```
import pandas as pd
df=pd.read csv('/content/Dataset .csv')
df.fillna(df.mean(numeric only=True), inplace=True)
df.fillna('unknown', inplace=True)
from sklearn.preprocessing import LabelEncoder
categorical cols = df.select dtypes(include=['object']).columns
label encoders = {}
for col in categorical cols:
    label_encoders[col] = LabelEncoder()
    df[col] = label_encoders[col].fit_transform(df[col])
from sklearn.preprocessing import StandardScaler
from sklearn.model selection import train test split
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error, r2_score
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
target_col = 'Aggregate rating'
X = df.drop(columns=[target_col])
y = df[target col]
numeric columns = X.select dtypes(include=['float64', 'int64']).columns
numeric transformer = StandardScaler()
preprocessor = ColumnTransformer(
    transformers=[
        ('num', numeric_transformer, numeric_columns),
    1)
pipeline = Pipeline(steps=[
    ('preprocessor', preprocessor),
    ('regressor', RandomForestRegressor(random_state=42))
])
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
pipeline.fit(X_train, y_train)
y_pred = pipeline.predict(X_test)
print("MSE:", mean_squared_error(y_test, y_pred))
print("R-squared:", r2_score(y_test, y_pred))
 → MSE: 0.027315152799581344
     R-squared: 0.9879991894430911
```

```
from sklearn.metrics.pairwise import cosine similarity
recommendation features = ['Cuisines', 'Price range', 'Aggregate rating']
feature matrix = df[recommendation features].values
similarity matrix = cosine similarity(feature matrix)
restaurant index = 0
similar indices = similarity matrix[restaurant index].argsort()[::-1][1:6]
print("Recommended restaurants:")
print(df.iloc[similar indices][['Restaurant Name', 'Cuisines', 'Aggregate rating']])
 Recommended restaurants:
           Restaurant Name Cuisines Aggregate rating
     361
                      2942
                                 924
                                                   4.8
     8821
                      6613
                                 626
                                                   3.3
     1962
                     1878
                                 625
                                                  3.3
     3784
                      1524
                                 604
                                                  3.1
                       788
                                 588
                                                  3.1
     4966
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=f"{row['Restaurant Name']}, Rating: {row['Aggregate rating']}"
   ).add_to(restaurant_map)
restaurant_map.save("restaurant_map.html")
grouped_data = df.groupby('City').agg({
    'Aggregate rating': 'mean',
    'Price range': 'mean',
    'Restaurant Name': 'count'
}).rename(columns={'Restaurant Name': 'Number of Restaurants'})
print(grouped_data)
→
           Aggregate rating Price range Number of Restaurants
     City
     0
                   4.300000
                               3.300000
                                                             20
    1
                  3.965000
                               2.650000
                                                             20
     2
                                                             21
                  4.161905
                               2.571429
     3
                  3.555000
                               1.700000
                                                             20
     4
                  3.395000
                               2.650000
                                                             20
     . . .
                                                            . . .
     136
                  3.900000
                               2.000000
                                                             1
     137
                  4.250000
                                                             20
                               3.250000
     138
                  3.200000
                               2.000000
                                                              1
```

 139
 3.300000
 2.000000
 1

 140
 4.292857
 2.857143
 14

[141 rows x 3 columns]