
Analyzing Swiggy

Wireframe Documentation

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Analysis:

As per the problem statement, we have defined the several Use Cases to perform the analysis on which helps in not only understanding the meaningful relationships between attributes but it also allows us to do our own research and come-up with our findings.

Basic Information:

```
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<class 'pandas.core.frame.DataFrame'>
Int64Index: 117 entries, 0 to 117
Data columns (total 7 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   shop_name             117 non-null    object
 1   cuisine               117 non-null    object
 2   location              117 non-null    object
 3   rating               117 non-null    float64
 4   cost_for_two          117 non-null    int64
 5   avg_cost_per_person   117 non-null    float64
 6   cost_category         117 non-null    object
dtypes: float64(2), int64(1), object(4)
memory usage: 7.3+ KB
None
```

The project analyzed the data of food delivery service provider, Swiggy, in Bangalore. The dataset used in this project was obtained from Google Drive and consisted of information about restaurants in Bangalore, their cuisines, ratings, and costs.

TOP 5 ROWS OF THE DATA:

```
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   shop_name      cuisine \
0   Kanti Sweets      Sweets
1   Mumbai Tiffin  North Indian, Home Food, Thalys, Combo
2   Sri Krishna sagar  South Indian, North Indian, Fast Food, Beverag...
3       Al Daaz  American, Arabian, Chinese, Desserts, Fast Foo...
4   Beijing Bites      Chinese, Thai

   location      rating  cost_for_two  avg_cost_per_person \
0  Koramangala, Koramangala      4.3         150         75.0
1      Sector 5, HSR      4.4         400        200.0
2    6th Block, Koramangala      4.1         126         63.0
3          HSR, HSR      4.4         400        200.0
4    5th Block, Koramangala      4.1         450        225.0
```

The top 5 rows of the dataset were displayed using the `head()` function. This showed the column names and some sample data from the dataset, including the restaurant name, location, rating, cuisines, and average cost for two people.

NUMBER OF UNIQUE VALUES FOR EACH ATTRIBUTE:

```
Number of unique values for each attribute:
```

```
shop_name      114
cuisine        79
location       65
rating         12
cost_for_two    30
avg_cost_per_person  30
cost_category    2
dtype: int64
```

The unique values for each attribute were determined using the `nunique()` function. This showed the number of unique values for each attribute in the dataset, including the number of unique restaurant names, unique locations, unique cuisines, and unique ratings.

AVERAGE COST FOR EACH CUISINE:

Average cost for each cuisine:

```
cuisine
American 450.000000
American, Arabian, Chinese, Desserts, Fast Food, Mughlai, North Indian 400.000000
American, Fast Food 350.000000
Andhra, Biryani 433.333333
Andhra, Biryani, Chinese, Desserts, Fast Food, Seafood, South Indian 225.000000
...
South Indian, North Indian, Fast Food, Beverages, Jain 126.000000
South Indian, Snacks, North Indian, Chinese 250.000000
Sweets 150.000000
Turkish, Portuguese, American 300.000000
Turkish, Portuguese, American, Grill 300.000000
Name: cost_for_two, Length: 79, dtype: float64
```

The average cost for each cuisine was calculated using the `groupby()` function. This grouped the data based on the cuisine and calculated the average cost for two people for each cuisine. The output showed the average cost for each cuisine in descending order, with the cuisine with the highest average cost listed first.

