

# internship-task

November 23, 2024

## 0.1 Level 1

### Task 1 : Data Exploration and Preprocessing

```
[1]: import pandas as pd
import numpy as np
```

```
[2]: df=pd.read_csv('Dataset .csv')
```

```
[3]: df.head(2)
```

```
[3]: Restaurant ID  Restaurant Name  Country Code      City \
0      6317637  Le Petit Souffle          162  Makati City
1      6304287  Izakaya Kikufuji          162  Makati City

                                Address \
0  Third Floor, Century City Mall, Kalayaan Avenu...
1  Little Tokyo, 2277 Chino Roces Avenue, Legaspi...

                                Locality \
0  Century City Mall, Poblacion, Makati City
1  Little Tokyo, Legaspi Village, Makati City

                                Locality Verbose  Longitude  Latitude \
0  Century City Mall, Poblacion, Makati City, Mak...  121.027535  14.565443
1  Little Tokyo, Legaspi Village, Makati City, Ma...  121.014101  14.553708

                                Cuisines  ...      Currency Has Table booking \
0  French, Japanese, Desserts  ...  Botswana Pula(P)          Yes
1  Japanese  ...  Botswana Pula(P)          Yes

Has Online delivery  Is delivering now  Switch to order menu  Price range \
0          No          No          No          3
1          No          No          No          3

Aggregate rating  Rating color  Rating text  Votes
0          4.8    Dark Green    Excellent    314
1          4.5    Dark Green    Excellent    591
```

[2 rows x 21 columns]

```
[4]: df.shape
```

```
[4]: (9551, 21)
```

```
[5]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9551 entries, 0 to 9550
Data columns (total 21 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Restaurant ID          9551 non-null   int64
1   Restaurant Name        9551 non-null   object
2   Country Code           9551 non-null   int64
3   City                   9551 non-null   object
4   Address                9551 non-null   object
5   Locality               9551 non-null   object
6   Locality Verbose       9551 non-null   object
7   Longitude              9551 non-null   float64
8   Latitude               9551 non-null   float64
9   Cuisines               9542 non-null   object
10  Average Cost for two   9551 non-null   int64
11  Currency               9551 non-null   object
12  Has Table booking      9551 non-null   object
13  Has Online delivery    9551 non-null   object
14  Is delivering now      9551 non-null   object
15  Switch to order menu   9551 non-null   object
16  Price range            9551 non-null   int64
17  Aggregate rating       9551 non-null   float64
18  Rating color           9551 non-null   object
19  Rating text            9551 non-null   object
20  Votes                  9551 non-null   int64
dtypes: float64(3), int64(5), object(13)
memory usage: 1.5+ MB
```

```
[6]: print(f'Number of rows:{df.shape[0]}')
      print(f'Number of columns:{df.shape[1]}')
```

```
Number of rows:9551
Number of columns:21
```

```
[7]: miss_value=df.isnull().sum()
      print("Missing Values:\n",miss_value)
```

```
Missing Values:
```

Restaurant ID	0
Restaurant Name	0
Country Code	0
City	0
Address	0
Locality	0
Locality Verbose	0
Longitude	0
Latitude	0
Cuisines	9
Average Cost for two	0
Currency	0
Has Table booking	0
Has Online delivery	0
Is delivering now	0
Switch to order menu	0
Price range	0
Aggregate rating	0
Rating color	0
Rating text	0
Votes	0

dtype: int64

```
[9]: df['Cuisines'].fillna('Unkown',inplace=True)
```

C:\Users\jeeva\AppData\Local\Temp\ipykernel\_14924\4047093895.py:1:

FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

```
df['Cuisines'].fillna('Unkown',inplace=True)
```

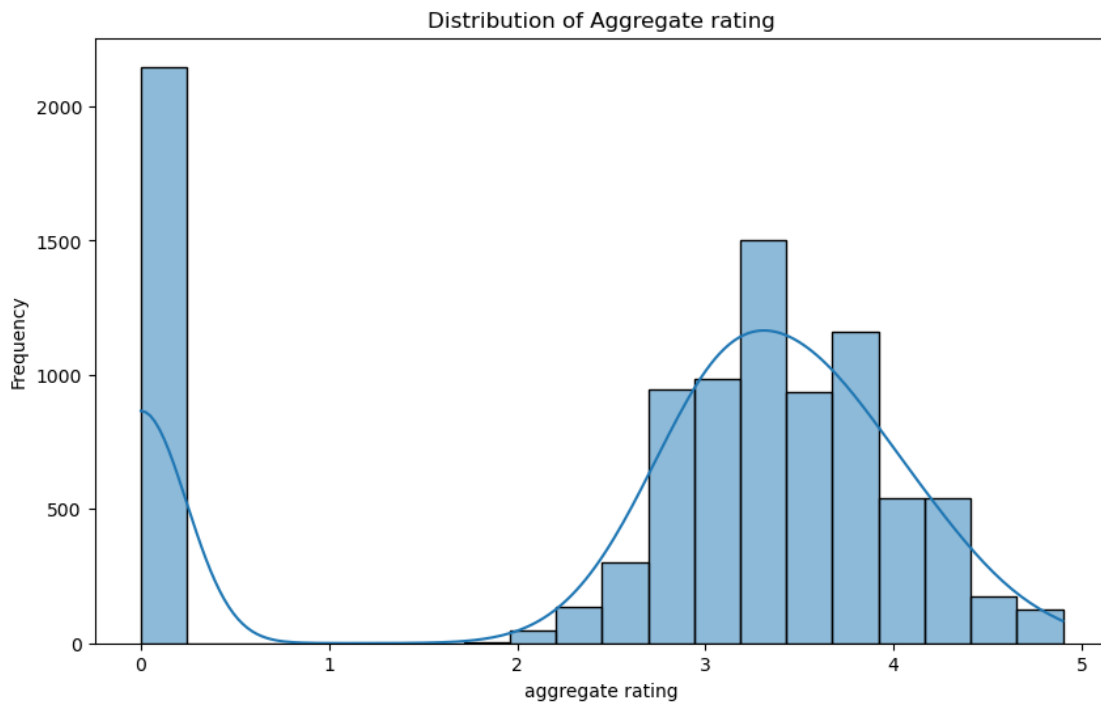
```
[16]: df.isnull().sum()
```

```
[16]: Restaurant ID      0
      Restaurant Name   0
      Country Code     0
      City             0
      Address          0
      Locality         0
```

