

# **PYTHON**: Mini Project

## 1. Importing Libraries

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

### 2. Set Options

```
In [3]: pd.set_option("display.max_columns",40)
    pd.set_option("display.max_rows",30)
    pd.set_option("display.float_format",lambda x :f"{x:,.4f}")
```

#### 3. Read Data

```
In [2]: df = pd.read_csv("/content/drive/MyDrive/DATA SCIENCE AND ML AI/PYTHON/Project
df.head()
```

Out[2]:		res_id	name	establishment	url	address	city o
	0	3400299	Bikanervala	Quick Bites	https://www.zomato.com/ agra/bikanervala- khanda	Kalyani Point, Near Tulsi Cinema, Bypass Road,	Agra
	1	3400005	Mama Chicken Mama Franky House	Quick Bites	https://www.zomato.com/ agra/mama-chicken- mama	Main Market, Sadar Bazaar, Agra Cantt, Agra	Agra
	2	3401013	Bhagat Halwai	Quick Bites	https://www.zomato.com/ agra/bhagat- halwai-2-sh	62/1, Near Easy Day, West Shivaji Nagar, Goalp	Agra
	3	3400290	Bhagat Halwai	Quick Bites	https://www.zomato.com/ agra/bhagat-halwai-civi	Near Anjana Cinema, Nehru Nagar, Civil Lines,	Agra
	4	3401744	The Salt Cafe Kitchen & Bar	Casual Dining	https://www.zomato.com/ agra/the-salt-cafe-kitc	1C,3rd Floor, Fatehabad Road, Tajganj, Agra	Agra

5 rows × 26 columns

# 4. Understand and Prepare the Data

In [4]: df.info()

```
RangeIndex: 211944 entries, 0 to 211943
Data columns (total 26 columns):
    Column
                         Non-Null Count
                                         Dtype
    -----
                                         ----
 0
    res id
                         211944 non-null int64
 1
    name
                         211944 non-null object
 2
    establishment
                         207117 non-null object
 3
    url
                         211944 non-null object
                        211810 non-null object
    address
 5
    city
                         211944 non-null object
6 city id
                        211944 non-null int64
 7
    locality
                         211944 non-null object
 8
    latitude
                        211944 non-null float64
                       211944 non-null float64
 9
    longitude
                        48757 non-null
 10 zipcode
                                         object
 11 country_id
                         211944 non-null int64
 12 locality_verbose
                         211944 non-null object
 13 cuisines
                         210553 non-null object
14 timings
                         208070 non-null object
 15 average cost for two 211944 non-null int64
                         211944 non-null int64
 16 price range
                         211944 non-null object
 17 currency
                         209875 non-null object
 18 highlights
 19 aggregate rating
                         211944 non-null float64
 20 rating_text
                         211944 non-null object
21 votes
                         211944 non-null int64
 22 photo count
                         211944 non-null int64
 23 opentable_support
                       211896 non-null float64
24 delivery
                         211944 non-null int64
25 takeaway
                         211944 non-null int64
dtypes: float64(4), int64(9), object(13)
memory usage: 42.0+ MB
```

<class 'pandas.core.frame.DataFrame'>

#### 5. Understand the variables

In [5]: df.describe()

Out[5]:		res_id	city_id	latitude	longitude	country_id	a١
	count	211,944.0000	211,944.0000	211,944.0000	211,944.0000	211,944.0000	
	mean	13,494,112.3481	4,746.7854	21.4995	77.6153	1.0000	
	std	7,883,721.9725	5,568.7664	22.7813	7.5001	0.0000	
	min	50.0000	1.0000	0.0000	0.0000	1.0000	
	25%	3,301,027.0000	11.0000	15.4961	74.8780	1.0000	
	50%	18,695,734.0000	34.0000	22.5142	77.4260	1.0000	
	<b>75</b> %	18,812,974.0000	11,306.0000	26.8412	80.2193	1.0000	
	max	19,159,790.0000	11,354.0000	10,000.0000	91.8328	1.0000	

