Zomato Data Analysis Using Python

Data Analytics Project Using Python

Project Overview

Project Title:

Zomato Data Analysis Using Python

Introduction:

Zomato is one of the largest food delivery platforms globally, with millions of monthly transacting customers and a vast network of restaurant partners. With the rise of data-driven decision-making, analyzing Zomato's data can provide valuable insights into customer behavior, restaurant performance, and overall market trends.

Objective:

The primary goal of this project is to conduct an in-depth Exploratory Data Analysis (EDA) on Zomato's dataset using Python. The analysis aims to uncover key insights, trends, and patterns that can help optimize business strategies, improve customer experience, and drive growth for both Zomato and its restaurant partners.

Dataset Description:

The dataset used in this project contains information on various restaurants listed on Zomato. It includes details such as restaurant names, online orders, book table, ratings, votes, cost for two people, and other relevant features. The data offers a comprehensive view of the restaurant landscape on the Zomato platform.

Key Questions Explored:

- What type of restaurant do the majority of customers order from?
- How many votes has each type of restaurant received from customers?
- What are the ratings that the majority of restaurants have received?
- Zomato has observed that most couples order most of their food online. What is their average spending on each order?
- Which mode(online or offline) has received the maximum rating?
- Which type of restaurants received more offline orders, so that Zomato can provide those customers with some good offers?

Tools and Techniques:

- Python Libraries: Pandas, NumPy, Matplotlib, Seaborn
- Data Cleaning: Converting data types.
- Data Visualization: Creating insightful visualizations to illustrate key findings, such as bar charts, scatter plots, and heatmaps.
- Exploratory Analysis: Investigating correlations, distributions, and trends within the data.

Expected Outcomes: The analysis is expected to provide actionable insights into customer preferences, popular restaurants, cost distribution, and rating trends. These findings can help Zomato enhance its platform's performance, improve customer satisfaction, and support restaurants in making data-driven decisions.

Importing Libraries

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

Loading the dataset

```
In [20]: data = pd.read_csv(r"C:\Users\ykuma\Downloads\Zomato data .csv")
data
```

Out[20]:

	name	online_order	book_table	rate	votes	approx_cost(for two people)	listea_in(type)
0	Jalsa	Yes	Yes	4.1/5	775	800	Buffet
1	Spice Elephant	Yes	No	4.1/5	787	800	Buffet
2	San Churro Cafe	Yes	No	3.8/5	918	800	Buffet

3	Addhuri Udupi Bhojana	No	No	3.7/5	88	300	Buffet
4	Grand Village	No	No	3.8/5	166	600	Buffet
				***	***		
143	Melting Melodies	No	No	3.3/5	0	100	Dining
144	New Indraprasta	No	No	3.3/5	0	150	Dining
145	Anna Kuteera	Yes	No	4.0/5	771	450	Dining
146	Darbar	No	No	3.0/5	98	800	Dining
147	Vijayalakshmi	Yes	No	3.9/5	47	200	Dining

148 rows × 7 columns

Data Cleaning

```
In [24]: # Converting the data type of the column 'Rate' to float and remove the denominator

def rating(value):
    value = str(value).split('/')
    value = value[0];
    return float(value)

data['rate'] = data['rate'].apply(rating)
data.head()
```

Out[24]:

	name	online_order	book_table	rate	votes	approx_cost(for two people)	listed_in(type)
0	Jalsa	Yes	Yes	4.1	775	800	Buffet
1	Spice Elephant	Yes	No	4.1	787	800	Buffet
2	San Churro Cafe	Yes	No	3.8	918	800	Buffet
3	Addhuri Udupi Bhojana	No	No	3.7	88	300	Buffet
4	Grand Village	No	No	3.8	166	600	Buffet