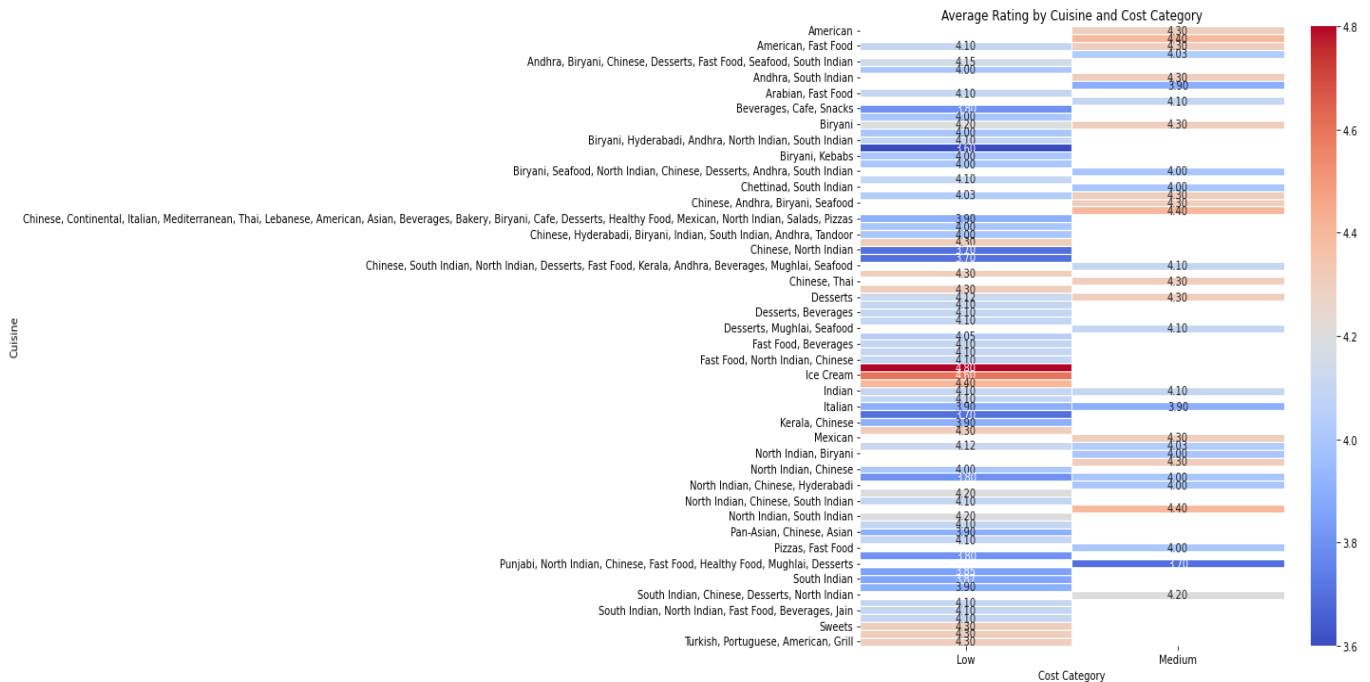
AVERAGE RATING FOR EACH CUISINE :

Average rating for each cuisine:

```
cuisine
American 4.300000
American, Arabian, Chinese, Desserts, Fast Food, Mughlai, North Indian 4.400000
American, Fast Food 4.166667
Andhra, Biryani 4.033333
Andhra, Biryani, Chinese, Desserts, Fast Food, Seafood, South Indian 4.150000
...
South Indian, North Indian, Fast Food, Beverages, Jain 4.100000
South Indian, Snacks, North Indian, Chinese 4.100000
Sweets 4.300000
Turkish, Portuguese, American 4.300000
Turkish, Portuguese, American, Grill 4.300000
Name: rating, Length: 79, dtype: float64
```