Locality Verbose	0
Longitude	0
Latitude	0
Cuisines	0
Average Cost for two	0
Currency	0
Has Table booking	0
Has Online delivery	0
Is delivering now	0
Switch to order menu	0
Price range	0
Aggregate rating	0
Rating color	0
Rating text	0
Votes	0
dtype:	int64

```
[18]: import matplotlib.pyplot as plt
import seaborn as sns
```

```
[20]: plt.figure(figsize=(10,6))
sns.histplot(df['Aggregate rating'],bins=20,kde=True)
plt.title('Distribution of Aggregate rating')
plt.xlabel('aggregate rating')
plt.ylabel('Frequency')
plt.show()
```



## Task 2 : Descriptive Analysis

```
[22]: class_count=df['Aggregate rating'].value_counts()  
      print(class_count)
```

```
Aggregate rating  
0.0      2148  
3.2       522  
3.1       519  
3.4       498  
3.3       483  
3.5       480  
3.0       468  
3.6       458  
3.7       427  
3.8       400  
2.9       381  
3.9       335  
2.8       315  
4.1       274  
4.0       266  
2.7       250  
4.2       221  
2.6       191  
4.3       174  
4.4       144  
2.5       110  
4.5        95  
2.4        87  
4.6        78  
4.9        61  
2.3        47  
4.7        42  
2.2        27  
4.8        25  
2.1        15  
2.0         7  
1.9         2  
1.8         1  
Name: count, dtype: int64
```

```
[23]: df.describe()
```

```
[23]:
```

	Restaurant ID	Country Code	Longitude	Latitude \
count	9.551000e+03	9551.000000	9551.000000	9551.000000

mean	9.051128e+06	18.365616	64.126574	25.854381
std	8.791521e+06	56.750546	41.467058	11.007935
min	5.300000e+01	1.000000	-157.948486	-41.330428
25%	3.019625e+05	1.000000	77.081343	28.478713
50%	6.004089e+06	1.000000	77.191964	28.570469
75%	1.835229e+07	1.000000	77.282006	28.642758
max	1.850065e+07	216.000000	174.832089	55.976980

	Average Cost for two	Price range	Aggregate rating	Votes
count	9551.000000	9551.000000	9551.000000	9551.000000
mean	1199.210763	1.804837	2.666370	156.909748
std	16121.183073	0.905609	1.516378	430.169145
min	0.000000	1.000000	0.000000	0.000000
25%	250.000000	1.000000	2.500000	5.000000
50%	400.000000	2.000000	3.200000	31.000000
75%	700.000000	2.000000	3.700000	131.000000
max	800000.000000	4.000000	4.900000	10934.000000

```
[24]: country_distribution=df['Country Code'].value_counts()
print(country_distribution)
```

```
Country Code
1      8652
216    434
215     80
30      60
214     60
189     60
148     40
208     34
14      24
162     22
94      21
184     20
166     20
191     20
37       4
Name: count, dtype: int64
```

```
[25]: city_distribution=df['City'].value_counts()
print(city_distribution)
```

```
City
New Delhi      5473
Gurgaon        1118
Noida          1080
Faridabad       251
Ghaziabad       25
```

```

Panchkula      1
Mc Millan      1
Mayfield       1
Macedon        1
Vineland Station 1
Name: count, Length: 141, dtype: int64

```

```
[26]: cuisine_distribution=df['Cuisines'].value_counts()
print(cuisine_distribution)
```

```

Cuisines
North Indian      936
North Indian, Chinese 511
Chinese           354
Fast Food         354
North Indian, Mughlai 334
...
Bengali, Fast Food 1
North Indian, Rajasthani, Asian 1
Chinese, Thai, Malaysian, Indonesian 1
Bakery, Desserts, North Indian, Bengali, South Indian 1
Italian, World Cuisine 1
Name: count, Length: 1826, dtype: int64

```

```
[27]: print("Top Cuisines:")
print(cuisine_distribution.head(10))
```

```

Top Cuisines:
Cuisines
North Indian      936
North Indian, Chinese 511
Chinese           354
Fast Food         354
North Indian, Mughlai 334
Cafe              299
Bakery            218
North Indian, Mughlai, Chinese 197
Bakery, Desserts  170
Street Food       149
Name: count, dtype: int64

```

```
[28]: print("Top Citis:")
print(city_distribution.head(10))
```

```

Top Citis:
City
New Delhi      5473

```

Gurgaon	1118
Noida	1080
Faridabad	251
Ghaziabad	25
Bhubaneswar	21
Amritsar	21
Ahmedabad	21
Lucknow	21
Guwahati	21

Name: count, dtype: int64

### Task 3 : Geospatial Analysis

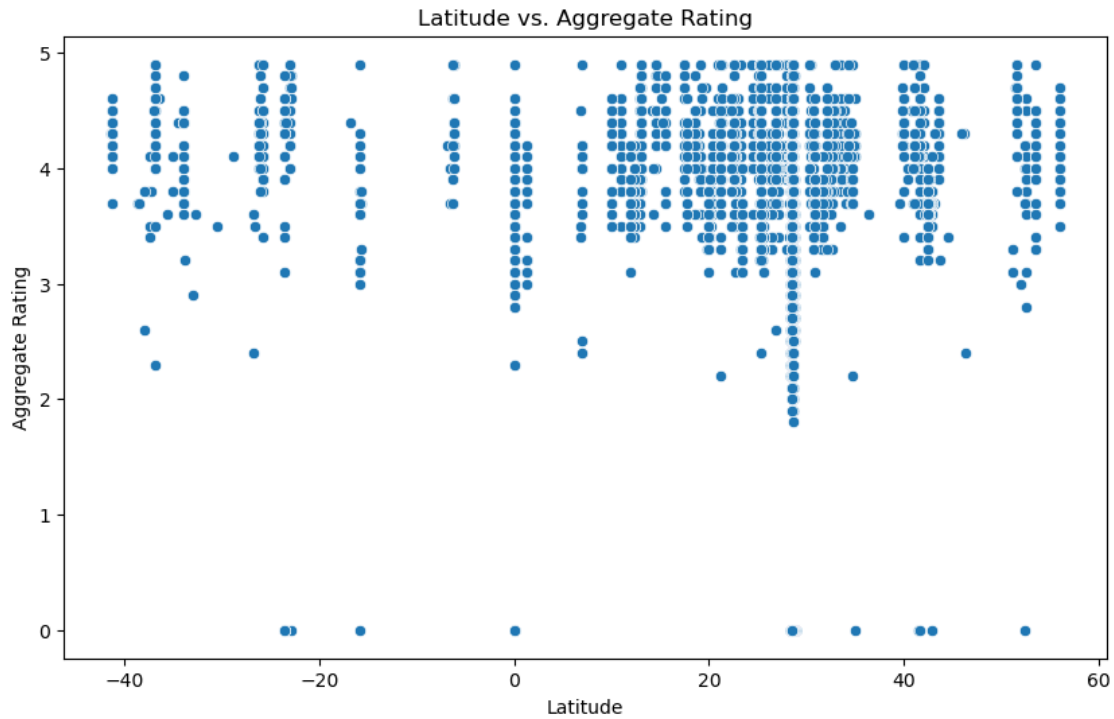
```
[30]: import folium
map_center=[df['Latitude'].mean(),df['Longitude'].mean()]
restaurant_map=folium.Map(location=map_center,zoom_start=12)

for _,row in df.iterrows():
    folium.
        ↪Marker(location=[row['Latitude'],row['Longitude']],popup=row['Restaurant_
        ↪Name'],).add_to(restaurant_map)
```