Summary of the dataframe

```
In [25]: data.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 148 entries, 0 to 147
         Data columns (total 7 columns):
                                          Non-Null Count Dtype
             Column
                                                         object
                                          148 non-null
             name
             online_order
                                         148 non-null
                                                         object
                                      148 non-null
             book_table
                                                         object
                                         148 non-null
                                                         float64
             rate
                                         148 non-null
                                                         int64
             votes
             approx_cost(for two people) 148 non-null
                                                         int64
             listed_in(type)
                                                         object
                                         148 non-null
         dtypes: float64(1), int64(2), object(4)
         memory usage: 8.2+ KB
```

NOTE:- There is no null values in the dataframe

Business Problem

20 -

Buffet

Cafes

listed_in(type)

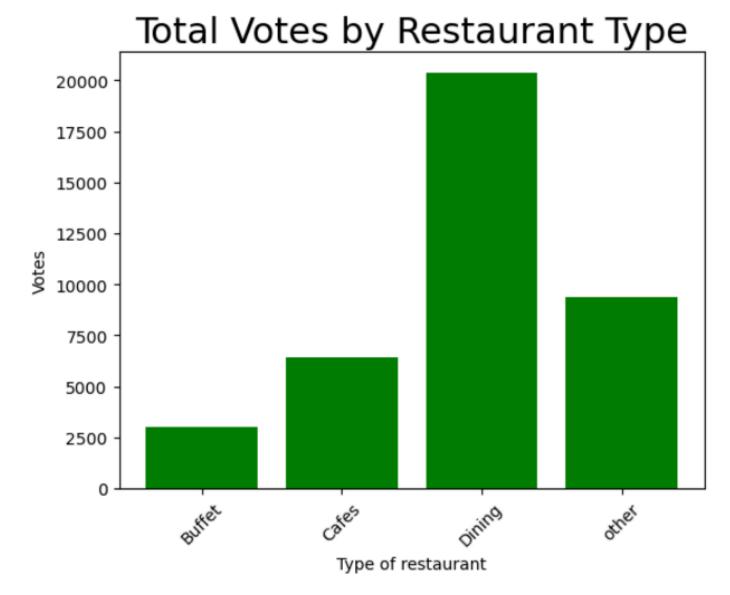
```
In [27]: # 1. What type of restaurant do the majority of customers order from ?
         sns.countplot(data= data, x='listed_in(type)')
         plt.xlabel('Type of Restaurant')
Out[27]: <Axes: xlabel='listed_in(type)', ylabel='count'>
             100
              80
          count
              60 -
              40
```

other

Dining

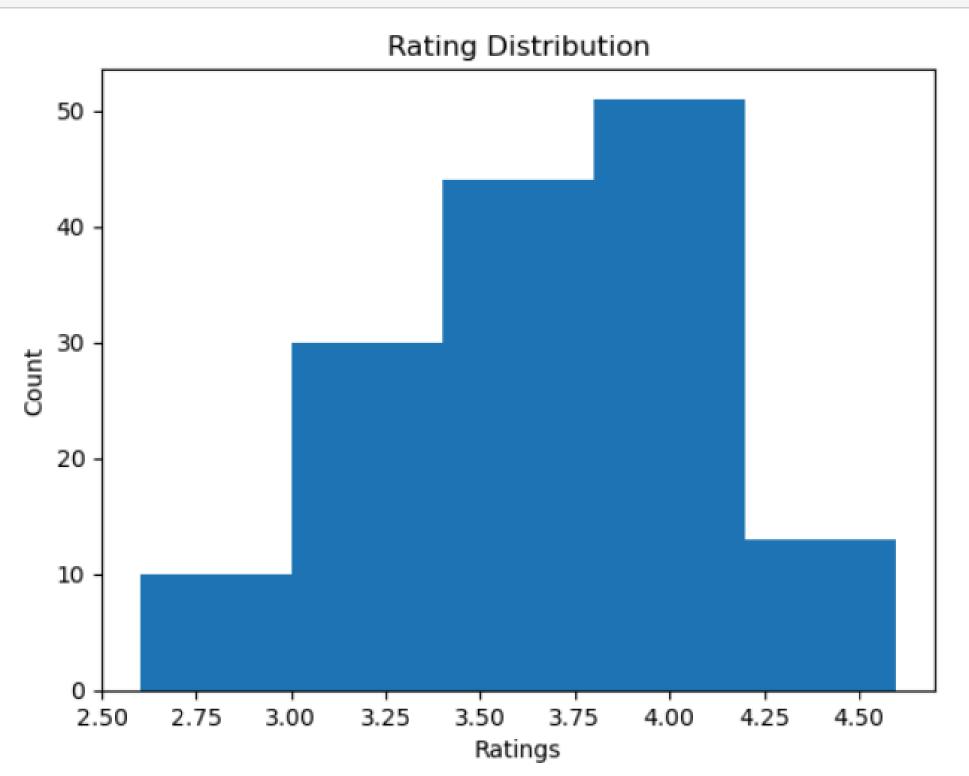
Conclusion:- Dining Restaurants are preferred by a larger number of individuals.

```
In [86]: # 2. How many votes has each type of restaurant received from customers ?
    grouped_data = data.groupby('listed_in(type)')['votes'].sum().reset_index()
    #plotting graph
    plt.bar(grouped_data['listed_in(type)'], grouped_data['votes'], color='green')
    plt.xlabel('Type of restaurant')
    plt.ylabel('Votes')
    plt.title('Total Votes by Restaurant Type', size=22)
    plt.xticks(rotation=45)
    plt.show()
```



