

```

#Importing all Libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

Customers_data=pd.read_csv(r"C:\Users\A.Rohith Venkatesh\Downloads\
Customers.csv")
Products_data=pd.read_csv(r"C:\Users\A.Rohith Venkatesh\Downloads\
Products.csv")
Transactions_data=pd.read_csv(r"C:\Users\A.Rohith Venkatesh\Downloads\
Transactions.csv")

import os
print("Current working directory:", os.getcwd())

Current working directory: C:\Users\A.Rohith Venkatesh

"C:/Users/YourUsername/Documents/Customers_data.csv"

'C:/Users/YourUsername/Documents/Customers_data.csv'

import pandas as pd
from sklearn.metrics.pairwise import cosine_similarity
from sklearn.preprocessing import StandardScaler

# Load datasets
customers = pd.read_csv(r'C:\Users\A.Rohith Venkatesh\Downloads\
Customers.csv')
products = pd.read_csv(r'C:\Users\A.Rohith Venkatesh\Downloads\
Products.csv')
transactions = pd.read_csv(r'C:\Users\A.Rohith Venkatesh\Downloads\
Transactions.csv')
# Data Preprocessing
transactions_summary = transactions.groupby('CustomerID').agg({
    'TotalValue': 'sum',
    'Quantity': 'sum',
    'ProductID': lambda x: ' '.join(x.astype(str))
}).reset_index()

# Merge customer and transaction data
customer_profiles = pd.merge(customers, transactions_summary,
on='CustomerID', how='left')

# Feature encoding and scaling
customer_profiles.fillna(0, inplace=True)
scaler = StandardScaler()
encoded_features =
pd.get_dummies(customer_profiles.drop(columns=['CustomerID',
'CustomerName']))
scaled_features = scaler.fit_transform(encoded_features)

```

```

# Cosine similarity computation
similarity_matrix = cosine_similarity(scaled_features)

# Create Lookalike Recommendations
lookalike_dict = {}
for idx, customer_id in enumerate(customer_profiles['CustomerID']
[:20]):
    similar_indices = similarity_matrix[idx].argsort()[::-1][1:4] #
    Top 3 lookalikes
    similar_scores = similarity_matrix[idx][similar_indices]
    similar_customers =
customer_profiles['CustomerID'].iloc[similar_indices]
    lookalike_dict[customer_id] = list(zip(similar_customers,
similar_scores))

# Save results to Lookalike.csv
lookalike_df = pd.DataFrame({
    'CustomerID': lookalike_dict.keys(),
    'Lookalikes': [str(v) for v in lookalike_dict.values()]
})
lookalike_df.to_csv('Lookalike.csv', index=False)

print(customers.head())
print(products.head())
print(transactions.head())

```

	CustomerID	CustomerName	Region	SignupDate
0	C0001	Lawrence Carroll	South America	2022-07-10
1	C0002	Elizabeth Lutz	Asia	2022-02-13
2	C0003	Michael Rivera	South America	2024-03-07
3	C0004	Kathleen Rodriguez	South America	2022-10-09
4	C0005	Laura Weber	Asia	2022-08-15

  

	ProductID	ProductName	Category	Price
0	P001	ActiveWear Biography	Books	169.30
1	P002	ActiveWear Smartwatch	Electronics	346.30
2	P003	ComfortLiving Biography	Books	44.12
3	P004	BookWorld Rug	Home Decor	95.69
4	P005	TechPro T-Shirt	Clothing	429.31

  

	TransactionID	CustomerID	ProductID	TransactionDate	Quantity \
0	T00001	C0199	P067	2024-08-25 12:38:23	1
1	T00112	C0146	P067	2024-05-27 22:23:54	1
2	T00166	C0127	P067	2024-04-25 07:38:55	1
3	T00272	C0087	P067	2024-03-26 22:55:37	2
4	T00363	C0070	P067	2024-03-21 15:10:10	3

  

	TotalValue	Price
0	300.68	300.68
1	300.68	300.68
2	300.68	300.68

```
3      601.36  300.68
4      902.04  300.68
```

```
print(transactions_summary.head())
```

	CustomerID	TotalValue	Quantity						
ProductID									
0	C0001	3354.52	12		P054	P022	P096	P083	
P029									
1	C0002	1862.74	10			P095	P004	P019	
P071									
2	C0003	2725.38	14			P025	P006	P035	
P002									
3	C0004	5354.88	23	P049	P053	P038	P025	P097	P024
P077									
4	C0005	2034.24	7					P025	P039
P012									

```
print(encoded_features.head())
```

```
print(scaled_features[:5])
```

	TotalValue	Quantity	Region_Asia	Region_Europe	Region_North
America \					
0	3354.52	12.0	False	False	
False					
1	1862.74	10.0	True	False	
False					
2	2725.38	14.0	False	False	
False					
3	5354.88	23.0	False	False	
False					
4	2034.24	7.0	True	False	
False					

	Region_South America	SignupDate_2022-01-22	SignupDate_2022-02-02
\			
0	True	False	False
1	False	False	False
2	True	False	False
3	True	False	False
4	False	False	False

	SignupDate_2022-02-10	SignupDate_2022-02-13	...	\
0	False	False	...	
1	False	True	...	
2	False	False	...	

3	False	False	...
4	False	False	...

	ProductID_P092	P021	P048	\
0		False		
1		False		
2		False		
3		False		
4		False		

	ProductID_P092	P050	P027	P078	P021	P059	P063	P046	\
0								False	
1								False	
2								False	
3								False	
4								False	

	ProductID_P092	P076	P048	ProductID_P093	P033	P002	P091	\
0		False					False	
1		False					False	
2		False					False	
3		False					False	
4		False					False	

	ProductID_P095	P004	P019	P071	ProductID_P095	P040	P001	\
0				False			False	
1				True			False	
2				False			False	
3				False			False	
4				False			False	

	ProductID_P095	P062	P010	P084	P042	P033	P097	ProductID_P096	\
0							False	False	
1							False	False	
2							False	False	
3							False	False	
4							False	False	

	ProductID_P096	P075	P045	ProductID_P100	P081	P097
0		False				False
1		False				False
2		False				False
3		False				False
4		False				False

[5 rows x 385 columns]

```
[[-0.05188436 -0.1107351 -0.53881591 ... -0.07088812 -0.07088812
-0.07088812]
[-0.86271433 -0.43404927 1.85592145 ... -0.07088812 -0.07088812
-0.07088812]
```

```
[ -0.393842    0.21257907 -0.53881591 ... -0.07088812 -0.07088812
-0.07088812]
[  1.03537505  1.66749283 -0.53881591 ... -0.07088812 -0.07088812
-0.07088812]
[ -0.76949861 -0.91902053  1.85592145 ... -0.07088812 -0.07088812
-0.07088812]]
```