Level 3

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model selection import train test split
from sklearn.linear model import LinearRegression
from sklearn.tree import DecisionTreeRegressor
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean squared error, r2 score
df = pd.read csv("Dataset .csv")
df
      Restaurant ID
                               Restaurant Name Country Code
City
            6317637
                              Le Petit Souffle
                                                          162
Makati City
            6304287
                              Izakaya Kikufuji
                                                          162
Makati City
            6300002
                        Heat - Edsa Shangri-La
                                                          162
Mandaluyong City
            6318506
                                          0oma
                                                          162
Mandaluyong City
                                   Sambo Kojin
                                                          162
            6314302
Mandaluyong City
. . .
                                   Naml \ Gurme
                                                          208
9546
            5915730
@rstanbul
9547
            5908749
                                  Ceviz Aûûac N
                                                      208
@ @ stanbul
9548
            5915807
                                                          208
                                         Hugga
@
@
g
s
t
anbul
9549
            5916112
                                   A@@k Kahve
                                                          208
@rstanbul
9550
            5927402 Walter's Coffee Roastery
                                                          208
@rstanbul
                                                  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...
3
      Third Floor, Mega Fashion Hall, SM Megamall, O...
4
      Third Floor, Mega Atrium, SM Megamall, Ortigas...
. . .
      Kemanke�� Karamustafa Pa��a Mahallesi, R\ht\m ...
9546
      Ko@@uyolu Mahallesi, Muhittin @@st@ nda@@ Cadd...
9547
```

```
9548
      Kuru@ e@@me Mahallesi, Muallim Naci Caddesi, N...
9549
      Kuru@ e@@me Mahallesi, Muallim Naci Caddesi, N...
9550
      Cafea@@a Mahallesi, Bademalt \ Sokak, No 21/B, ...
                                         Locality \
0
       Century City Mall, Poblacion, Makati City
1
      Little Tokyo, Legaspi Village, Makati City
2
      Edsa Shangri-La, Ortigas, Mandaluyong City
3
          SM Megamall, Ortigas, Mandaluyong City
4
          SM Megamall, Ortigas, Mandaluyong City
9546
                                         Karak🖟 y
9547
                                        Ko@@uyolu
                                      Kuruû eûûme
9548
9549
                                      Kuru@ e@@me
                                             Moda
9550
                                        Locality Verbose
                                                           Longitude \
0
      Century City Mall, Poblacion, Makati City, Mak...
                                                          121.027535
1
      Little Tokyo, Legaspi Village, Makati City, Ma...
                                                          121.014101
2
      Edsa Shangri-La, Ortigas, Mandaluyong City, Ma...
                                                          121.056831
      SM Megamall, Ortigas, Mandaluyong City, Mandal...
3
                                                          121.056475
      SM Megamall, Ortigas, Mandaluyong City, Mandal...
4
                                                          121.057508
9546
                                     Karakû y, ûûstanbul
                                                           28.977392
                                    Ko@@uyolu, @@stanbul
9547
                                                           29.041297
9548
                                  Kuru@ e@@me, @@stanbul
                                                           29.034640
9549
                                  Kuru@ e@@me, @@stanbul
                                                           29.036019
9550
                                         Moda, 🚱 stanbul
                                                           29.026016
       Latitude
                                          Cuisines ...
Currency \
      14.565443
                       French, Japanese, Desserts
                                                         Botswana
Pula(P)
      14.553708
                                                         Botswana
                                          Japanese
Pula(P)
                 Seafood, Asian, Filipino, Indian
      14.581404
                                                         Botswana
Pula(P)
      14.585318
                                   Japanese, Sushi
                                                         Botswana
Pula(P)
      14.584450
                                  Japanese, Korean ...
                                                         Botswana
Pula(P)
9546 41.022793
                                           Turkish ...
                                                         Turkish
Lira(TL)
9547 41.009847
                  World Cuisine, Patisserie, Cafe
                                                         Turkish
Lira(TL)
9548 41.055817
                           Italian, World Cuisine ...
                                                         Turkish
```

Lira 9549 Lira	41.057979		Restaura	nt Cafe	Turkish
	40.984776			Cafe	Turkish
0 1 2 3 4	Has Table book	cing Has Or Yes Yes Yes No Yes	nline delivery No No No No No	Is deliverin	ng now \ No No No No No No No
9546 9547 9548 9549 9550		No No No No No	No No No No No		No No No No No
\	Switch to orde	er menu Pri	.ce range Agg	regate rating	Rating color
ò		No	3	4.8	B Dark Green
1		No	3	4.5	Dark Green
2		No	4	4.4	Green
3		No	4	4.9	Dark Green
4		No	4	4.8	B Dark Green
9546		No	3	4.1	. Green
9547		No	3	4.2	green
9548		No	4	3.7	Yellow
9549		No	4	4.0	Green
9550		No	2	4.6	Green
0 1 2 3 4	Rating text Vo Excellent Excellent Very Good Excellent Excellent	otes 314 591 270 365 229			

