

"Analyzing Swiggy - Bangalore Outlet"

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INTRODUCTION AND PROBLEM STATEMENT

The online food ordering market includes foods prepared by restaurants, prepared by independent people, and groceries being ordered online and then picked up or delivered. The first online food ordering service, World Wide Waiter (now known as Waiter.com), was founded in 1995. Online food ordering is the process of ordering food from a website or other application. The product can be either ready-to-eat food or food that has not been specially prepared for direct consumption.

Do ETL: Extract-Transform-Load the dataset and find for me some information from this large data. This is a form of data mining. What all information can be achieved by mining this data,would be explained in a class by the trainer Find key metrics and factors and show the meaningful relationships between attributes. Do your own research and come up with your findings.

```
In [1]: #importing necessary libraries
```

```
import matplotlib.pyplot as plt
import numpy as np
import sklearn as sklearn
import seaborn as sns
import pandas as pd
```

```
In [2]: %matplotlib inline
```

```
In [3]: # importing the dataset (Swiggy)
df=pd.read_csv('Swiggy data.csv')
```

```
In [4]: size= df.shape
print('The Number of Rows: ',size[0], ' and Number of Columns: ',size[1])
```

The Number of Rows: 118 and Number of Columns: 5

In [5]: `df.head()`

Out[5]:

	Shop_Name	Cuisine	Location	Rating	Cost_for_Two
0	Kanti Sweets	Sweets	Koramangala, Koramangala	4.3	₹ 150
1	Mumbai Tiffin	North Indian, Home Food, Thalís, Combo	Sector 5, HSR	4.4	₹ 400
2	Sri Krishna sagar	South Indian, North Indian, Fast Food, Beverag...	6th Block, Koramangala	4.1	₹ 126
3	Al Daaz	American, Arabian, Chinese, Desserts, Fast Foo...	HSR, HSR	4.4	₹ 400
4	Beijing Bites	Chinese, Thai	5th Block, Koramangala	4.1	₹ 450

PRE-PROCESSING DATASET FOR EXPLORATION

In [6]: `df.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 118 entries, 0 to 117
Data columns (total 5 columns):
 #   Column          Non-Null Count  Dtype
---  -
 0   Shop_Name       118 non-null    object
 1   Cuisine         118 non-null    object
 2   Location        118 non-null    object
 3   Rating          118 non-null    object
 4   Cost_for_Two    118 non-null    object
dtypes: object(5)
memory usage: 4.7+ KB
```

Rating and Cost_for_Two should be Floating type

In [7]: `df['Rating'].unique()`

Out[7]: `array(['4.3', '4.4', '4.1', '4.2', '3.9', '3.8', '4', '3.7', '3.6', '4.8', '4.5', '4.6', '--'], dtype=object)`

```
In [8]: # Data Points with rating as -- should be replaced with 0
df['Rating'].replace('--',0,inplace=True)
```

```
In [9]: # now changing the Data Type of Rating Attribute from String to Float
df['Rating']=[float(i) for i in df['Rating']]
```

```
In [10]: df['Cost_for_Two'].unique()
```

```
Out[10]: array(['₹ 150', '₹ 400', '₹ 126', '₹ 450', '₹ 350', '₹ 200', '₹ 500',
               '₹ 247', '₹ 550', '₹ 300', '₹ 129', '₹ 250', '₹ 268', '₹ 600',
               '₹ 527', '₹ 130', '₹ 257', '₹ 280', '₹ 399', '₹ 220', '₹ 800',
               '₹ 100', '₹ 178', '₹ 120', '₹ 251', '₹ 650', '₹ 132', '₹ 153',
               '₹ 219', '₹ 193'], dtype=object)
```

```
In [11]: # now changing the Data Type of Cost_for_Two Attribute from String to Float
list1=[]
list2=[]
for i in df['Cost_for_Two']:
    res=i.split()
    list1.append(res[0])
    list2.append(float(res[1]))
# list2
# list1
```

```
In [12]: df['Cost for Two']=list2
df.drop(['Cost_for_Two'],axis=1,inplace=True)
```

