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import pandas as pd
import numpy as np
from \ sklearn.preprocessing \ import \ StandardScaler
from sklearn.metrics.pairwise import cosine_similarity
from sklearn.cluster import KMeans
from sklearn.metrics import davies_bouldin_score
import matplotlib.pyplot as plt
import seaborn as sns
from datetime import datetime
customers_df = pd.read_csv('Customers.csv')
products_df = pd.read_csv('Products.csv')
transactions_df = pd.read_csv('Transactions.csv')
def perform_clustering():
    # Merge customer and transaction data
    customer\_transactions = transactions\_df.groupby('CustomerID').agg(\{'TotalValue': 'sum', 'Quantity': 'sum'\}).reset\_index()
    \verb|customer_profiles| = \verb|customers_df.merge| (customer_transactions, on='CustomerID', how='left').fillna(0) |
    # Standardize numeric features
    scaler = StandardScaler()
    features = scaler.fit transform(customer profiles[['TotalValue', 'Quantity']])
    # Perform clustering
    kmeans = KMeans(n_clusters=4, random_state=42)
    customer_profiles['Cluster'] = kmeans.fit_predict(features)
    # Calculate DB Index
    db_index = davies_bouldin_score(features, customer_profiles['Cluster'])
    print(f"Davies-Bouldin Index: {db_index}")
    # Visualize clusters
    plt.figure(figsize=(10, 6))
    sns.scatterplot(
        x=features[:, 0], y=features[:, 1], hue=customer_profiles['Cluster'], palette='viridis'
    plt.title('Customer Clusters')
    plt.xlabel('TotalValue (scaled)')
    plt.ylabel('Quantity (scaled)')
    plt.legend(title='Cluster')
    plt.show()
# Execute task
perform_clustering()
```

→ Davies-Bouldin Index: 0.7102764046737506