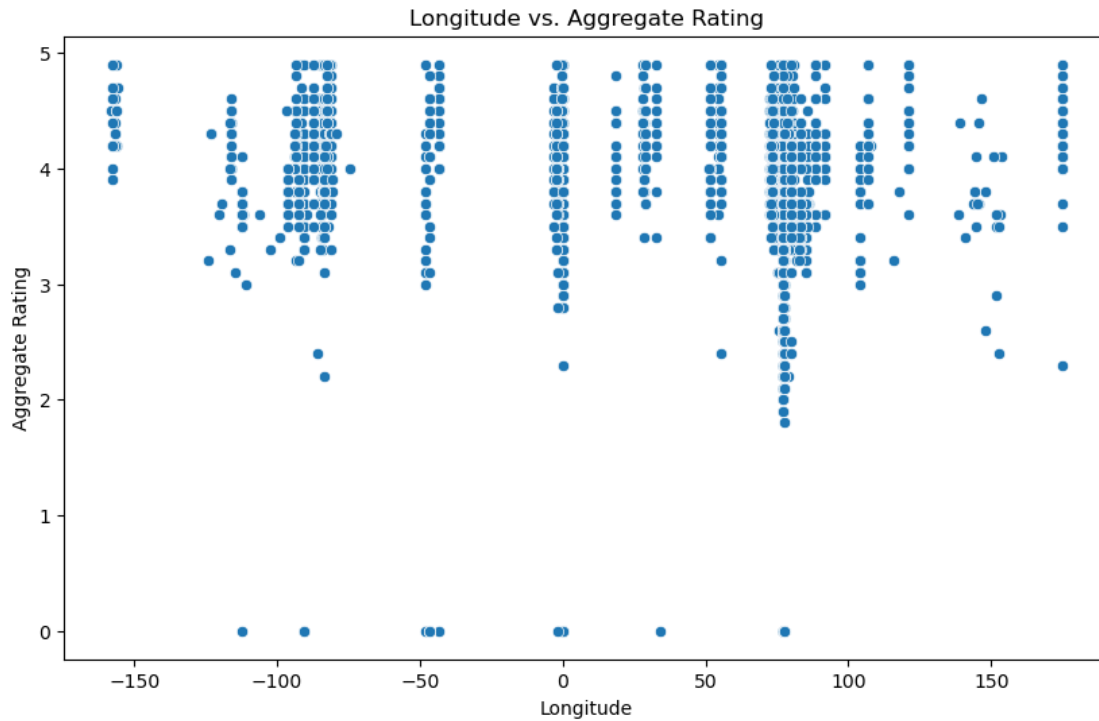
```
[31]: restaurant_map.save('restaurant_map.html')
```

```
[32]: plt.figure(figsize=(10,6))
sns.scatterplot(data=df,x='Latitude',y='Aggregate rating')
plt.title('Latitude vs. Aggregate Rating')
plt.xlabel('Latitude')
plt.ylabel('Aggregate Rating')
plt.show()
```





```
[33]: plt.figure(figsize=(10,6))
sns.scatterplot(data=df,x='Longitude',y='Aggregate rating')
plt.title('Longitude vs. Aggregate Rating')
plt.xlabel('Longitude')
plt.ylabel('Aggregate Rating')
plt.show()
```



```
[34]: corr_matrix=df[['Latitude','Longitude','Aggregate rating']].corr()
      print(corr_matrix)
```

	Latitude	Longitude	Aggregate rating
Latitude	1.000000	0.043207	0.000516
Longitude	0.043207	1.000000	-0.116818
Aggregate rating	0.000516	-0.116818	1.000000

## 0.2 Level 2

### Task 1 : Table Booking and Online Delivery

```
[38]: tabel_booking_percentage=(df['Has Table booking'].
      ↪value_counts(normalize=True)*100).get('yes',0)
      print(f'Percentage of restaurant offering booking:{tabel_booking_percentage:.
      ↪2f}%')
```

Percentage of restaurant offering booking:0.00%

```
[39]: online_percentage=(df['Has Online delivery'].value_counts(normalize=True)*100).
      ↪get('yes',0)
      print(f'Percentage of restaurant offering booking:{online_percentage:.2f}%')
```

Percentage of restaurant offering booking:0.00%

## Task 2 : Price Range Analysis

```
[49]: ava_rating_with_booking=df[df['Has Table booking']=='Yes']['Aggregate rating'].  
      ↪mean()  
      ava_rating_without_booking=df[df['Has Table booking']=='No']['Aggregate_  
      ↪rating'].mean()
```

```
[50]: print(f'Avarage rating of with table booking:{ava_rating_with_booking:.2f}')  
      print(f'Avarage rating of without table booking:{ava_rating_without_booking:.  
      ↪2f}')
```