```
9546
       Very Good
                   788
9547
       Very Good
                  1034
9548
            Good
                   661
9549
       Very Good
                   901
9550
       Very Good
                   591
[9551 rows x 21 columns]
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
                           9551 non-null
    Longitude
                                            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
                                            obiect
 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
#Drop rows with missing values
df.dropna(inplace=True)
#After handling missing values
df.shape
(9542, 21)
```

Task 1: Predictive Modeling

```
# Considering some potential features
selected features = ['Average Cost for two', 'Price range', 'Votes']
X = df[selected features]
y = df['Aggregate rating']
# Split the dataset into training and testing sets (80% train, 20%
test)
X train, X test, y train, y test = train test split(X, y,
test size=0.2, random state=42)
# Build the regression model (Linear Regression as an example)
model = LinearRegression()
model.fit(X train, y train)
LinearRegression()
# Predict aggregate ratings on the test set
v pred = model.predict(X test)
# Evaluate the model
mse = mean_squared_error(y_test, y_pred)
rmse = mean squared error(y test, y pred, squared=False)
r2 = r2 score(y test, y pred)
# Mean Squared Error
mse
1.7193421524563215
# Root Mean Squared Error
rmse
1.3112368788500122
# R^2 Score
0.24920612400080278
# Initialize and train different regression models
models = {
    'Linear Regression': LinearRegression(),
    'Decision Tree': DecisionTreeRegressor(random state=42),
    'Random Forest': RandomForestRegressor(n estimators=100,
random state=42)
results = {}
for name, model in models.items():
```

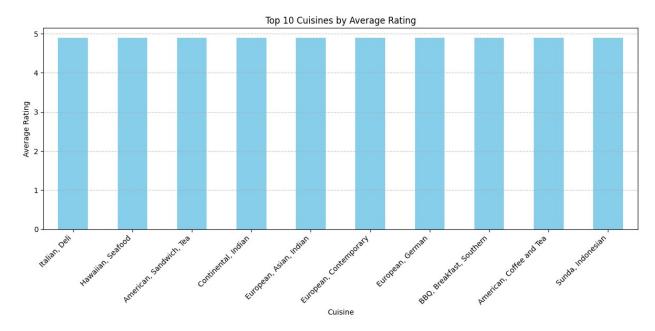
```
model.fit(X train, y train)
   y pred = model.predict(X test)
   mse = mean squared error(y test, y pred)
    rmse = mean squared error(y test, y pred, squared=False)
    r2 = r2_score(y_test, y_pred)
    results[name] = {'MSE': mse, 'RMSE': rmse, 'R^2': r2}
# Print results
print("Regression Model Performance:")
for name, metrics in results.items():
   print(f"Model: {name}")
   print(f"MSE: {metrics['MSE']:.2f}")
   print(f"RMSE: {metrics['RMSE']:.2f}")
   print(f"R^2: {metrics['R^2']:.2f}")
   print("-----")
Regression Model Performance:
Model: Linear Regression
MSE: 1.72
RMSE: 1.31
R^2: 0.25
Model: Decision Tree
MSE: 0.20
RMSE: 0.44
R^2: 0.91
Model: Random Forest
MSE: 0.14
RMSE: 0.37
R^2: 0.94
```

Task 2: Customer Preference Analysis

