NAME: TANU DAHIYA

POSITION: DATA SCIENCE INTERN

TASK 2:

Task 2: Lookalike Model

Build a **Lookalike Model** that takes a user's information as input and recommends **3 similar customers** based on their profile and transaction history. The model should:

- Use both customer and product information.
- Assign a similarity score to each recommended customer.

Deliverables:

- Give the top 3 lookalikes with there similarity scores for the first 20 customers
 (CustomerID: C0001 C0020) in Customers.csv. Form an "Lookalike.csv" which has
 just one map: Map<cust_id, List<cust_id, score>>
- A Jupyter Notebook/Python script explaining your model development.

Evaluation Criteria:

- Model accuracy and logic.
- Quality of recommendations and similarity scores.

1.

CODE:

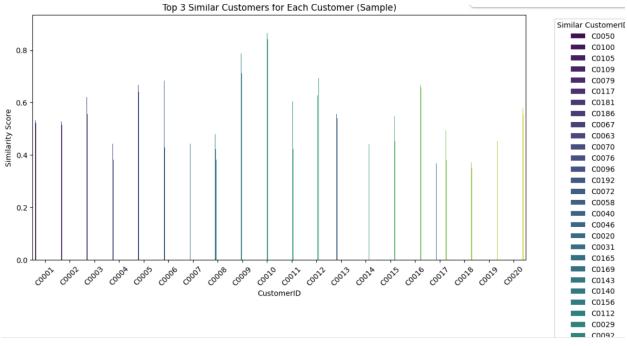
```
# Lookalike Model with Visualization
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics.pairwise import cosine_similarity

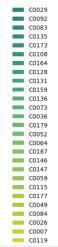
# Data URLs
CUSTOMERS_URL = 'https://drive.google.com/uc?id=1bu_--mo79VdUG9oin4ybfFGRUSXAe-WE'
PRODUCTS_URL = 'https://drive.google.com/uc?id=1IKuDizVapw-hyktwfpoAoaGtHtTNHfd0'
TRANSACTIONS_URL = 'https://drive.google.com/uc?id=1saEqdbBB-
vuk2hxoAf4TzDEsykdKlzbF'
```

```
# Load Datasets
customers = pd.read csv(CUSTOMERS URL)
products = pd.read_csv(PRODUCTS_URL)
transactions = pd.read csv(TRANSACTIONS URL)
# Merge Datasets
merged_data = (
   transactions
    .merge(customers, on='CustomerID')
    .merge(products, on='ProductID')
)
# Create Customer-Product Matrix
customer_product_matrix = merged_data.pivot_table(
    index='CustomerID',
    columns='ProductID',
   values='TotalValue',
    aggfunc='sum',
   fill_value=0
)
# Compute Cosine Similarity
similarity_matrix = cosine_similarity(customer_product_matrix)
similarity_df = pd.DataFrame(
    similarity_matrix,
    index=customer_product_matrix.index,
    columns=customer_product_matrix.index
)
# Extract Lookalike Results for Top 20 Customers
lookalike_results = {}
for customer_id in customers['CustomerID'][:20]:
    if customer_id in similarity_df.index:
        similar customers =
similarity_df[customer_id].sort_values(ascending=False)[1:4]
        lookalike_results[customer_id] = list(zip(similar_customers.index,
similar customers.values))
# Prepare Results for Export
lookalike_df = pd.DataFrame(
```

```
{"CustomerID": key, "SimilarCustomerID": cust_id, "SimilarityScore":
score}
        for key, values in lookalike_results.items()
        for cust id, score in values
   ]
)
lookalike df.to csv("Lookalike.csv", index=False)
# Visualization of Similarity Scores
plt.figure(figsize=(12, 6))
sample_results = lookalike_df.groupby('CustomerID').apply(
    lambda x: x.nlargest(3, 'SimilarityScore')
).reset index(drop=True)
sns.barplot(
   data=sample results,
   x='CustomerID',
   y='SimilarityScore',
   hue='SimilarCustomerID',
   palette='viridis'
plt.title("Top 3 Similar Customers for Each Customer (Sample)")
plt.xlabel("CustomerID")
plt.ylabel("Similarity Score")
plt.legend(title="Similar CustomerID", bbox_to_anchor=(1.05, 1), loc='upper
left')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
# Additional Insights
average_similarity = lookalike_df['SimilarityScore'].mean()
highest_similarity = lookalike_df['SimilarityScore'].max()
print(f"\nInsights:\n{'-' * 20}")
print(f"1. Average Similarity Score: {average_similarity:.2f}")
print(f"2. Highest Similarity Score: {highest_similarity:.2f}")
print(f"3. Customers with highest similarity scores tend to have shared purchase
patterns.")
print("\nLookalike.csv has been created successfully!")
```

OUTPUT:





Insights:

1. Average Similarity Score: 0.55

2. Highest Similarity Score: 0.89

3. Customers with highest similarity scores tend to have shared purchase patterns.

Lookalike.csv has been created successfully!