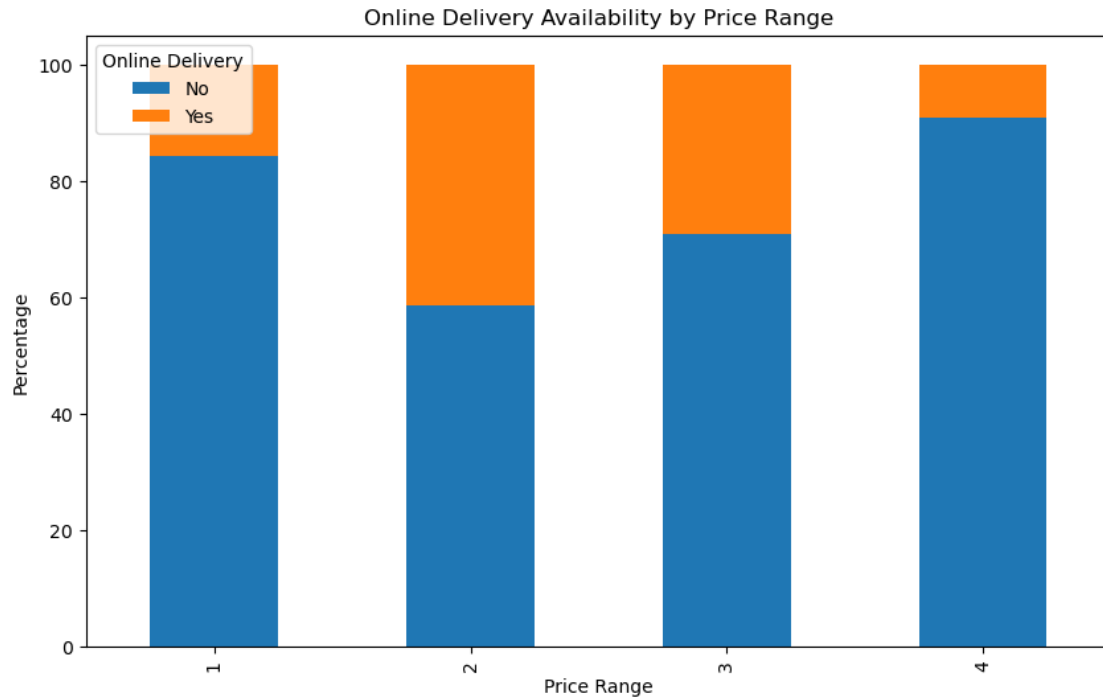
Avarage rating of with table booking:3.44  
Avarage rating of without table booking:2.56

```
[51]: online_delivery_price=df.groupby('Price range')['Has Online delivery'].  
      ↪value_counts(normalize=True).unstack().fillna(0)*100
```

```
[52]: print(online_delivery_price)
```

Has Online delivery	No	Yes
Price range		
1	84.225923	15.774077
2	58.689367	41.310633
3	70.809659	29.190341
4	90.955631	9.044369

```
[53]: online_delivery_price.plot(kind='bar',stacked=True,figsize=(10,6))  
      plt.title('Online Delivery Availability by Price Range')  
      plt.xlabel("Price Range")  
      plt.ylabel('Percentage')  
      plt.legend(title='Online Delivery',loc='upper left')  
      plt.show()
```



```
[68]: df.head(2)
```

```
[68]:
```

	Restaurant ID	Restaurant Name	Country Code	City \
0	6317637	Le Petit Souffle	162	Makati City
1	6304287	Izakaya Kikufuji	162	Makati City

	Address \
0	Third Floor, Century City Mall, Kalayaan Avenu...
1	Little Tokyo, 2277 Chino Roces Avenue, Legaspi...

	Locality \
0	Century City Mall, Poblacion, Makati City
1	Little Tokyo, Legaspi Village, Makati City

	Locality Verbose	Longitude	Latitude \
0	Century City Mall, Poblacion, Makati City, Mak...	121.027535	14.565443
1	Little Tokyo, Legaspi Village, Makati City, Ma...	121.014101	14.553708

	Cuisines ...	Currency	Has Table booking \
0	French, Japanese, Desserts ...	Botswana Pula(P)	Yes
1	Japanese ...	Botswana Pula(P)	Yes

	Has Online delivery	Is delivering now	Switch to order menu	Price range \
0	No	No	No	3

1	No	No	No	3
---	----	----	----	---

	Aggregate rating	Rating color	Rating text	Votes
0	4.8	Dark Green	Excellent	314
1	4.5	Dark Green	Excellent	591

[2 rows x 21 columns]

### Task 3 : Feature Engineering

```
[70]: df['Restaturant_Name_length']=df['Restaurant Name'].apply(len)
```

```
[76]: df['Address_length']=df['Address'].apply(len)
```

```
[84]: df[['Restaurant Name','Restaturant_Name_length','Address','Address_length']].
      ↪head()
```

```
[84]:
```

	Restaurant Name	Restaturant_Name_length	\
0	Le Petit Souffle	16	
1	Izakaya Kikufuji	16	
2	Heat - Edsa Shangri-La	22	
3	Ooma	4	
4	Sambo Kojin	11	

	Address	Address_length
0	Third Floor, Century City Mall, Kalayaan Avenu...	71
1	Little Tokyo, 2277 Chino Roces Avenue, Legaspi...	67
2	Edsa Shangri-La, 1 Garden Way, Ortigas, Mandal...	56
3	Third Floor, Mega Fashion Hall, SM Megamall, O...	70
4	Third Floor, Mega Atrium, SM Megamall, Ortigas...	64

```
[90]: df['Has_Table_Booking']=df['Has Table booking'].apply(lambda x: 1 if x == 'Yes'
      ↪else 0)
df['Has_Online_Delivery']=df['Has Online delivery'].apply(lambda x: 1 if x ==
      ↪'Yes' else 0)
```

```
[92]: df[['Has Table booking','Has_Table_Booking','Has Online_
      ↪delivery','Has_Online_Delivery']].head()
```

```
[92]:
```

	Has Table booking	Has_Table_Booking	Has Online delivery	\
0	Yes	1	No	
1	Yes	1	No	
2	Yes	1	No	
3	No	0	No	
4	Yes	1	No	

Has\_Online\_Delivery

0	0
1	0
2	0
3	0
4	0

### 0.3 Level 3

#### Task 1: Predictive Modeling

```
[102]: feature=['Restaturant_Name_length','Address_length','Has_Table_Booking','Has_Online_Delivery']
       target = ['Aggregate rating']
```

```
[104]: X=df[feature]
       y=df[target]
```

```
[106]: from sklearn.model_selection import train_test_split
```

```
[108]: X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.
       ↪2,random_state=42)
```

#### Linear Regression

```
[111]: from sklearn.linear_model import LinearRegression
       from sklearn.metrics import mean_absolute_error, mean_squared_error,r2_score
```

```
[113]: lr=LinearRegression()
       lr.fit(X_train,y_train)
```

```
[113]: LinearRegression()
```

```
[115]: y_pred_lr=lr.predict(X_test)
```

```
[117]: mae_lr=mean_absolute_error(y_test,y_pred_lr)
       msse_lr=mean_squared_error(y_test,y_pred_lr)
       r2_lr=r2_score(y_test,y_pred_lr)
```

```
[119]: print(f"Linear Regression - MAE:{mae_lr}, MSE:{msse_lr},R2:{r2_lr}")
```

```
Linear Regression - MAE:1.193199733256996,
MSE:2.1032980983357454,R2:0.07592382118320307
```

#### Decision Tree

```
[125]: from sklearn.tree import DecisionTreeRegressor
```

```
[127]: dt=DecisionTreeRegressor()
       dt.fit(X_train,y_train)
```

```
[127]: DecisionTreeRegressor()
```

```
[129]: y_pred_dt=dt.predict(X_test)
```

```
[131]: mae_dt=mean_absolute_error(y_test,y_pred_dt)
msse_dt=mean_squared_error(y_test,y_pred_dt)
r2_dt=r2_score(y_test,y_pred_dt)
```

```
[133]: print(f"DecisionTreeRegressor - MAE:{mae_dt}, MSE:{msse_dt},R2:{r2_dt}")
```

```
DecisionTreeRegressor - MAE:1.3223126575185862,
MSE:3.0066380393776058,R2:-0.32095521443759667
```