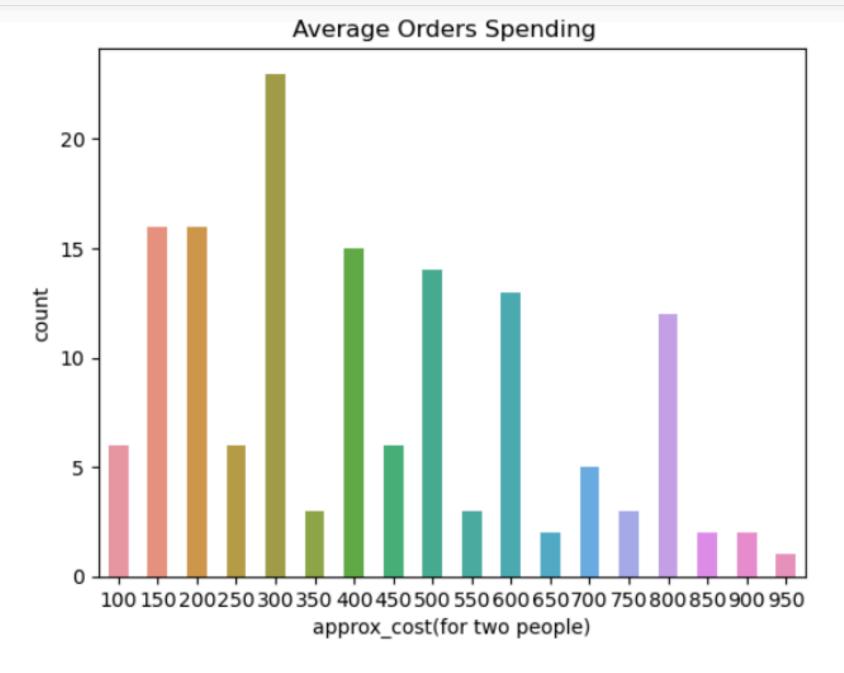
Conclusion: - Dining Restaurants has received maximum votes.

In [49]: # 3. what are the ratings that the majority of restaurants have received ?
 plt.hist(data['rate'],bins=5)
 plt.title('Rating Distribution')
 plt.xlabel('Ratings')
 plt.ylabel('Count')
 plt.show()