```
# Group the data by cuisine and calculate the average rating for each
cuisine
avg_rating_by_cuisine = df.groupby('Cuisines')['Aggregate
rating'].mean().sort_values(ascending=False)

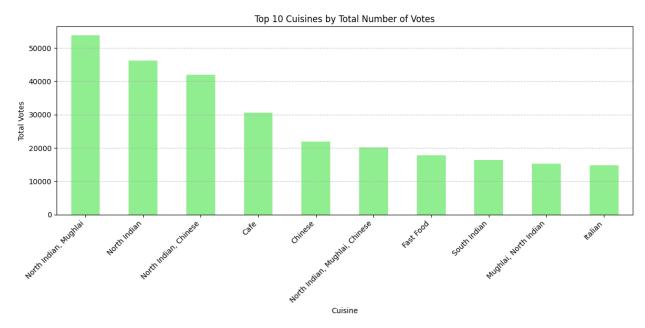
# Plot the top cuisines by average rating
plt.figure(figsize=(12, 6))
avg_rating_by_cuisine.head(10).plot(kind='bar', color='skyblue')
plt.title('Top 10 Cuisines by Average Rating')
plt.xlabel('Cuisine')
plt.ylabel('Average Rating')
plt.ylabel('Average Rating')
plt.xticks(rotation=45, ha='right')
plt.grid(axis='y', linestyle='--', alpha=0.7)
```

```
plt.tight_layout()
plt.show()
```

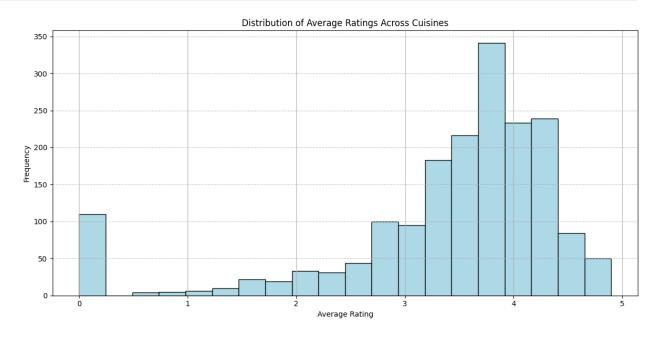


```
# Group the data by cuisine and calculate the total number of votes
for each cuisine
total_votes_by_cuisine = df.groupby('Cuisines')
['Votes'].sum().sort_values(ascending=False)

# Plot the top cuisines by total number of votes
plt.figure(figsize=(12, 6))
total_votes_by_cuisine.head(10).plot(kind='bar', color='lightgreen')
plt.title('Top 10 Cuisines by Total Number of Votes')
plt.xlabel('Cuisine')
plt.ylabel('Total Votes')
plt.ylabel('Total Votes')
plt.sticks(rotation=45, ha='right')
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()
plt.show()
```

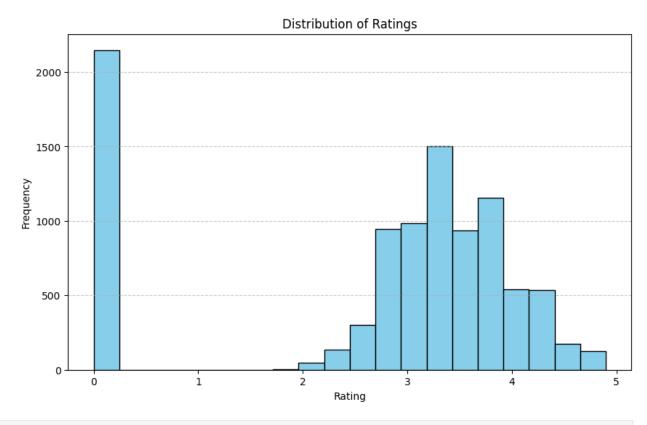