### Random Forest

```
[136]: from sklearn.ensemble import RandomForestRegressor
```

```
[138]: rf=RandomForestRegressor(random_state=42)
rf.fit(X_train,y_train)
```

```
C:\Users\jeeva\anaconda3\Lib\site-packages\sklearn\base.py:1474:
DataConversionWarning: A column-vector y was passed when a 1d array was
expected. Please change the shape of y to (n_samples,), for example using
ravel().
```

```
return fit_method(estimator, *args, **kwargs)
```

```
[138]: RandomForestRegressor(random_state=42)
```

```
[140]: y_pred_rf=rf.predict(X_test)
```

```
[142]: mae_rf=mean_absolute_error(y_test,y_pred_rf)
msse_rf=mean_squared_error(y_test,y_pred_rf)
r2_rf=r2_score(y_test,y_pred_rf)
```

```
[144]: print(f"Random Forest - MAE:{mae_rf}, MSE:{msse_rf},R2:{r2_rf}")
```

```
Random Forest - MAE:1.2526774928676685,
MSE:2.5957098373762104,R2:-0.14041544074885026
```

### Task 2: Customer Preference Analysis

```
[149]: cuisine_rating=df.groupby('Cuisines')['Aggregate rating'].mean().
↪sort_values(ascending=False)
```

```
[155]: cuisine_rating.head(10)
```

```
[155]: Cuisines
Continental, Indian      4.9
BBQ, Breakfast, Southern 4.9
```

Italian, Deli	4.9
American, Caribbean, Seafood	4.9
Burger, Bar Food, Steak	4.9
American, Burger, Grill	4.9
Italian, Bakery, Continental	4.9
European, Asian, Indian	4.9
European, Contemporary	4.9
American, Coffee and Tea	4.9

Name: Aggregate rating, dtype: float64

```
[157]: cuisine_votes = df.groupby('Cuisines')['Votes'].sum().
      ↪sort_values(ascending=False)
```

```
[161]: cuisine_votes.head(10)
```

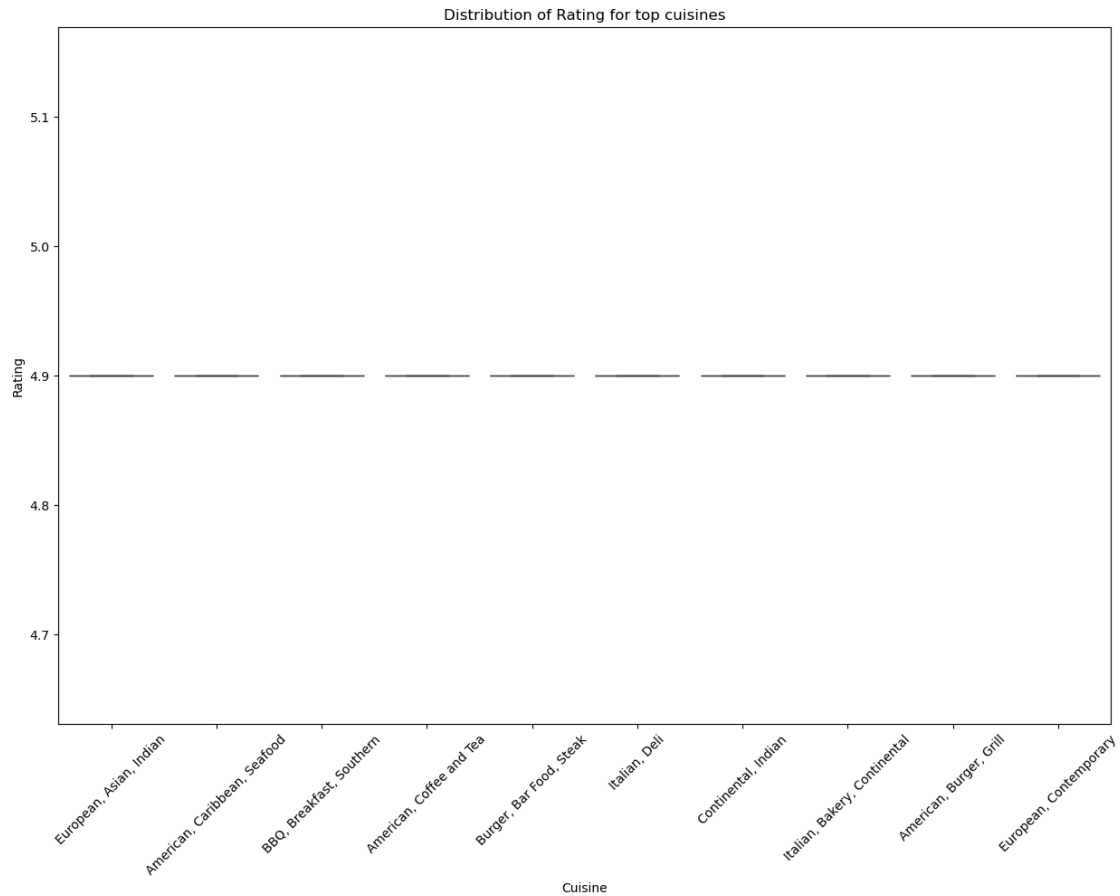
```
[161]: Cuisines
North Indian, Mughlai      53747
North Indian               46241
North Indian, Chinese     42012
Cafe                      30657
Chinese                   21925
North Indian, Mughlai, Chinese 20115
Fast Food                 17852
South Indian              16433
Mughlai, North Indian     15275
Italian                   14799
Name: Votes, dtype: int64
```

```
[163]: top_cuisine=cuisine_rating.head(10).index
```

```
[165]: df_top_cusines=df[df['Cuisines'].isin(top_cuisine)]
```

```
[169]: plt.figure(figsize=(15,10))
sns.boxplot(x='Cuisines',y='Aggregate rating', data=df_top_cusines)
plt.xticks(rotation=45)
plt.title('Distribution of Rating for top cuisines')
plt.xlabel('Cuisine')
plt.ylabel('Rating')
plt.show()
```