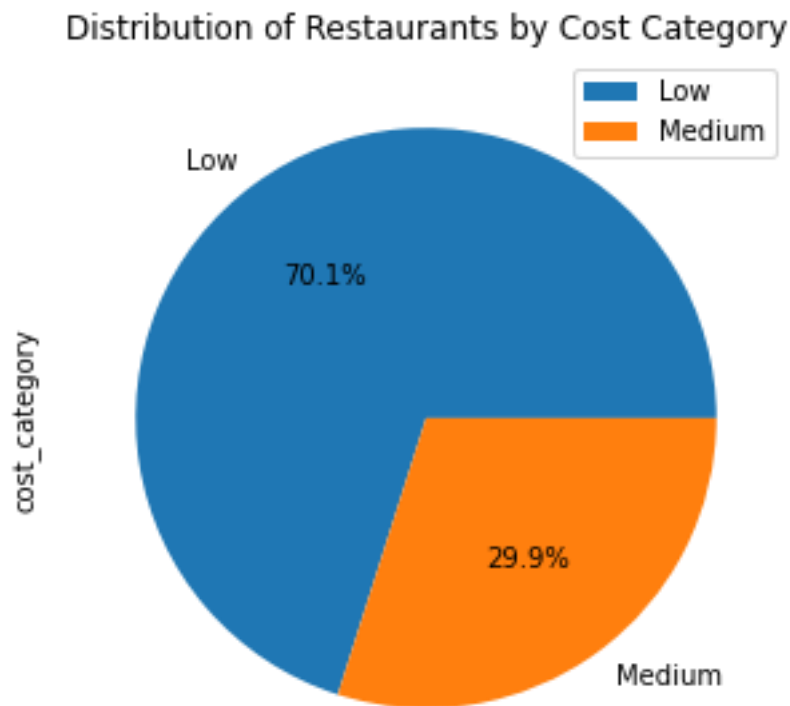
The average rating for each cuisine was also calculated using the `groupby()` function. This grouped the data based on the cuisine and calculated the average rating for each cuisine. The output showed the average rating for each cuisine in descending order, with the cuisine with the highest average rating listed first.

Average Rating by Cuisine And Cost Category :



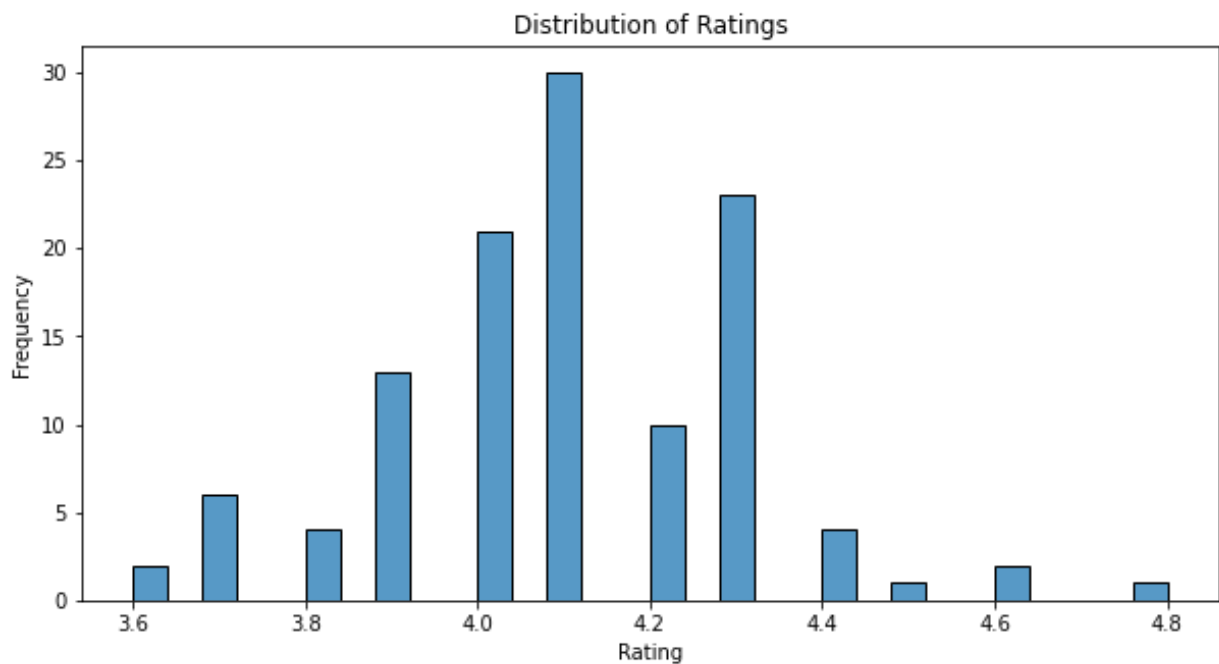
The heatmap displays the average cost and rating for each cuisine in Bangalore. This analysis helps to identify which cuisines have the highest average cost and rating. This information can help customers make informed choices when deciding on which cuisine to order.