In [6]: df.dtypes

Out[6]: **0** 

res_id	int64
name	object
establishment	object
url	object
address	object
city	object
city_id	int64
locality	object
latitude	float64
longitude	float64
zipcode	object
country_id	int64
locality_verbose	object
cuisines	object
timings	object
average_cost_for_two	int64
price_range	int64
currency	object
highlights	object
aggregate_rating	float64
rating_text	object
votes	int64
photo_count	int64
opentable_support	float64
delivery	int64
takeaway	int64

dtype: object

# 6. Check for Missing Values

In [7]: df.isnull().sum()

Out[7]:

	0
res_id	0
name	0
establishment	4827
url	0
address	134
city	0
city_id	0
locality	0
latitude	0
longitude	0
zipcode	163187
country_id	0
locality_verbose	0
cuisines	1391
timings	3874
average_cost_for_two	0
price_range	0
currency	0
highlights	2069
aggregate_rating	0
rating_text	0
votes	0
photo_count	0
opentable_support	48
delivery	0
takeaway	0

dtype: int64

In [8]: df.isnull().mean()

Out[8]: **0** 

res_id	0.0000
name	0.0000
establishment	0.0228
url	0.0000
address	0.0006
city	0.0000
city_id	0.0000
locality	0.0000
latitude	0.0000
longitude	0.0000
zipcode	0.7700
country_id	0.0000
locality_verbose	0.0000
cuisines	0.0066
timings	0.0183
average_cost_for_two	0.0000
price_range	0.0000
currency	0.0000
highlights	0.0098
aggregate_rating	0.0000
rating_text	0.0000
votes	0.0000
photo_count	0.0000
opentable_support	0.0002
delivery	0.0000
takeaway	0.0000

dtype: float64

As i analyze the data ..'opentable\_support' column havin all the values in 0 .. so we can also drop this column

```
In [9]: df = df.drop('opentable_support', axis=1)
In [10]: df
```

Out[10]:		res_id	name	establishment	url	address
	0	3400299	Bikanervala	Quick Bites	https://www.zomato.com/ agra/bikanervala- khanda	Kalyani Point, Near Tulsi Cinema, Bypass Road,
	1	3400005	Mama Chicken Mama Franky House	Quick Bites	https://www.zomato.com/ agra/mama-chicken- mama	Main Market, Sadar Bazaar, Agra Cantt, Agra
	2	3401013	Bhagat Halwai	Quick Bites	https://www.zomato.com/ agra/bhagat- halwai-2-sh	62/1, Near Easy Day, West Shivaji Nagar, Goalp
	3	3400290	Bhagat Halwai	Quick Bites	https://www.zomato.com/ agra/bhagat-halwai-civi	Near Anjana Cinema, Nehru Nagar, Civil Lines, 
	4	3401744	The Salt Cafe Kitchen & Bar	Casual Dining	https://www.zomato.com/ agra/the-salt-cafe-kitc	1C,3rd Floor, Fatehabad Road, Tajganj, Agra
	211939	3202251	Kali Mirch Cafe And Restaurant	Casual Dining	https://www.zomato.com/ vadodara/kali-mirch-caf	Manu Smriti Complex, Near Navrachna School, Gl
	211940	3200996	Raju Omlet	Quick Bites	https://www.zomato.com/ vadodara/raju-omlet- kar	Mahalaxmi Apartment, Opposite B O B, Karoli Ba
	211941	18984164	The Grand Thakar	Casual Dining	https://www.zomato.com/ vadodara/the-grand- thak	3rd Floor, Shreem Shalini

address	url	establishment	name	res_id	
Mall, Opposite Conqu					
G-2, Vedant Platina, Near Cosmos, Akota, Vadodara	https://www.zomato.com/ vadodara/ subway-1-akota	Quick Bites	Subway	3201138	211942
Shop 7, Ground Floor, Opposite Natubhai Circle	https://www.zomato.com/ vadodara/freshcos-the- h	Café	Freshcos - The Health Cafe	18879846	211943

211944 rows × 25 columns

# \*# -- As we can see values in zipcode column mostly null (77%)..so we can drop this column \*

```
In [11]: df.drop('zipcode', axis=1, inplace=True)
In [12]: df.isna().sum()
```