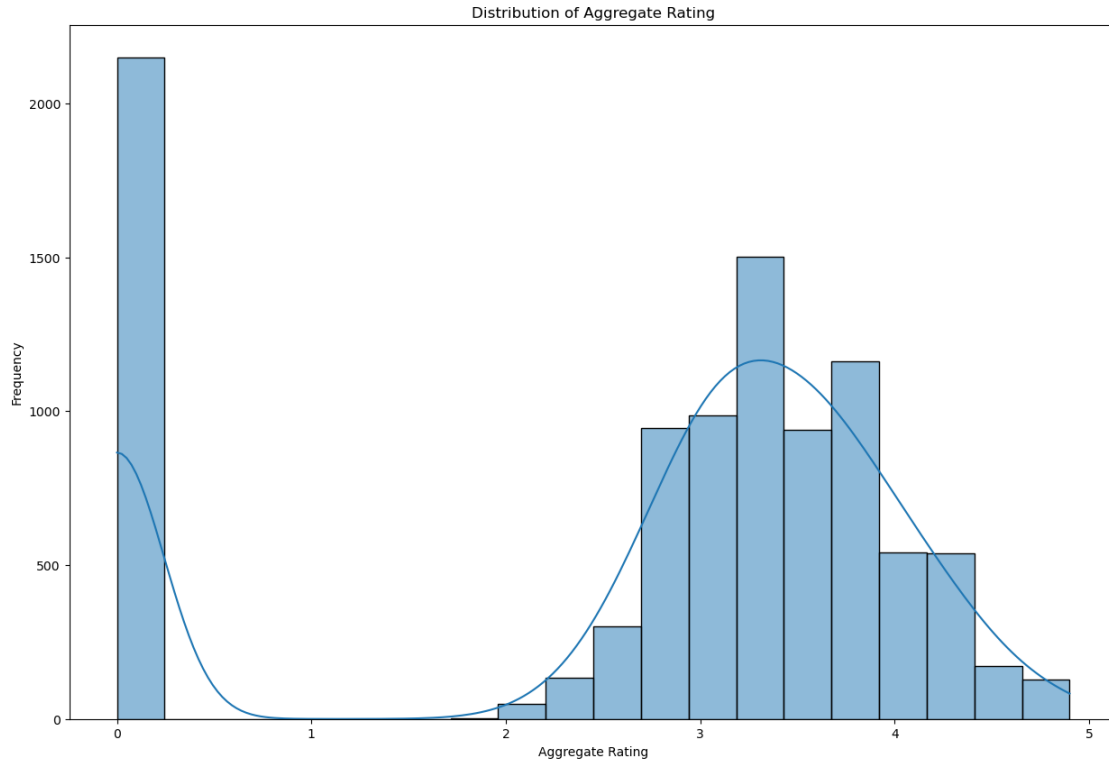


## Task 3: Data Visualization

### 1. Distribution of Rating

#### Histogram of Rating

```
[174]: plt.figure(figsize=(15,10))
sns.histplot(df['Aggregate rating'],bins=20,kde=True)
plt.title('Distribution of Aggregate Rating')
plt.xlabel('Aggregate Rating')
plt.ylabel('Frequency')
plt.show()
```



### Bar Plot of Rating

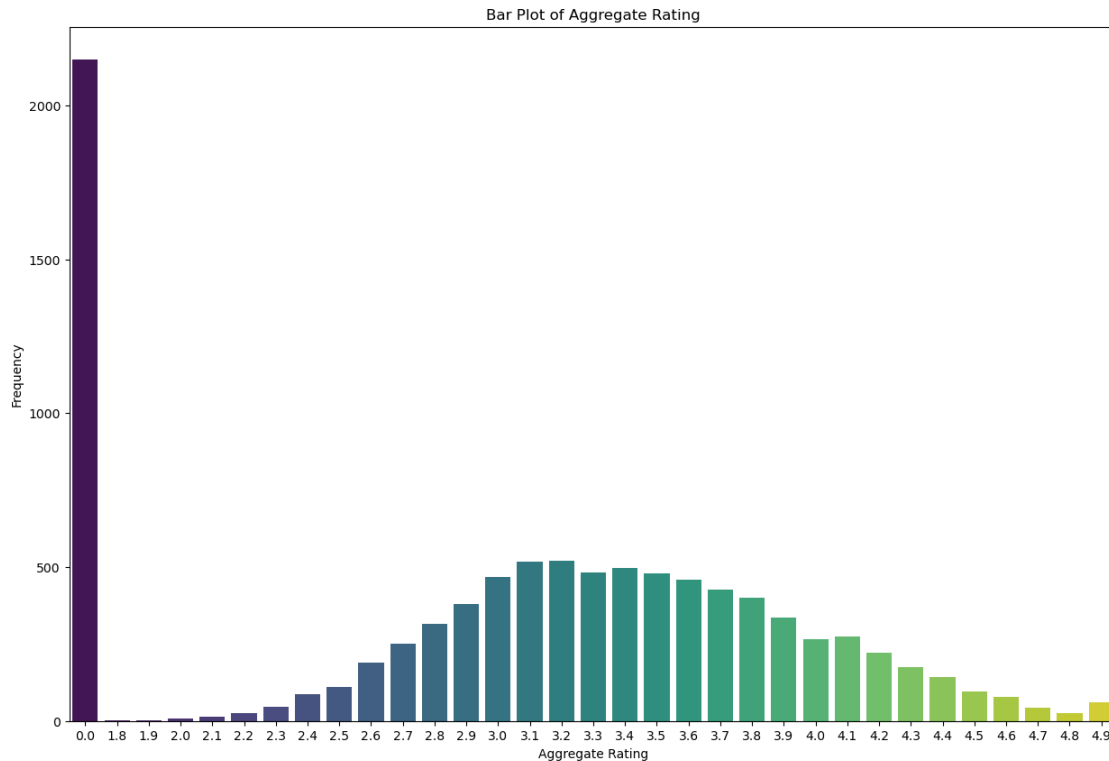
```
[179]: rating_counts=df['Aggregate rating'].value_counts().sort_index()
```

```
[181]: plt.figure(figsize=(15,10))
sns.barplot(x=rating_counts.index,y=rating_counts.values,palette='viridis')
plt.title('Bar Plot of Aggregate Rating')
plt.xlabel('Aggregate Rating')
plt.ylabel('Frequency')
plt.show()
```

