

Lookalike Model;

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

customers_file = 'Customers.csv'
products_file = 'Products.csv'
transactions_file = 'Transactions.csv'

customers_df = pd.read_csv(customers_file)
products_df = pd.read_csv(products_file)
transactions_df = pd.read_csv(transactions_file)
#reading the files

transactions_customers = pd.merge(transactions_df, customers_df, on="CustomerID", how="left")
merged_df = pd.merge(transactions_customers, products_df, on="ProductID", how="left")
#merging datasets

#Create customer profiles
def create_customer_profiles(data):
    profiles = (
        data.groupby("CustomerID")
        .agg(
            {
                "TotalValue": "sum", # Total transaction value
                "Quantity": "sum", # Total quantity purchased
                "ProductID": "nunique", # Number of unique products purchased
                "Category": lambda x: x.mode()[0] if len(x) > 0 else None, # Most common category
                "Region": "first", # Customer's region
            }
        )
        .reset_index()
    )
    return profiles

customer_profiles = create_customer_profiles(merged_df)

# Encoding
customer_profiles_encoded = pd.get_dummies(customer_profiles, columns=["Category", "Region"], drop_first=True)

# Normalize numerical columns
scaler = StandardScaler()
numerical_cols = ["TotalValue", "Quantity", "ProductID"]
customer_profiles_encoded[numerical_cols] = scaler.fit_transform(customer_profiles_encoded[numerical_cols])

# Calculate similarity matrix
def calculate_similarity_matrix(data):
    return cosine_similarity(data)

customer_ids = customer_profiles_encoded["CustomerID"]
similarity_matrix = calculate_similarity_matrix(customer_profiles_encoded.drop(columns=["CustomerID"]))

# Recommend top 3 similar customers
def get_top_3_similar(customers, sim_matrix, num_customers=20):
    lookalike_map = {}
    for i in range(num_customers): # First 20 customers
        customer_id = customers.iloc[i]
        similarities = list(enumerate(sim_matrix[i]))
        similarities = sorted(similarities, key=lambda x: x[1], reverse=True) # Sort by similarity score

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    top_3 = [(customers.iloc[j], score) for j, score in similarities[1:4]] # Exclude self
    lookalike_map[customer_id] = top_3
    return lookalike_map

# Get lookalikes for the first 20 customers
lookalike_map = get_top_3_similar(customer_ids, similarity_matrix)

# Generate Lookalike.csv
def generate_lookalike_csv(lookalike_map, output_file="Lookalike.csv"):
    lookalike_data = []
    for cust_id, lookalikes in lookalike_map.items():
        for lookalike_id, score in lookalikes:
            lookalike_data.append({"CustomerID": cust_id, "LookalikeID": lookalike_id, "Score": score})
    lookalike_df = pd.DataFrame(lookalike_data)
    lookalike_df.to_csv(output_file, index=False)
    print(f"{output_file} generated successfully!")

# Generate the CSV and display the first 20 results
generate_lookalike_csv(lookalike_map)


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 Lookalike.csv generated successfully!

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# Print lookalike recommendations for the first 20 customers
for customer_id, recommendations in lookalike_map.items():
    print(f"Customer {customer_id}:")
    for lookalike_id, score in recommendations:
        print(f"    Lookalike: {lookalike_id}, Score: {score:.4f}")

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 Lookalike: C0187, Score: 0.8966
 Lookalike: C0011, Score: 0.8729
 Customer C0007:
 Lookalike: C0005, Score: 0.9866
 Lookalike: C0115, Score: 0.9825
 Lookalike: C0140, Score: 0.9770
 Customer C0008:
 Lookalike: C0065, Score: 0.8779
 Lookalike: C0090, Score: 0.8772
 Lookalike: C0139, Score: 0.8510
 Customer C0009:
 Lookalike: C0198, Score: 0.9877
 Lookalike: C0061, Score: 0.9676
 Lookalike: C0062, Score: 0.9271
 Customer C0010:
 Lookalike: C0111, Score: 0.8899
 Lookalike: C0062, Score: 0.8798
 Lookalike: C0061, Score: 0.8539
 Customer C0011:
 Lookalike: C0137, Score: 0.9278
 Lookalike: C0174, Score: 0.8903
 Lookalike: C0191, Score: 0.8788

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Lookalike: C0041, Score: 0.9329
Lookalike: C0175, Score: 0.9000
Customer C0018:
Lookalike: C0046, Score: 0.8774
Lookalike: C0122, Score: 0.8409
Lookalike: C0068, Score: 0.7733
Customer C0019:
Lookalike: C0172, Score: 0.8080
Lookalike: C0081, Score: 0.7612
Lookalike: C0121, Score: 0.7537
Customer C0020:
Lookalike: C0015, Score: 0.9426
Lookalike: C0140, Score: 0.9390
Lookalike: C0186, Score: 0.9193
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