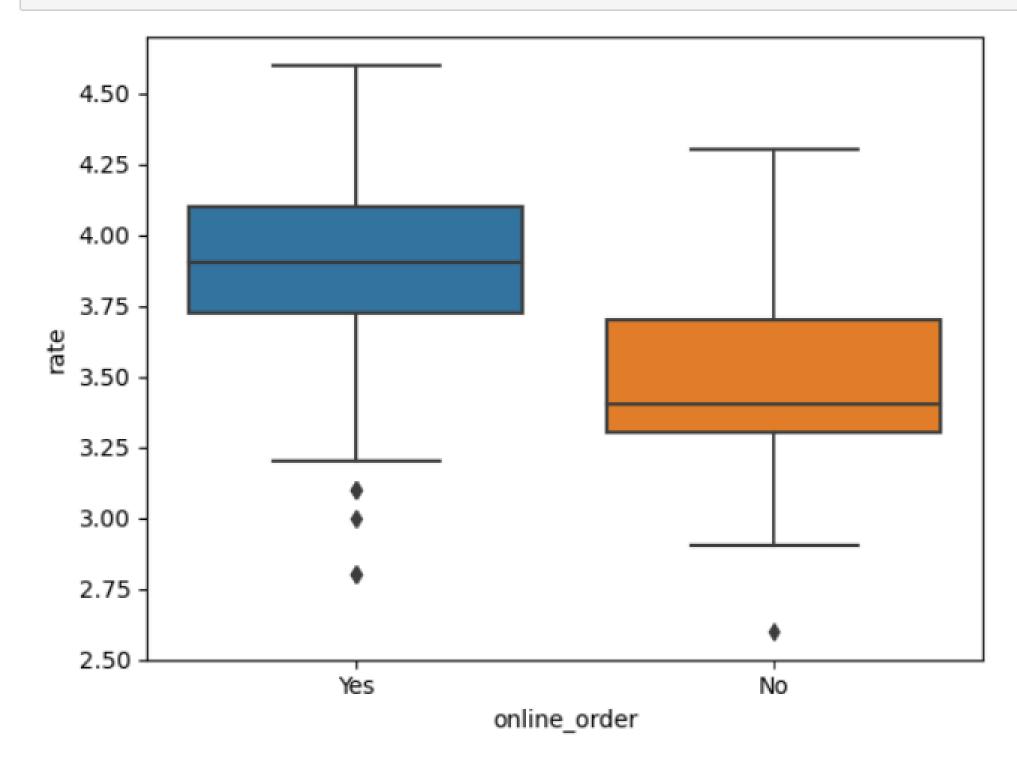
Conclusion:- The majority restaurants received ratings from 3.5 to 4.0

```
In [69]: # 4. Zomato has observed that most of couples order most of their food online. what is their average spending on each order ?
    couple_data= data['approx_cost(for two people)']
    plt.title('Average Orders Spending ')
    plt.xlabel('Cost')
    sns.countplot(x=couple_data, width=0.5)
    plt.show()
```



