```
In [8]: |# Step 1: Aggregate transaction data
customer transactions = transactions.groupby('CustomerID').agg(
    TotalSpend=('TotalValue', 'sum'),
    TotalQuantity=('Quantity', 'sum'),
    TransactionCount=('TransactionID', 'count')
).reset_index()
# Step 2: Merge with customer profiles
customer profiles = customers.merge(customer transactions, on='CustomerID', h
# One-hot encode the Region column
customer profiles = pd.get dummies(customer profiles, columns=['Region'], dro
# Drop unnecessary columns
customer profiles features = customer profiles.drop(columns=['CustomerID', 'C
from sklearn.metrics.pairwise import cosine_similarity
import numpy as np
# Compute cosine similarity
similarity matrix = cosine similarity(customer profiles features)
# Create a DataFrame for similarity matrix
similarity df = pd.DataFrame(similarity matrix, index=customer profiles['Customer')
# Generate top 3 recommendations for the first 20 customers
lookalike dict = {}
for cust id in customer profiles['CustomerID'][:20]:
    # Sort similarity scores in descending order, exclude the customer themse
    similar customers = similarity df[cust id].sort values(ascending=False)[1
    lookalike_dict[cust_id] = list(zip(similar_customers.index, similar_customers.index)
# Convert the Lookalike dictionary into a DataFrame
lookalike df = pd.DataFrame.from dict(lookalike dict, orient='index', columns
lookalike_df.to_csv("Lookalike.csv")
print(lookalike df.head())
                               Top1
                                                            Top2 \
C0001 (C0011, 0.9999999840608153)
                                     (C0012, 0.9999999811673553)
C0002
       (C0034, 0.9999999147211551)
                                     (C0043, 0.9999998706917115)
       (C0136, 0.9999998983359859)
                                     (C0190, 0.9999998793526346)
C0003
       (C0195, 0.999999946388577)
                                     (C0039, 0.9999999944704254)
C0004
C0005
       (C0145, 0.9999999957807079)
                                     (C0173, 0.999999952585269)
                               Top3
C0001
       (C0191, 0.9999999794383162)
C0002
       (C0030, 0.9999998693464531)
C0003
       (C0142, 0.9999998567679897)
C0004 (C0169, 0.9999999781031476)
C0005
       (C0178, 0.9999999929629075)
```