Pie chart of the distribution of restaurants by cost category:



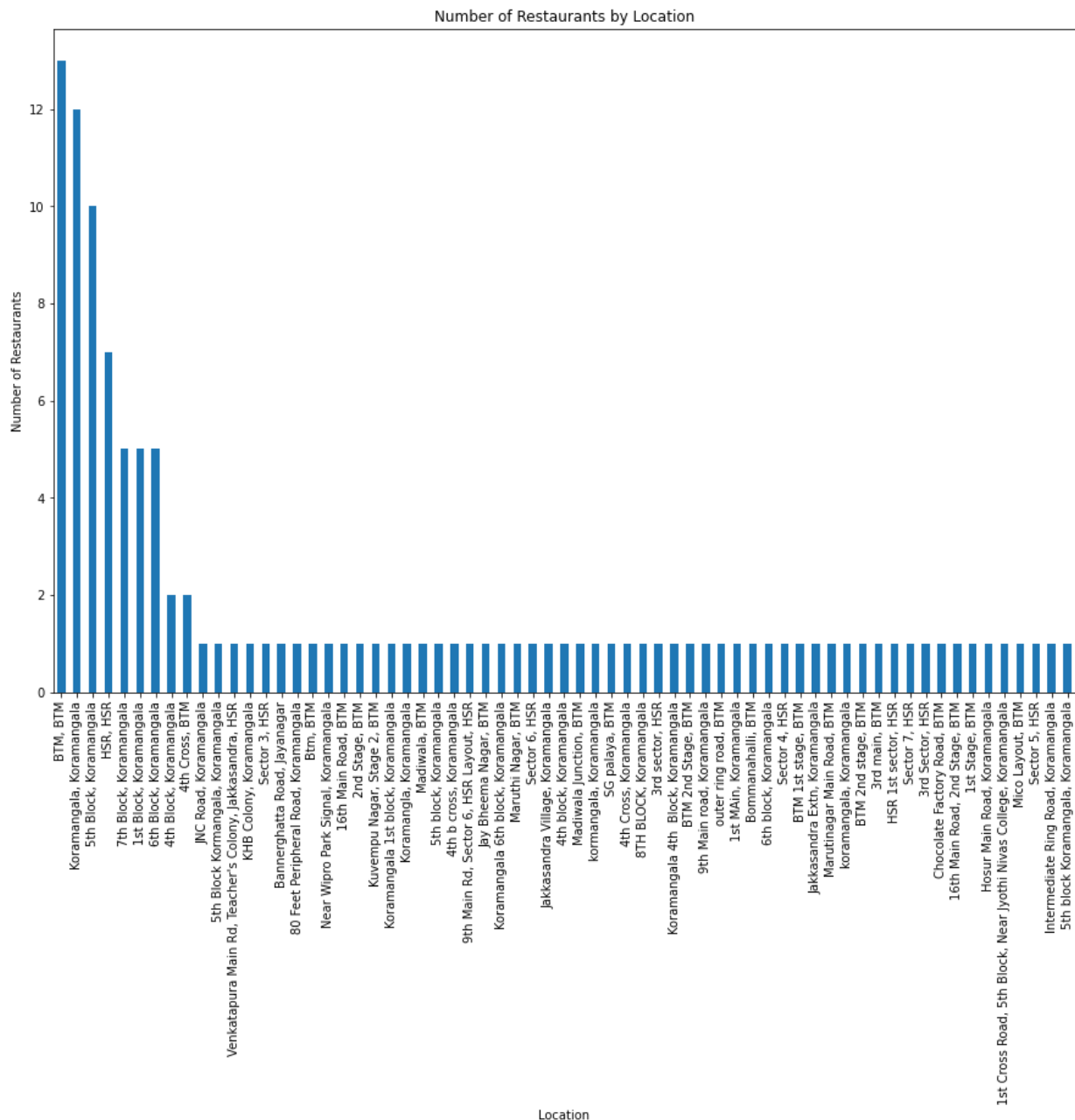
The pie chart displays the distribution of restaurants based on their cost category. The analysis helps to identify the percentage of restaurants that fall under each cost category. This information can help customers choose restaurants based on their budget.