```
# Plot the distribution of average ratings across cuisines
plt.figure(figsize=(12, 6))
avg_rating_by_cuisine.hist(bins=20, color='lightblue',
edgecolor='black')
plt.title('Distribution of Average Ratings Across Cuisines')
plt.xlabel('Average Rating')
plt.ylabel('Frequency')
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()
plt.show()
```



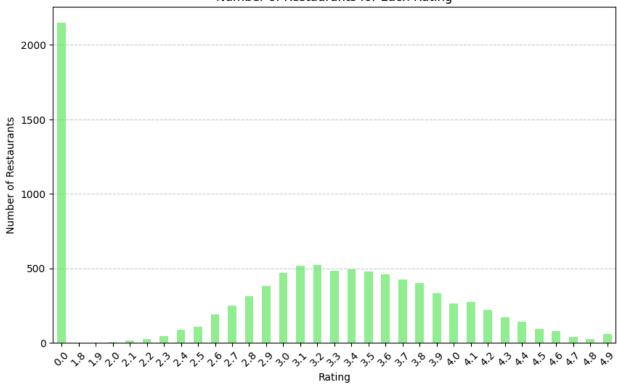
Task 3: Data Visualization

```
# Create a histogram to represent the distribution of ratings
plt.figure(figsize=(10, 6))
plt.hist(df['Aggregate rating'], bins=20, color='skyblue',
edgecolor='black')
plt.title('Distribution of Ratings')
plt.xlabel('Rating')
plt.ylabel('Frequency')
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.show()
```



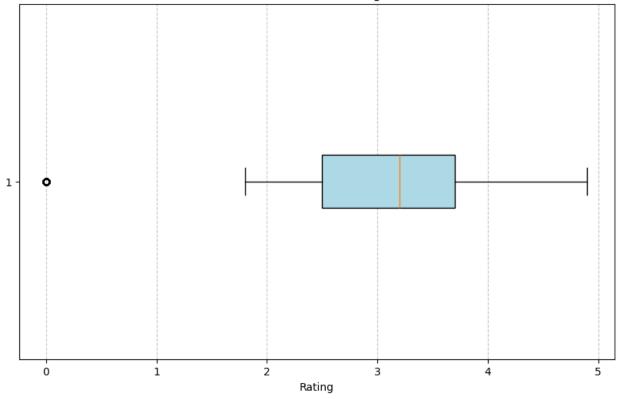
```
# Create a bar plot to represent the number of restaurants for each
rating
plt.figure(figsize=(10, 6))
df['Aggregate rating'].value_counts().sort_index().plot(kind='bar',
color='lightgreen')
plt.title('Number of Restaurants for Each Rating')
plt.xlabel('Rating')
plt.ylabel('Number of Restaurants')
plt.ylabel('Number of Restaurants')
plt.xticks(rotation=45)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.show()
```

Number of Restaurants for Each Rating

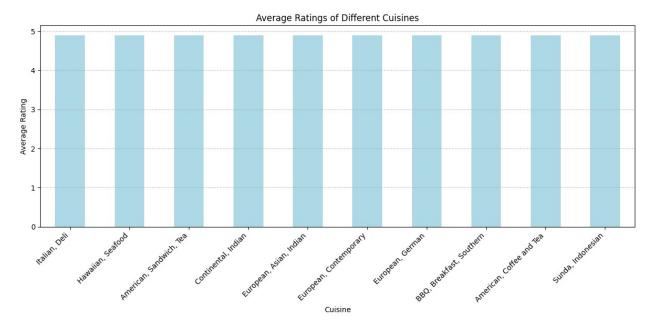


```
# Create a box plot to represent the distribution of ratings
plt.figure(figsize=(10, 6))
plt.boxplot(df['Aggregate rating'], vert=False, patch_artist=True,
boxprops=dict(facecolor='lightblue'))
plt.title('Distribution of Ratings')
plt.xlabel('Rating')
plt.grid(axis='x', linestyle='--', alpha=0.7)
plt.show()
```



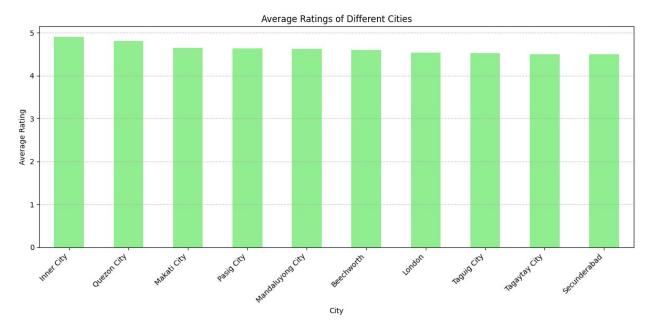