Out[12]:		0
	res_id	0
	name	0
	establishment	4827
	url	0
	address	134
	city	0
	city_id	0
	locality	0
	latitude	0
	longitude	0
	country_id	0
	locality_verbose	0
	cuisines	1391
	timings	3874
	average_cost_for_two	0
	price_range	0
	currency	0
	highlights	2069
	aggregate_rating	0
	rating_text	0
	votes	0
	photo_count	0
	delivery	0
	takeaway	0

dtype: int64

Now as we can see in the column 'esatablishment' some values are missing .. we can fill at those missing values by mode (this column has object data type)

```
In [13]: df['establishment']=df['establishment'].fillna(df['establishment'].mode()[0])
In [14]: df.isna().mean()
```

Out[14]:	0

0.0000
0.0000
0.0000
0.0000
0.0006
0.0000
0.0000
0.0000
0.0000
0.0000
0.0000
0.0000
0.0066
0.0183
0.0000
0.0000
0.0000
0.0098
0.0000
0.0000
0.0000
0.0000
0.0000
0.0000

dtype: float64

In [15]: df.dtypes

Out[15]: **0** 

res_id	int64
name	object
establishment	object
url	object
address	object
city	object
city_id	int64
locality	object
latitude	float64
longitude	float64
country_id	int64
locality_verbose	object
cuisines	object
timings	object
average_cost_for_two	int64
price_range	int64
currency	object
highlights	object
aggregate_rating	float64
rating_text	object
votes	int64
photo_count	int64
delivery	int64
takeaway	int64

dtype: object

\*here all the columns('cuisines','timing','highlights'), have some missing values having object data type .. so i am applying same mode method to fill the values same as before \*

```
In [16]: df['cuisines']=df['cuisines'].fillna(df['cuisines'].mode()[0])
In [17]: df['timings']=df['timings'].fillna(df['timings'].mode()[0])
In [18]: df['highlights']=df['highlights'].fillna(df['highlights'].mode()[0])
In [19]: df.isna().mean()
```

Out[19]: 0 res id 0.0000 0.0000 name establishment 0.0000 0.0000 url address 0.0006 city 0.0000 city id 0.0000 locality 0.0000 latitude 0.0000 longitude 0.0000 country\_id 0.0000 locality\_verbose 0.0000 cuisines 0.0000 timings 0.0000 average\_cost\_for\_two 0.0000 price\_range 0.0000 currency 0.0000 highlights 0.0000 aggregate\_rating 0.0000 rating\_text 0.0000 votes 0.0000 photo\_count 0.0000 delivery 0.0000 **takeaway** 0.0000

dtype: float64

# \*and for the address i just fill the value as in the city column \*

```
In [20]: df['address'] = df['address'].fillna(df['city'])
```

In [21]: df

Out[21]:		res_id	name	establishment	url	address
	0	3400299	Bikanervala	Quick Bites	https://www.zomato.com/ agra/bikanervala- khanda	Kalyani Point, Near Tulsi Cinema, Bypass Road,
	1	3400005	Mama Chicken Mama Franky House	Quick Bites	https://www.zomato.com/ agra/mama-chicken- mama	Main Market, Sadar Bazaar, Agra Cantt, Agra
	2	3401013	Bhagat Halwai	Quick Bites	https://www.zomato.com/ agra/bhagat- halwai-2-sh	62/1, Near Easy Day, West Shivaji Nagar, Goalp
	3	3400290	Bhagat Halwai	Quick Bites	https://www.zomato.com/ agra/bhagat-halwai-civi	Near Anjana Cinema, Nehru Nagar, Civil Lines, 
	4	3401744	The Salt Cafe Kitchen & Bar	Casual Dining	https://www.zomato.com/ agra/the-salt-cafe-kitc	1C,3rd Floor, Fatehabad Road, Tajganj, Agra
	211939	3202251	Kali Mirch Cafe And Restaurant	Casual Dining	https://www.zomato.com/ vadodara/kali-mirch-caf	Manu Smriti Complex, Near Navrachna School, Gl
	211940	3200996	Raju Omlet	Quick Bites	https://www.zomato.com/ vadodara/raju-omlet- kar	Mahalaxmi Apartment, Opposite B O B, Karoli Ba
	211941	18984164	The Grand Thakar	Casual Dining	https://www.zomato.com/ vadodara/the-grand- thak	3rd Floor, Shreem Shalini

address	url	establishment	name	res_id	
Mall, Opposite Conqu					
G-2, Vedant Platina, Near Cosmos, Akota, Vadodara	https://www.zomato.com/ vadodara/ subway-1-akota	Quick Bites	Subway	3201138	211942
Shop 7, Ground Floor, Opposite Natubhai Circle	https://www.zomato.com/ vadodara/freshcos-the- h	Café	Freshcos - The Health Cafe	18879846	211943