Histplot for distribution of ratings:



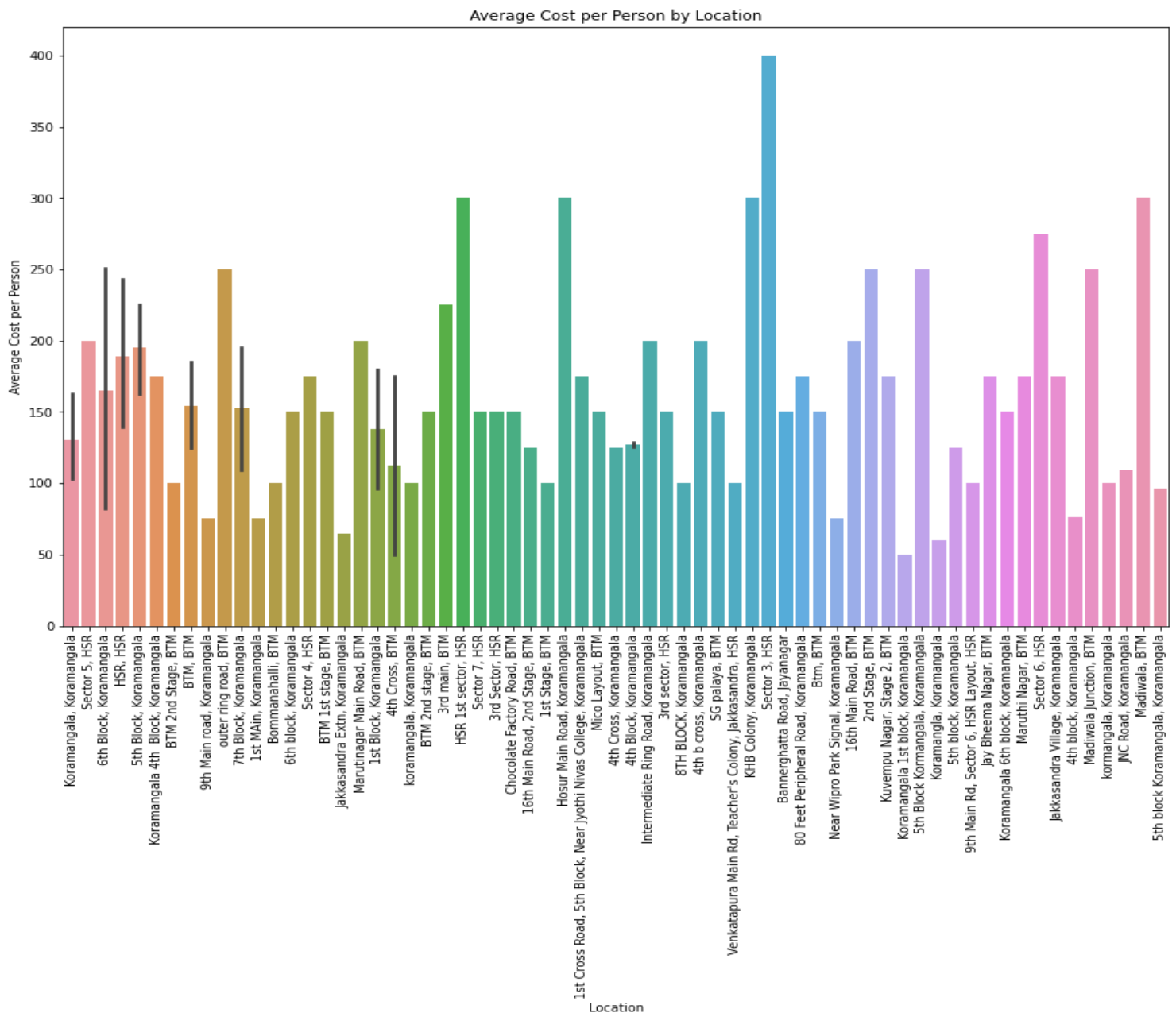
The histogram displays the distribution of ratings for all the restaurants in Bangalore. The analysis helps to identify the distribution of ratings and the most common rating score given by customers.