C:\Users\jeeva\AppData\Local\Temp\ipykernel\_14924\3444833412.py:2:  
FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(x=rating_counts.index,y=rating_counts.values,palette='viridis')
```



### Average Rating by Cuisine

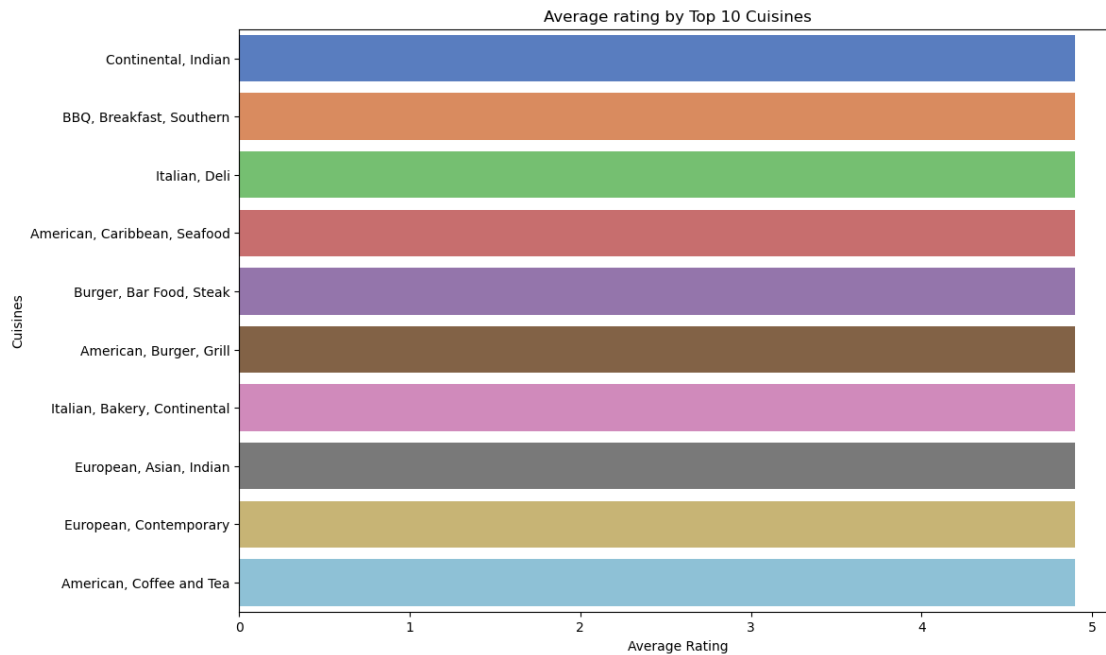
```
[184]: ava_rating_cuisine=df.groupby('Cuisines')['Aggregate rating'].mean().
        ↪sort_values(ascending=False).head(10)
```

```
[188]: plt.figure(figsize=(12,8))
sns.barplot(y=ava_rating_cuisine.index,x=ava_rating_cuisine.
        ↪values,palette='muted')
plt.title('Average rating by Top 10 Cuisines')
plt.xlabel('Average Rating')
plt.ylabel('Cuisines')
plt.show()
```

C:\Users\jeeva\AppData\Local\Temp\ipykernel\_14924\106019245.py:2: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(y=ava_rating_cuisine.index,x=ava_rating_cuisine.values,palette='muted')
```



### Average Rating by city

```
[191]: ava_rating_city = df.groupby('City')['Aggregate rating'].mean().
      ↪ sort_values(ascending=False).head(10)
```

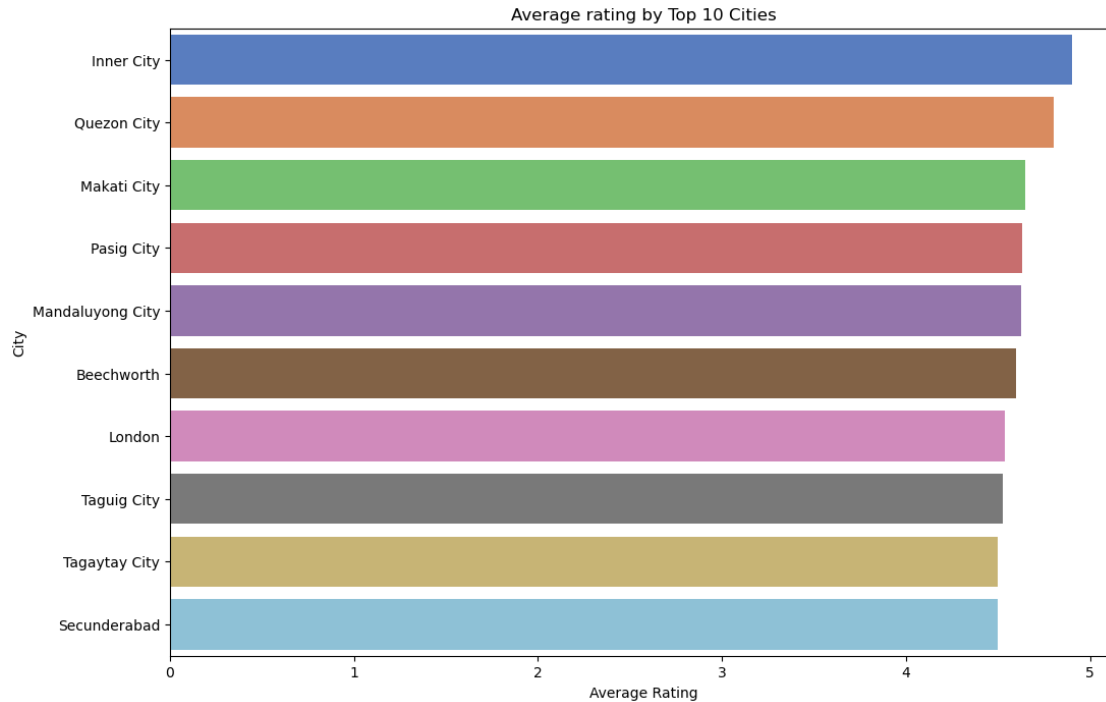
```
[193]: plt.figure(figsize=(12,8))
sns.barplot(y=ava_rating_city.index,x=ava_rating_city.values,palette='muted')
plt.title('Average rating by Top 10 Cities')
plt.xlabel('Average Rating')
plt.ylabel('City')
plt.show()
```

C:\Users\jeeva\AppData\Local\Temp\ipykernel\_14924\2114028040.py:2:

FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(y=ava_rating_city.index,x=ava_rating_city.values,palette='muted')
```



### Scatter plot: Restaurant Name Length vs. Aggregate Rating