```
In [13]: df
```

Out[13]:

	Shop_Name	Cuisine	Location	Rating	Cost for Two
0	Kanti Sweets	Sweets	Koramangala, Koramangala	4.3	150.0
1	Mumbai Tiffin	North Indian, Home Food, Thalís, Combo	Sector 5, HSR	4.4	400.0
2	Sri Krishna sagar	South Indian, North Indian, Fast Food, Beverag...	6th Block, Koramangala	4.1	126.0
3	Al Daaz	American, Arabian, Chinese, Desserts, Fast Foo...	HSR, HSR	4.4	400.0
4	Beijing Bites	Chinese, Thai	5th Block, Koramangala	4.1	450.0
...
113	Wok Paper Scissors	Pan-Asian, Chinese, Asian	JNC Road, Koramangala	3.9	219.0
114	Savoury Restaurant	Arabian, Middle Eastern, North Indian, Grill, ...	Madiwala, BTM	4.1	600.0
115	Royal Treat	North Indian, Chinese, Seafood, Biryani	5th block Koramangala, Koramangala	4.2	193.0
116	Thali 99	North Indian	Koramangala, Koramangala	4.3	200.0
117	Mani's Dum Biryani	Andhra, Biryani	1st Block, Koramangala	4.2	400.0

118 rows × 5 columns

In [14]: `# rechecking dataset
df.info()`

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 118 entries, 0 to 117  
Data columns (total 5 columns):  
#   Column      Non-Null Count  Dtype  
---  -  
0   Shop_Name    118 non-null    object  
1   Cuisine      118 non-null    object  
2   Location     118 non-null    object  
3   Rating       118 non-null    float64  
4   Cost for Two 118 non-null    float64  
dtypes: float64(2), object(3)  
memory usage: 4.7+ KB
```

we can observe that there is **NO Missing Values** in any attribute.

Now, Separating out the values in Cuisine Attribute into multiple attributes based on the Cusines

```
In [15]: # Split cuisine column into multiple columns using one-hot encoding
cuisine_dummies = df['Cuisine'].str.get_dummies(',')

# Normalize column names by removing leading/trailing spaces
cuisine_dummies.columns = cuisine_dummies.columns.str.strip()

# Merge columns with the same names into a single column
cuisine_merged = cuisine_dummies.groupby(cuisine_dummies.columns, axis=1).sum()

# Concatenate the merged column with the original dataset
data = pd.concat([df, cuisine_merged], axis=1)

# Drop the original cuisine columns
data.drop('Cuisine', axis=1, inplace=True)
```

```
In [16]: data.head()
```

```
Out[16]:
```

	Shop_Name	Location	Rating	Cost for Two	American	Andhra	Arabian	Asian	Bakery	Beverages	...	Rajasthani	Salads	Seafood	Snacks	South Indian	Sweets	Tan...
0	Kanti Sweets	Koramangala, Koramangala	4.3	150.0	0	0	0	0	0	0	...	0	0	0	0	0	1	
1	Mumbai Tiffin	Sector 5, HSR	4.4	400.0	0	0	0	0	0	0	...	0	0	0	0	0	0	
2	Sri Krishna sagar	6th Block, Koramangala	4.1	126.0	0	0	0	0	0	1	...	0	0	0	0	1	0	
3	Al Daaz	HSR, HSR	4.4	400.0	1	0	1	0	0	0	...	0	0	0	0	0	0	
4	Beijing Bites	5th Block, Koramangala	4.1	450.0	0	0	0	0	0	0	...	0	0	0	0	0	0	

5 rows × 52 columns

```
In [17]: # updated shape
size2= data.shape
print('The Number of Rows: ',size2[0], ' and Number of Columns: ',size2[1])
```

The Number of Rows: 118 and Number of Columns: 52

```
In [18]: data.describe()
```

```
Out[18]:
```

	Rating	Cost for Two	American	Andhra	Arabian	Asian	Bakery	Beverages	Biryani	Cafe	...	Rajasthani	Salads	Seaf
count	118.000000	118.000000	118.000000	118.000000	118.000000	118.000000	118.000000	118.000000	118.000000	118.000000	...	118.000000	118.000000	118.000000
mean	4.061864	321.008475	0.093220	0.127119	0.033898	0.025424	0.025424	0.076271	0.220339	0.016949	...	0.008475	0.016949	0.076271
std	0.430845	137.286804	0.291981	0.334526	0.181739	0.158080	0.158080	0.266563	0.416243	0.129631	...	0.092057	0.129631	0.266563
min	0.000000	100.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000
25%	4.000000	204.750000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000
50%	4.100000	300.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000
75%	4.300000	400.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000
max	4.800000	800.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	...	1.000000	1.000000	1.000000