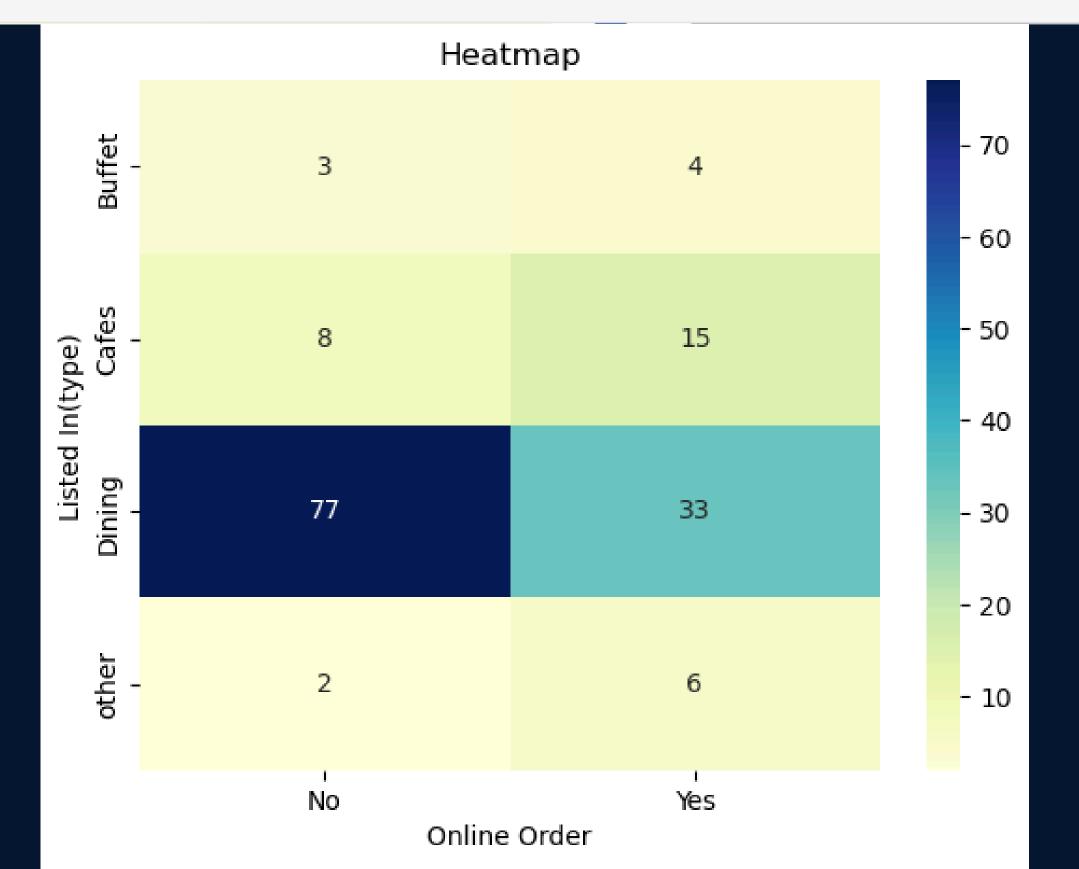
Conclusion:- The majority of couples prefer restaurants with an approximate cost of 300 rupees.

```
In [70]: # 5. Which mode has received the maximum rating ?
    sns.boxplot(data=data, x='online_order', y= 'rate')
    plt.show()
```



Conclusion:- Offline orders receive lower rating as compare to onine orders.

In [89]: # 6. Which type of restaurant received more offline orders, so that Zomato can provide those customers with some good offers ?
 pivot_table= data.pivot_table(index= 'listed_in(type)', columns= 'online_order', aggfunc ='size', fill_value=0)
 sns.heatmap(pivot_table, annot=True, cmap='YlGnBu', fmt='d')
 plt.title('Heatmap')
 plt.xlabel('Online Order')
 plt.ylabel('Listed In(type)')
 plt.show()



Conclusion:- Dining Restaurants primarily accept offline orders, whereas cafes primarilty receive online orders. This suggests that clients prefer to place orders in person at restaurants, but prefer online ordering at cafes.

Suggestions for Improvement

- 1. **Improve Customer Engagement:** The data suggests that certain types of restaurants receive significantly more votes (indicating higher customer engagement) compared to others. Stakeholders should focus on promoting these popular restaurant types on the platform to attract more customers.
- **2. Enhance Restaurant Listings:** Based on the analysis, dining options have the highest votes. Stakeholders could invest in expanding the number of dining restaurants on the platform or enhancing the visibility of these listings to capitalize on customer preferences.
- **3. Customer Feedback Mechanism:** The high level of votes indicates that customers are willing to provide feedback. Stakeholders should consider implementing a more structured feedback mechanism to gather insights for continuous improvement of services.

- **4. Targeted Marketing Campaigns:** Use the insights from the analysis to create targeted marketing campaigns. For example, promoting the types of restaurants that are less popular to balance customer distribution across different categories.
- 5. Data-Driven Decision Making: Encourage stakeholders to continue using data analytics to inform decision-making processes. Regularly analyzing customer preferences and market trends will allow Zomato to stay competitive and responsive to changes in customer behavior.

These suggestions will help stakeholders in making informed decisions to improve the overall customer experience and business performance on the Zomato platform



Thank You