211944 rows × 24 columns

In [22]: df.isna().sum()

Out[22]: 0 res id 0 name 0 establishment 0 url 0 address 0 city 0 city\_id 0 locality 0 latitude 0 longitude 0 country\_id 0 **locality\_verbose** 0 cuisines 0 timings 0 average\_cost\_for\_two 0 price\_range 0 currency 0 highlights 0 aggregate\_rating 0 rating\_text 0 votes 0 photo\_count 0 delivery 0 takeaway 0

dtype: int64

## 7. Study Correlation

```
In [23]: num_columns = ['average_cost_for_two','price_range','aggregate_rating','votes'
corr = df[num_columns].corr()
```

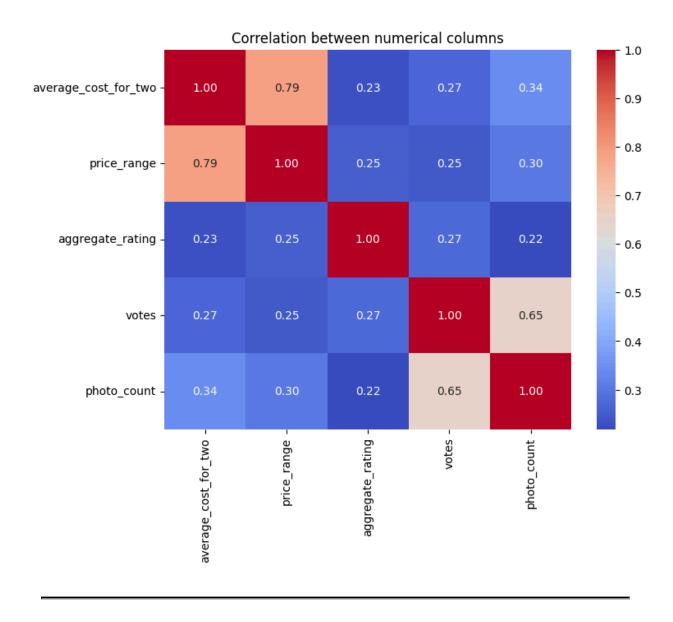
corr Out[23]: average cost for two price range aggregate rating 1.0000 average cost for two 0.7859 0.2323 0.26 0.7859 price\_range 1.0000 0.2523 0.24 1.0000 0.27 aggregate\_rating 0.2323 0.2523 votes 0.2690 0.2450 0.2720 1.00 0.3419 0.3014 0.2185 0.65 photo\_count In [ ]: ## Outcomes: # average cost for two <-> price range (0.7859): it shows strong positive -> h # average cost for two <-> aggregate rating(0.2323) : Weak positive -> expensi # average cost for two <-> photo count(0.3419) : Moderate positive -> expensiv # aggregate rating <-> votes(0.2720) : Weak positive -> more votes usually mea # aggregate rating <-> price range(0.2523) : Weak positive -> premium restaura # votes <-> photo count(0.6550) : Strong positive -> more photos uploaded usua In [24]: # Creating Heatmap:

sns.heatmap(corr, annot=True, cmap="coolwarm", fmt=".2f")

plt.title("Correlation between numerical columns")

plt.figure(figsize=(8,6))

plt.show()