8 rows × 50 columns

EXPLORATORY DATA ANALYSIS (EDA) AND VISUALIZATION

1. Cuisine Analysis

```
In [19]: # creating a sub dataframe having only cuisine columns
data2=data.iloc[:,4:]
data2.head(3)
```

Out[19]:

	American	Andhra	Arabian	Asian	Bakery	Beverages	Biryani	Cafe	Chaat	Chettinad	...	Rajasthani	Salads	Seafood	Snacks	South Indian	Sweets	Tandoor	Th
0	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	1	0	
1	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0	0	
2	0	0	0	0	0	1	0	0	0	0	...	0	0	0	0	1	0	0	

3 rows × 48 columns

```
In [20]: # Get the cuisine columns (assuming they are already encoded using get-dummies)
cuisine_columns = [col for col in data2.columns]

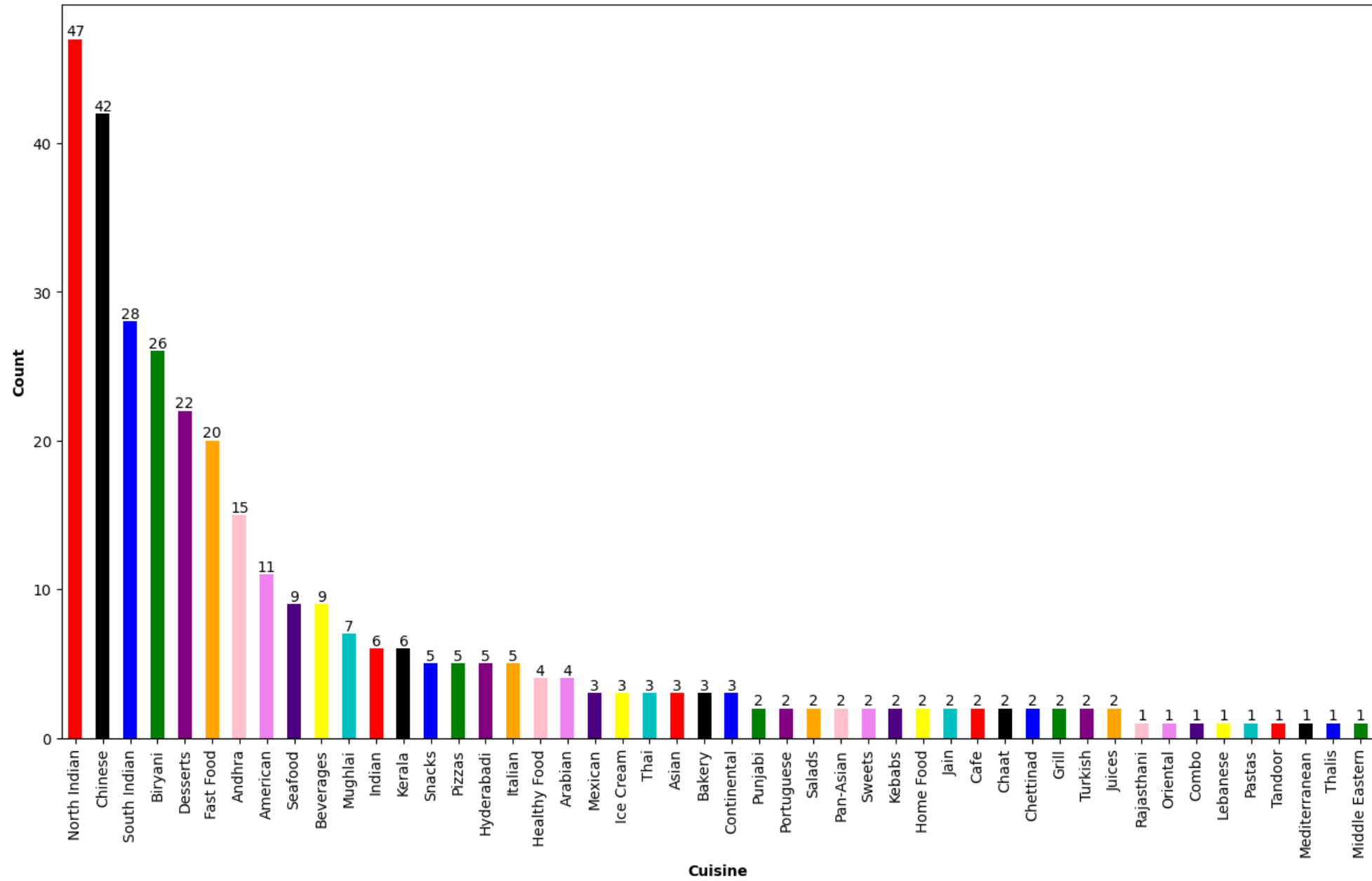
# Calculate the cuisine counts
cuisine_counts = data2[cuisine_columns].sum().sort_values(ascending=False)
print (cuisine_counts)
```