Number of restaurants by location:



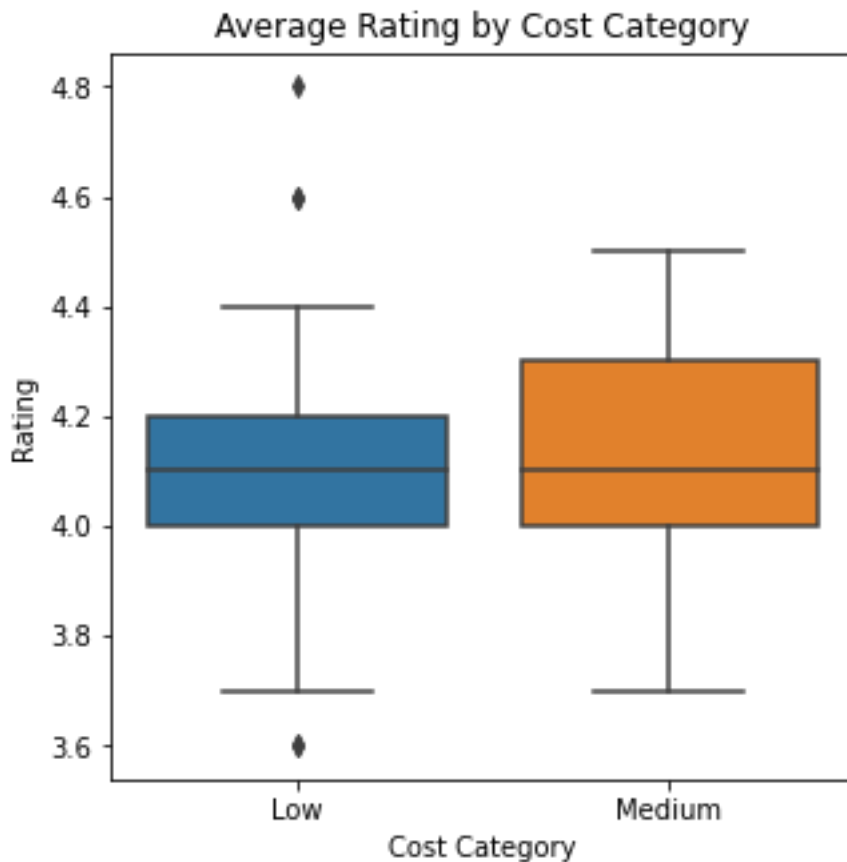
The analysis displays the number of restaurants by location. This information can help customers identify the areas with the highest number of restaurants and make informed choices when deciding on which area to order from.

Barplot for Average cost per person by location:



The barplot displays the average cost per person by location. This analysis helps to identify the areas with the highest and lowest average cost per person. This information can help customers choose restaurants based on their budget.

Box plot for average rating by cost category:



The box plot displays the average rating by cost category. This analysis helps to identify which cost category has the highest average rating and which has the lowest. This information can help customers choose restaurants based on their budget and still enjoy good ratings.

Overall, the analysis performed in the project provides useful insights for customers in Bangalore who use Swiggy for food delivery services. The information can help customers make informed choices on which cuisine, restaurant, and location to order from based on their budget, preferred rating, and location.