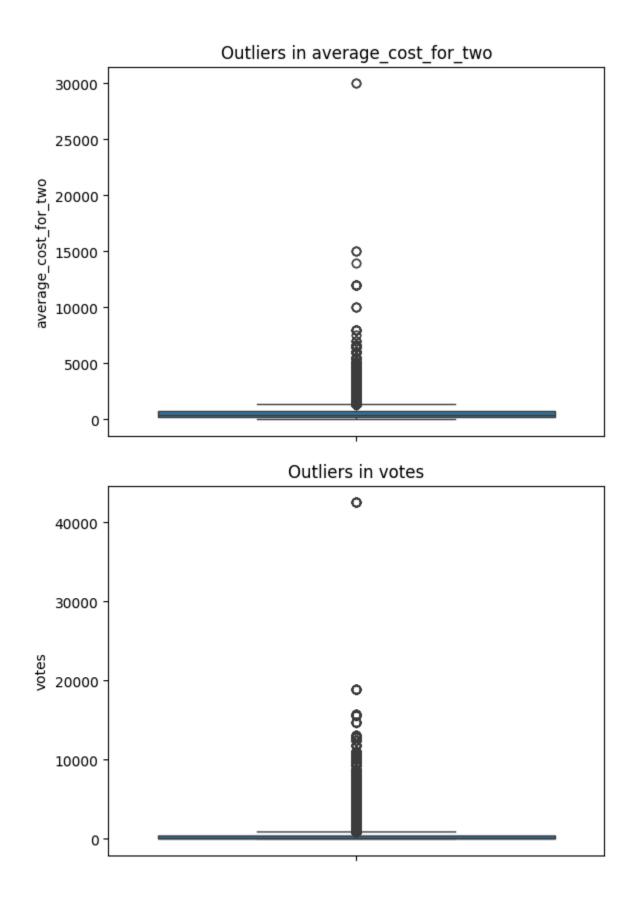


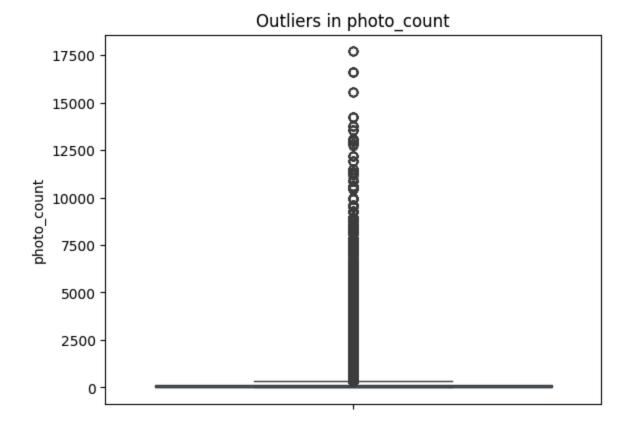
#### 8. Detect Outliers

```
In [36]: sns.boxplot(y=df['average_cost_for_two'])
    plt.title("Outliers in average_cost_for_two")
    plt.show()

sns.boxplot(y=df['votes'])
    plt.title("Outliers in votes")
    plt.show()

sns.boxplot(y=df['photo_count'])
    plt.title("Outliers in photo_count")
    plt.show()
```





```
In [ ]: # --> As i can see average_cost_for_two columns has one very high outlier (₹36
# --> As i can see votes columns has one very high outlier (above 40000) and s
# --> As i can see photo_count has 3 big outlier(above 15000)
In [ ]:
```