North Indian	47
Chinese	42
South Indian	28
Biryani	26
Desserts	22
Fast Food	20
Andhra	15
American	11
Seafood	9
Beverages	9
Mughlai	7
Indian	6
Kerala	6
Snacks	5
Pizzas	5
Hyderabadi	5
Italian	5
Healthy Food	4
Arabian	4
Mexican	3
Ice Cream	3
Thai	3
Asian	3
Bakery	3
Continental	3
Punjabi	2
Portuguese	2
Salads	2
Pan-Asian	2
Sweets	2
Kebabs	2
Home Food	2
Jain	2
Cafe	2
Chaat	2
Chettinad	2
Grill	2
Turkish	2
Juices	2
Rajasthani	1
Oriental	1


```
Combo          1
Lebanese       1
Pastas         1
Tandoor        1
Mediterranean  1
Thalis         1
Middle Eastern 1
dtype: int64
```

```
In [21]: # Plot the cuisine distribution
cd=['red','black','blue','green','purple','orange','pink','violet','indigo','yellow','c']
plt.figure(figsize=(16, 9))
cuisine_counts.plot(kind='bar',color=cd)
plt.xlabel('Cuisine',fontweight='bold')
plt.ylabel('Count',fontweight='bold')
plt.title('Cuisine Distribution',fontweight='bold',size=24)
plt.xticks(rotation=90)
for i, v in enumerate(cuisine_counts.values):
    plt.annotate(str(v), xy=(i, v),ha='center',color='black',va='bottom')
plt.show()
```

Cuisine Distribution



From the Above graph we can observe that **Top 3 Cuisines in Bangalore (Swiggy Network)** are:

1. North Indian
2. Chinese
3. South Indian

and the least available Cuisines are:

1. Rajasthani
2. Oriental
3. Combo
4. Lebanese
5. Pastas
6. Tandoor
7. Mediterranean
8. Thalís
9. Middle Eastern

2. Location Analysis

```
In [22]: location_counts = data['Location'].value_counts()
```

```
In [23]: location_counts
```

```

Out[23]: BTM, BTM 13
          Koramangala, Koramangala 12
          5th Block, Koramangala 10
          HSR, HSR 8
          7th Block, Koramangala 5
          ..
          1st Cross Road, 5th Block, Near Jyothi Nivas College, Koramangala 1
          Mico Layout, BTM 1
          Sector 5, HSR 1
          Intermediate Ring Road, Koramangala 1
          5th block Koramangala, Koramangala 1
          Name: Location, Length: 65, dtype: int64

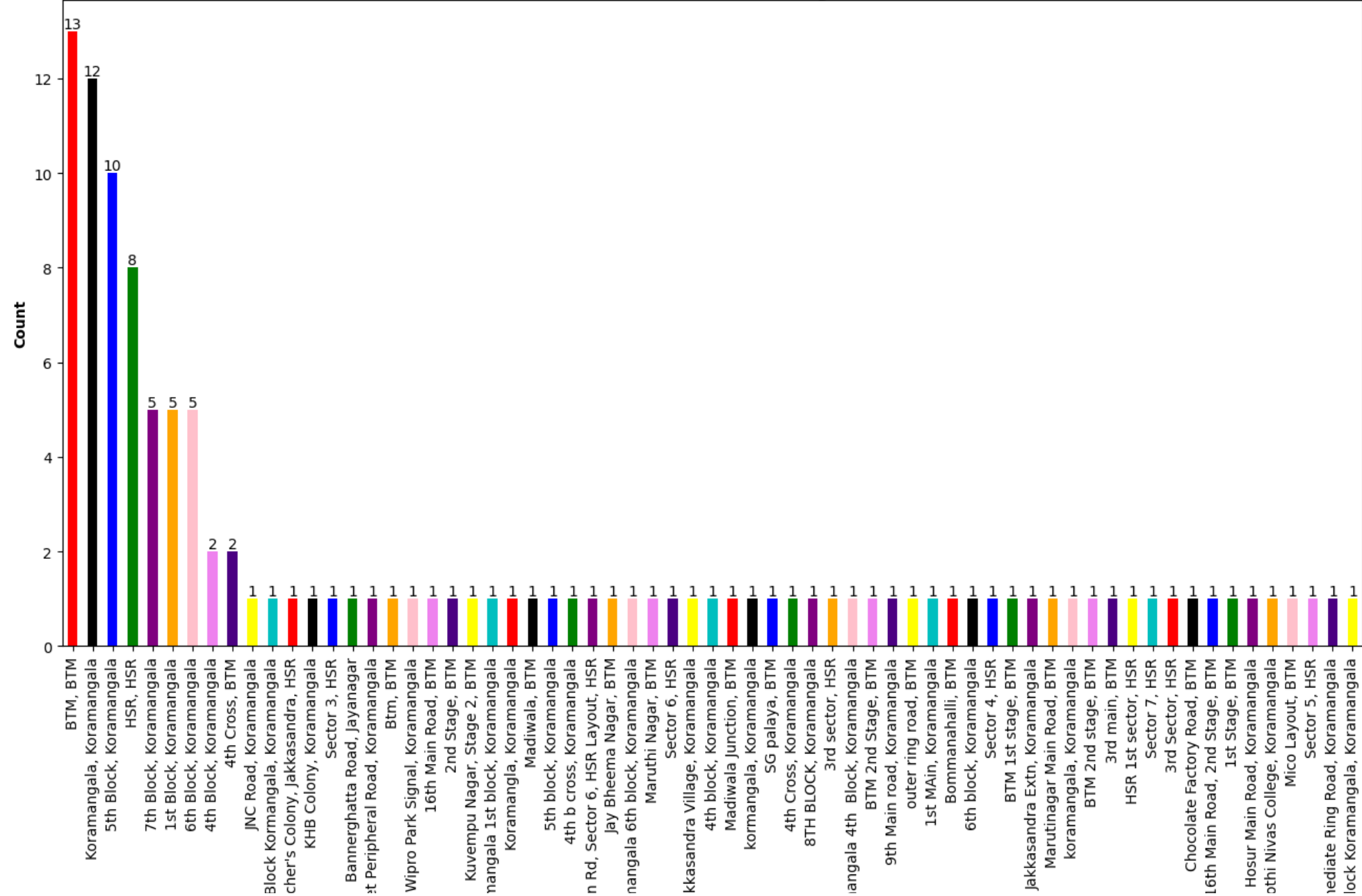
```

```

In [24]: # Plotting the Location distribution
plt.figure(figsize=(16, 8))
location_counts.plot(kind='bar',color=cd)
plt.xlabel('Location', fontweight='bold')
plt.ylabel('Count', fontweight='bold')
plt.title('Outlet Distribution by Location', fontweight='bold',size=24)
plt.xticks(rotation=90)
for i, v in enumerate(location_counts.values):
    plt.annotate(str(v), xy=(i, v),ha='center',color='black',va='bottom')
plt.show()

```

Outlet Distribution by Location



Only 9 Locations have more than 1 Outlets, rest 56 Locations have exactly 1 Outlet.

Top 3 Locations with most number of outlets:

1. BTM, BTM --> 13
2. Koramangala, Koramangala --> 12
3. 5th Block, Koramangala --> 10

3. Rating Analysis

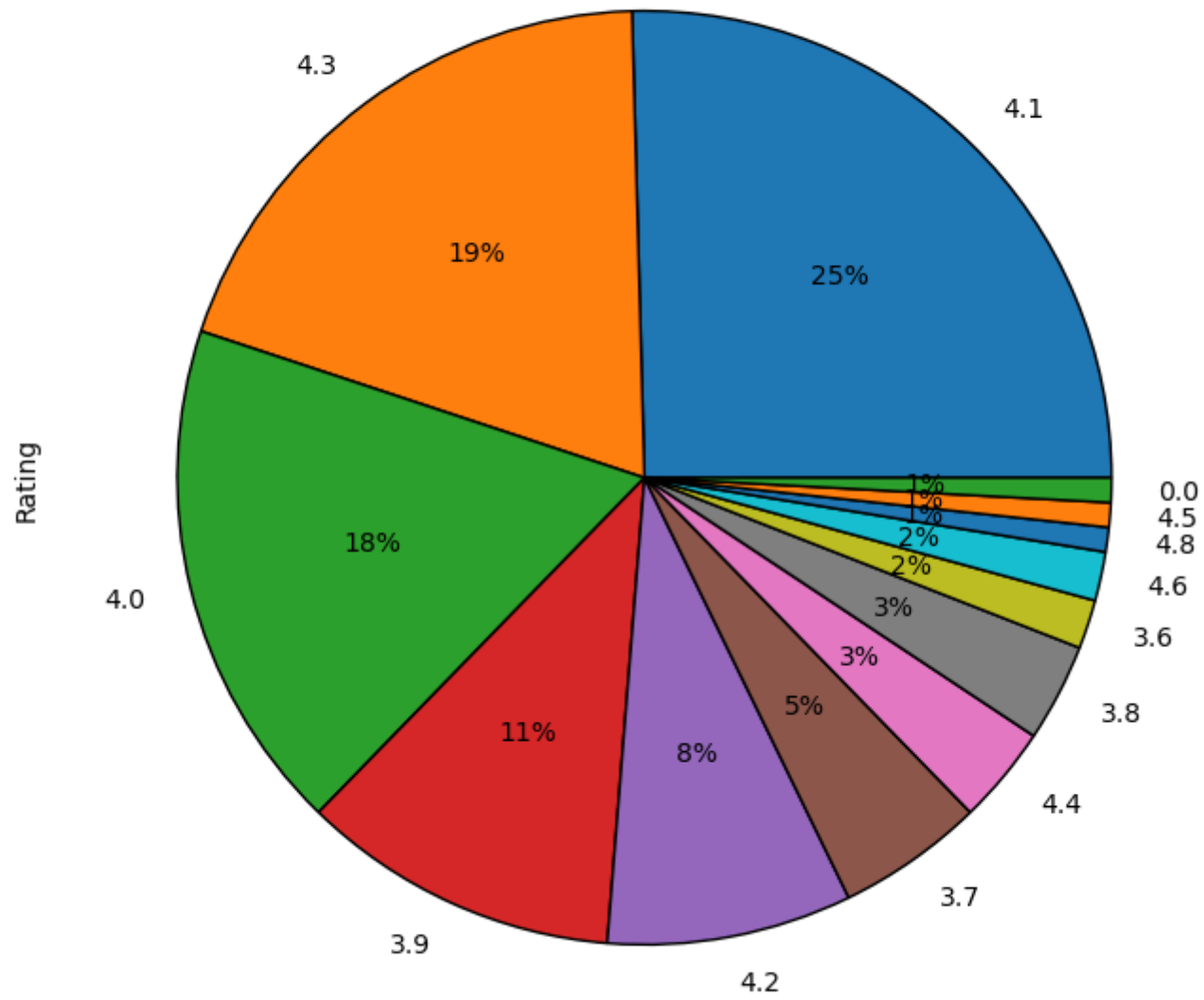
```
In [25]: rating_counts = data['Rating'].value_counts()  
rating_counts
```

```
Out[25]: 4.1    30
         4.3    23
         4.0    21
         3.9    13
         4.2    10
         3.7     6
         4.4     4
         3.8     4
         3.6     2
         4.6     2
         4.8     1
         4.5     1
         0.0     1
         Name: Rating, dtype: int64
```

```
In [26]: average_rating=data['Rating'].mean()
```

```
In [27]: # Plotting the rating distribution
rating_counts.plot.pie(subplots=True, figsize=(14, 8), wedgeprops={"edgecolor": "0", 'linewidth': 1, 'antialiased': True},
autopct='%1.0f%%')
plt.title('Rating Distribution', fontweight=600)
plt.show()
```

Rating Distribution




```
In [28]: # Additional statistics
print('Average Rating:', average_rating)
print('Maximum Rating:', data['Rating'].max())
print('Minimum Rating:', data['Rating'].min())
```

