 0 Address 0 0
L L C C H H H S S F A	Locality Verbose 0 Locality Verbose 0 Longitude 0 Latitude 0 Cuisines 9 Average Cost for two 0 Currency 0 Has Table booking 1 Is delivery 0 Is delivering now 0 Switch to order menu 0 Price range 0 Aggregate rating 0 Rating color 0 Rating text 0
.1]: [.1]: [.3]: [<pre>dtype: int64 [features for features in df.columns if df[features].isnull().sum()>0] ['Cuisines'] sns.heatmap(df.isnull(),yticklabels=False,cmap='viridis') <pre> "></pre></pre>
.4]: c	Country Code Country 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 100 191 Sri Lanka
1 1 1 5]: c	11 208 Turkey 12 214 UAE 13 215 United Kingdom 14 216 United States df1.info() cclass 'pandas.core.frame.DataFrame'> RangeIndex: 15 entries, 0 to 14 Data columns (total 2 columns): # Column Non-Null Count Dtype
6]: (6]: (7]: (7]: 3	0 Country Code 15 non-null int64 1 Country 15 non-null object thypes: int64(1), object(1) memory usage: 368.0+ bytes df1.shape (15, 2) df1.size
8]:	
9]: c 9]: c 0 0 0 0 1]: [1]: c	max 216.00000 df1.isnull().sum() Country Code 0 Country 0 dtype: int64 [features for features in df1.columns if df1[features].isnull().sum()>0] [] df1.columns
1]: I 2]: f 3]: f	Index(['Country Code', 'Country'], dtype='object') fdf = pd.merge(df,df1,on='Country Code',how='left') fdf.head() Restaurant Restaurant Restaurant Name Code City Address Locality Verbose Longitude Latitude Cuisines Has Has Online Dolarion Makkati City Mall Ralayaan Makati City Makkati City Makkati City Makkati City Makkati City Makkati City Dolarion Makkati City Dolarion D
2	Little Tokyo, 2277 Chino Legaspi Legaspi Legaspi
5	Mandaluyong Fashion Ortigas Ortigas Japanese Japanese
4]: C 4]: R C C A L L L C	df.dtypes Restaurant ID int64 Restaurant Name object Country Code int64 City object Address object Locality object Locality verbose object Locality Verbose object Locality terbose object Locality object Locality object Locality Verbose object Locality object Locality object Locality Verbose object Locality object Locality Verbose object Locality object Locality object Locality Verbose object Locality object Locality object Locality Verbose object Locality object Locality object Locality object object Locality object object Locality object object object Locality object
C H H H I I S S F R V C C C C C C C C C C C C C C C C C C	Currency object Has Table booking object Has Online delivery object Has Online delivery object Us delivering now object Switch to order menu object Price range int64 Aggregate rating float64 Rating color object Rating text object Votes int64 dtype: object Restaurant ID int64
C C C C C C C C C C C C C C C C C C C	Restaurant Name object Country Code int64 City object Address object Locality object Locality object Locality Verbose object Locality float64 Latitude float64 Cursines object Average Cost for two int64 Currency object Has Table booking object Has Online delivery object Switch to order menu object
F A A F R V C C C C C C C C C C C C C C C C C C	Price range int64 Aggregate rating float64 Rating color object Rating text object Votes int64 Country object dtype: object Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address',
3]: C	<pre>'Price range', 'Aggregate rating', 'Rating color', 'Rating text', 'Votes', 'Country'], dtype='object') country_name = fdf.Country.value_counts().index country_name Index(['India', 'United States', 'United Kingdom', 'Brazil', 'UAE', 'South Africa', 'New Zealand', 'Turkey', 'Australia', 'Phillipines', 'Indonesia', 'Singapore', 'Qatar', 'Sri Lanka', 'Canada'], dtype='object') country_value = fdf.Country.value_counts().values</pre>
⊙]: (## Pie chart for top country plt.pie(country_value[:3],labels=country_name[:3],autopct='%1.2f%%') ([<matplotlib.patches.wedge 0x1ad0c6f7700="" at="">,</matplotlib.patches.wedge>
I	India 94.39% United Kingdom United States
	Comato have maximum records or transaction are from India after USA and then UK. fdf.columns Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address',
7]: r	dtype='object') ratings=fdf.groupby(['Aggregate rating','Rating color', 'Rating text']).size().reset_index().rename(columns={0:'rating_count'}) ratings 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
1 1 1	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 315 12 2.9 Orange Average 381 13 3.0 Orange Average 468
1 1 1 1 1 1 2 2	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 335 22 3.9 Yellow Good 335
2 2 2 2 2 2 3 3 3	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
3	1. Rting between 4.5 to 4.9> Excellent 2. Rating between 4.0 to 3.4> Very Good 3. Rating between 3.5 to 3.9> Good 4. Rating between 3.0 to 3.4> Average 5. Rating between 2.5 to 2.9> Average 6. Rating between 1.8 to 2.4> Poor
3]:	6. Rating between 1.8 to 2.4> Poor 7. If rating is 0> rated ratings.head() Aggregate rating Rating color Rating text rating_count 0 0.0 White Not rated 2148
4]: i n	
	1500 - 15
	on 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 Aggregate rating", y="rating_count", hue='Rating color', palette=['blue', 'red', 'orange', 'yellow', 'green', 'green'], data=ratings) *AxesSubplot:xlabel='Aggregate rating', ylabel='rating_count'> Rating color White Red Orange Yellow Green Dark Green
	1500 - Dark Green 500 - 00 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 Aggregate rating
Ð]: s	Observation 1. Not rated count is very high that is in Blue Color 2. Maximum number rating is between 2.5 to 3.4 sns.countplot(x="Rating color", data= ratings, palette=['blue', 'red', 'orange', 'yellow', 'green', 'green']) <a ,="" href="mailto:xaxessubplot:xlabel='Rating color" ylabel="count">
***************************************	10 - 8 - 9 - 9 - 9 - 9 - 9 - 9 - 9 - 9 - 9
6]: f	White Red Orange Rating color Sellow Green Dark Green ###counteries name that has given zero rating fdf.columns Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address', 'Locality', 'Locality', 'Locality Verbose', 'Longitude', 'Latitude', 'Cuisines',
0)]:	'Locality', 'Locality Verbose', 'Longitude', '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') fdf[fdf['Rating color']=='White'].groupby("Country").size().reset_index() Country 0 Brazil 5
3 4]: f 4]: 0 1 2	### def Groupby(['Aggregate rating', 'Country']).size().reset_index().head(5) ### Aggregate rating
4 C 6]: #	Observations Maximum number of 0 rating are from India # Which currency is used by which country? fdf.columns Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address',
3]:	
1 1 1	4 Indonesia Indone
1 5]: # 5]: f 1	# Which countries do have online deliveries option? fdf[fdf['Has Online delivery'] == 'Yes'].groupby('Country').size().reset_index() Country 0 India 2423 UAE 28
7]: _	County C
1 1 1	
	fdf.groupby(['Country','Has Online delivery']).size().reset_index() Country
163]: f	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
1 5]: f 5]:1 1 1	14 UAE Yes 28 15 United Kingdom No 80 16 United States No 434
13	Observation 1. Online Deliveries are available in Inida & UAE only. # Pie chart for cities distributions fdf.columns
1	1. Online Deliveries are available in Inida & UAE only. # Pie chart for cities distributions fdf.columns Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address',
106]: f 106]:	1. Online Deliveries are available in Inida & UAE only. # Pie chart for cities distributions Index(['Restaurant ID', 'Restaurant Name', 'Country Code', 'City', 'Address',