```
[200]: df.head(3)
```

```
[200]:
```

	Restaurant ID	Restaurant Name	Country Code	City \
0	6317637	Le Petit Souffle	162	Makati City
1	6304287	Izakaya Kikufuji	162	Makati City
2	6300002	Heat - Edsa Shangri-La	162	Mandaluyong City

	Address \
0	Third Floor, Century City Mall, Kalayaan Avenu...
1	Little Tokyo, 2277 Chino Roces Avenue, Legaspi...
2	Edsa Shangri-La, 1 Garden Way, Ortigas, Mandal...

	Locality \
0	Century City Mall, Poblacion, Makati City
1	Little Tokyo, Legaspi Village, Makati City
2	Edsa Shangri-La, Ortigas, Mandaluyong City

	Locality Verbose	Longitude	Latitude \
0	Century City Mall, Poblacion, Makati City, Mak...	121.027535	14.565443
1	Little Tokyo, Legaspi Village, Makati City, Ma...	121.014101	14.553708
2	Edsa Shangri-La, Ortigas, Mandaluyong City, Ma...	121.056831	14.581404

	Cuisines ...	Switch to order menu	Price range \
--	--------------	----------------------	---------------

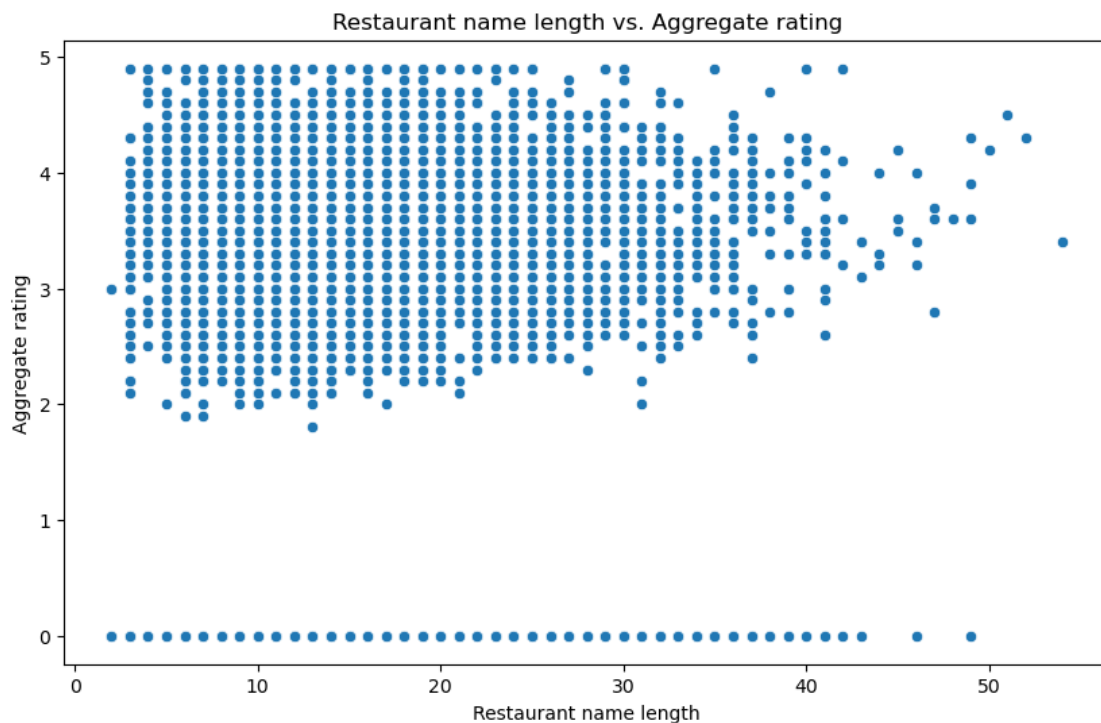
0	French, Japanese, Desserts ...	No	3
1	Japanese ...	No	3
2	Seafood, Asian, Filipino, Indian ...	No	4

	Aggregate rating	Rating	color	Rating text	Votes	Restaturant_Name_length \
0	4.8	Dark	Green	Excellent	314	16
1	4.5	Dark	Green	Excellent	591	16
2	4.4	Green		Very Good	270	22

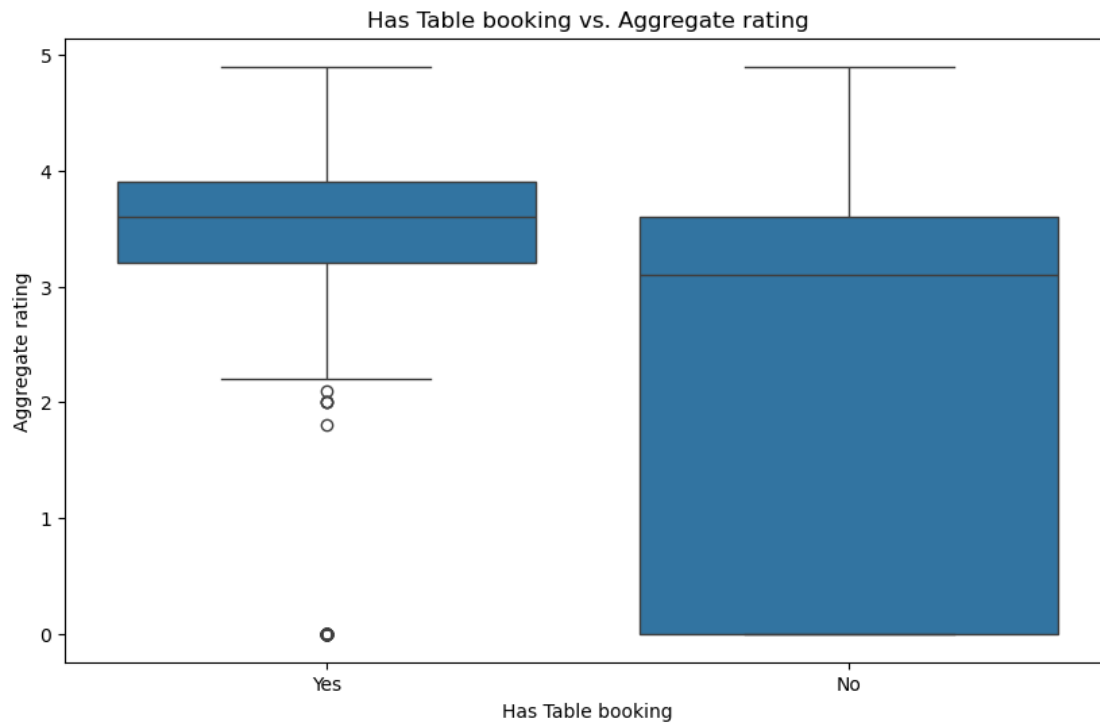
	Address_length	Has_Table_Booking	Has_Online_Delivery
0	71	1	0
1	67	1	0
2	56	1	0

[3 rows x 25 columns]

```
[202]: plt.figure(figsize=(10,6))
sns.scatterplot(data=df,x='Restaturant_Name_length',y='Aggregate rating')
plt.title('Restaurant name length vs. Aggregate rating')
plt.xlabel('Restaurant name length')
plt.ylabel('Aggregate rating')
plt.show()
```



```
[206]: plt.figure(figsize=(10,6))
sns.boxplot(data=df,x='Has Table booking',y='Aggregate rating')
plt.title('Has Table booking vs. Aggregate rating')
plt.xlabel('Has Table booking')
plt.ylabel('Aggregate rating')
plt.show()
```



```
[ ]:
```