```
# Plot the average ratings of different cuisines
plt.figure(figsize=(12, 6))
avg_rating_by_cuisine.head(10).plot(kind='bar', color='lightblue')
plt.title('Average Ratings of Different Cuisines')
plt.xlabel('Cuisine')
plt.ylabel('Average Rating')
plt.ylabel('Average Rating')
plt.xticks(rotation=45, ha='right')
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()
plt.show()
```



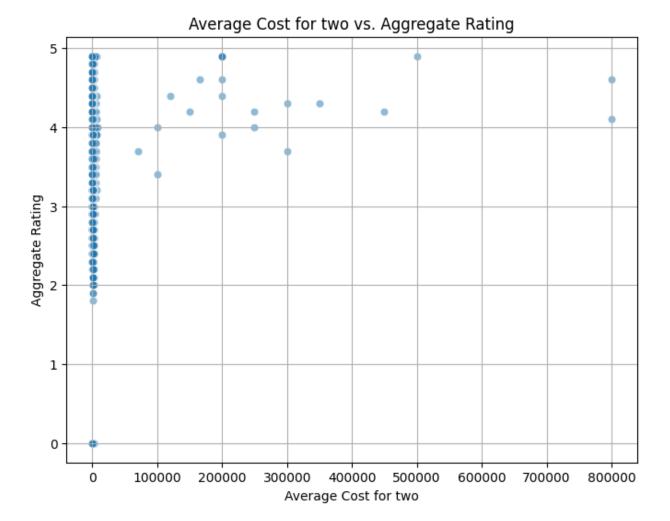
```
# Calculate the average rating for each city
avg_rating_by_city = df.groupby('City')['Aggregate
rating'].mean().sort_values(ascending=False)

# Plot the average ratings of different cities
plt.figure(figsize=(12, 6))
avg_rating_by_city.head(10).plot(kind='bar', color='lightgreen')
plt.title('Average Ratings of Different Cities')
plt.xlabel('City')
plt.ylabel('Average Rating')
plt.xticks(rotation=45, ha='right')
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.tight_layout()
plt.show()
```

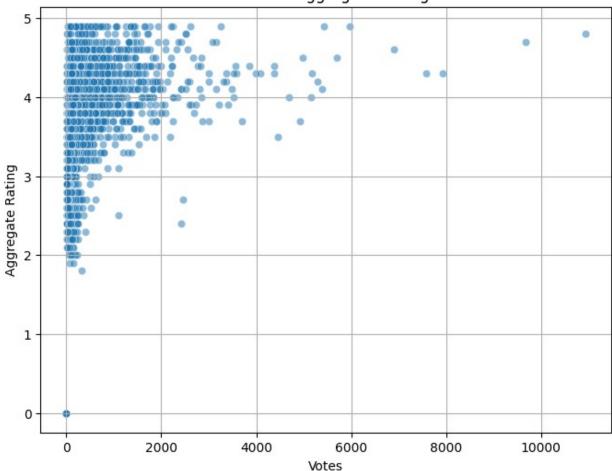


```
# Visualize numerical features vs. target variable (aggregate rating)
num_features = ['Average Cost for two', 'Votes'] # Add more numerical
features if needed

for feature in num_features:
    plt.figure(figsize=(8, 6))
    sns.scatterplot(data=df, x=feature, y='Aggregate rating',
alpha=0.5)
    plt.title(f'{feature} vs. Aggregate Rating')
    plt.xlabel(feature)
    plt.ylabel('Aggregate Rating')
    plt.grid(True)
    plt.show()
```







```
# Visualize categorical features vs. target variable (aggregate
rating)
cat_features = ['Price range', 'Has Table booking', 'Has Online
delivery'] # Add more categorical features if needed

for feature in cat_features:
    plt.figure(figsize=(8, 6))
    sns.boxplot(data=df, x=feature, y='Aggregate rating')
    plt.title(f'{feature} vs. Aggregate Rating')
    plt.xlabel(feature)
    plt.ylabel('Aggregate Rating')
    plt.grid(True)
    plt.show()
```