# 9. Creating a new column 'region'

```
In [25]: df['city'].unique()
```

```
Out[25]: array(['Agra', 'Ahmedabad', 'Gandhinagar', 'Ajmer', 'Alappuzha', 'Allahabad', 'Amravati', 'Amritsar', 'Aurangabad', 'Bangalore',
                      'Bhopal', 'Bhubaneshwar', 'Chandigarh', 'Mohali', 'Panchkula',
                      'Zirakpur', 'Nayagaon', 'Chennai', 'Coimbatore', 'Cuttack', 'Darjeeling', 'Dehradun', 'New Delhi', 'Gurgaon', 'Noida',
                       'Faridabad', 'Ghaziabad', 'Greater Noida', 'Dharamshala',
                      'Gangtok', 'Goa', 'Gorakhpur', 'Guntur', 'Guwahati', 'Gwalior', 'Haridwar', 'Hyderabad', 'Secunderabad', 'Indore', 'Jabalpur',
                       'Jaipur', 'Jalandhar', 'Jammu', 'Jamnagar', 'Jamshedpur', 'Jhansi', 'Jodhpur', 'Junagadh', 'Kanpur', 'Kharagpur', 'Kochi', 'Kolhapur',
                      'Kolkata', 'Howrah', 'Kota', 'Lucknow', 'Ludhiana', 'Madurai', 'Manali', 'Mangalore', 'Manipal', 'Udupi', 'Meerut', 'Mumbai',
                       'Thane', 'Navi Mumbai', 'Mussoorie', 'Mysore', 'Nagpur',
                       'Nainital', 'Nashik', 'Neemrana', 'Ooty', 'Palakkad', 'Patiala',
                      'Patna', 'Puducherry', 'Pune', 'Pushkar', 'Raipur', 'Rajkot', 'Ranchi', 'Rishikesh', 'Salem', 'Shimla', 'Siliguri', 'Srinagar', 'Surat', 'Thrissur', 'Tirupati', 'Trichy', 'Trivandrum', 'Udaipur',
                       'Varanasi', 'Vellore', 'Vijayawada', 'Vizag', 'Vadodara'],
                     dtype=object)
In [26]: # Defining some cities :
             city region = {
                  'Agra': 'North-East',
                  'Delhi': 'North', 'New Delhi': 'North', 'Gurgaon': 'North', 'Noida': 'Nort
                  'Mumbai': 'West', 'Pune': 'West',
                  'Bangalore': 'South', 'Chennai': 'South', 'Hyderabad': 'South',
                  'Kolkata': 'East', 'Patna': 'East',
                  'Bhopal': 'Central', 'Indore': 'Central'
             }
             # Apply mapping, using map
             df['region'] = df['city'].map(city region)
             # Replace NaN with "Other"
             df['region'] = df['region'].fillna('Other')
             print(df[['city','region']].head(25))
```

```
city
        region
   Agra North-East
0
1
  Agra North-East
2
  Agra North-East
3 Agra North-East
4 Agra North-East
5 Agra North-East
6 Agra North-East
7 Agra North-East
8 Agra North-East
9 Agra North-East
10 Agra North-East
11 Agra North-East
12 Agra North-East
13 Agra North-East
14 Agra North-East
15 Agra North-East
16 Agra North-East
17 Agra North-East
18 Agra North-East
19 Agra North-East
20 Agra North-East
21 Agra North-East
22 Agra North-East
23 Agra North-East
24 Agra North-East
```

In [27]: df.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 211944 entries, 0 to 211943 Data columns (total 25 columns):

#	Column	Non-Null Count	Dtype			
0	res_id	211944 non-null	int64			
1	name	211944 non-null	object			
2	establishment	211944 non-null	object			
3	url	211944 non-null	object			
4	address	211944 non-null	object			
5	city	211944 non-null	object			
6	city_id	211944 non-null	int64			
7	locality	211944 non-null	object			
8	latitude	211944 non-null	float64			
9	longitude	211944 non-null	float64			
10	country_id	211944 non-null	int64			
11	locality_verbose	211944 non-null	object			
12	cuisines	211944 non-null	object			
13	timings	211944 non-null	object			
14	average_cost_for_two	211944 non-null	int64			
15	price_range	211944 non-null	int64			
16	currency	211944 non-null	object			
17	highlights	211944 non-null	object			
18	aggregate_rating	211944 non-null	float64			
19	rating_text	211944 non-null	object			
20	votes	211944 non-null	int64			
21	photo_count	211944 non-null	int64			
22	delivery	211944 non-null	int64			
23	takeaway	211944 non-null	int64			
24	region	211944 non-null	object			
dtypes: float64(3), int64(9), object(13)						

dtypes: float64(3), int64(9), object(13)

memory usage: 40.4+ MB

```
In [28]: df.isna().sum()
```

Out[28]: 0 res\_id 0 name 0 establishment 0 url 0 address 0 city 0 city\_id 0 locality 0 latitude 0 longitude 0 country\_id 0 locality\_verbose 0 cuisines 0 timings 0 average\_cost\_for\_two 0 price\_range 0 currency 0 highlights 0 aggregate\_rating 0 rating\_text 0 votes 0 photo\_count 0 delivery 0 takeaway 0 region 0

dtype: int64