Average Rating: 4.061864406779662

Maximum Rating: 4.8

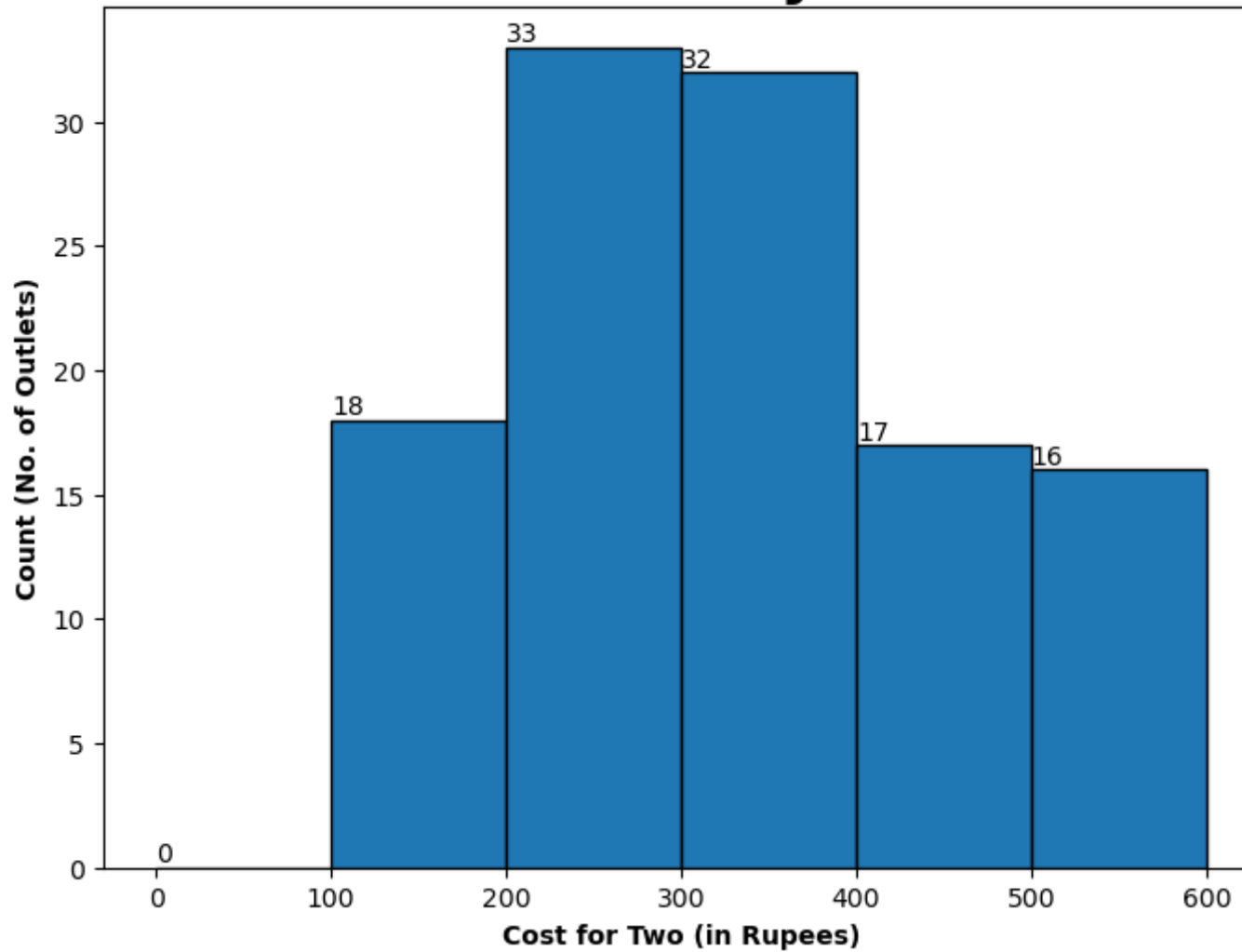
Minimum Rating: 0.0

4. Cost Analysis

```
In [29]: # Cost Analysis
plt.figure(figsize=(8, 6))
bins1=[0,100,200,300,400,500,600]
n, bins, patches = plt.hist(data['Cost for Two'], bins=bins1, edgecolor='k')
plt.xlabel('Cost for Two (in Rupees)', fontweight='bold')
plt.ylabel('Count (No. of Outlets)', fontweight='bold')
plt.title('Cost Analysis', fontweight='bold', size=24)
for i in range(len(patches)):
    plt.text(bins[i], n[i], int(n[i]), ha='left', va='bottom', wrap=True)

plt.show()
```

Cost Analysis



In [30]: *# Additional statistics*

```
print('Maximum Cost:', data['Cost for Two'].max())  
print('Minimum Cost:', data['Cost for Two'].min())  
print('Average Cost:', data['Cost for Two'].mean())
```

Maximum Cost: 800.0
Minimum Cost: 100.0
Average Cost: 321.00847457627117

THANK YOU