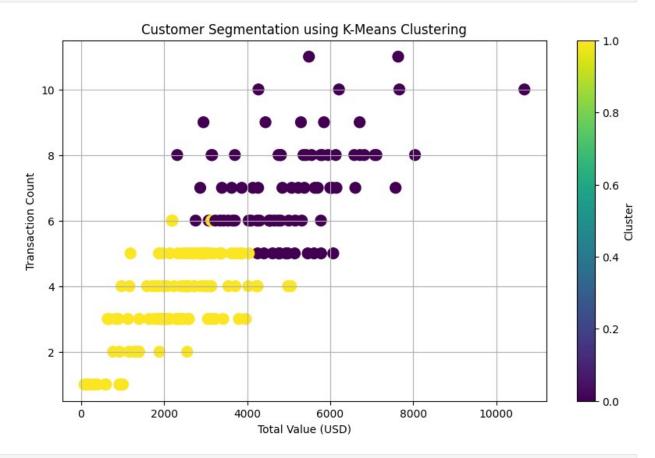
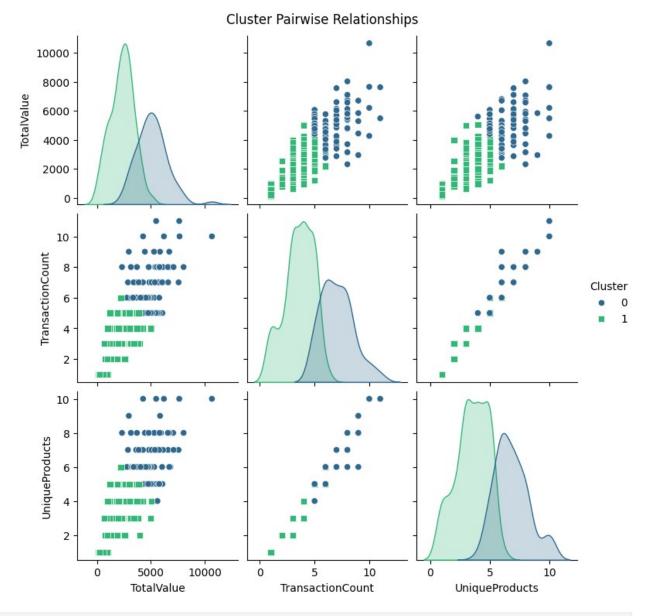
```
import pandas as pd
import numpy as np
import seaborn as sns
from sklearn.cluster import KMeans, AgglomerativeClustering
from sklearn.metrics import silhouette score, davies bouldin score
from sklearn.preprocessing import StandardScaler
import matplotlib.pyplot as plt
customers = pd.read csv('Customers.csv')
transactions = pd.read_csv('Transactions.csv')
merged_data = transactions.merge(customers, on='CustomerID')
customer profile =
merged_data.groupby('CustomerID').agg({'TotalValue':
'sum','TransactionID': 'count','ProductID': 'nunique', }
).reset index().rename(columns={'TransactionID': 'TransactionCount',
'ProductID': 'UniqueProducts'})
scaler = StandardScaler()
features scaled = scaler.fit transform(customer profile[['TotalValue',
'TransactionCount', 'UniqueProducts']])
best score = -1
best num clusters = 0
num clusters range = range(2, 11)
for n clusters in num clusters range:
    kmeans = KMeans(n clusters=n clusters, random state=42)
    cluster labels = kmeans.fit predict(features scaled)
    silhouette avg = silhouette score(features scaled, cluster labels)
    db index = davies bouldin score(features scaled, cluster labels)
    print(f'Clusters: {n clusters}, Silhouette Score:
{silhouette avg:.4f}, DB Index: {db index:.4f}')
    if silhouette avg > best score:
        best score = silhouette avg
        best num clusters = n clusters
        best kmeans model = kmeans
print(f'\nBest Silhouette Score: {best score: .4f} with Clusters:
{best num clusters}')
Clusters: 2, Silhouette Score: 0.4859, DB Index: 0.7327
Clusters: 3, Silhouette Score: 0.4323, DB Index: 0.7636
Clusters: 4, Silhouette Score: 0.3744, DB Index: 0.9182
Clusters: 5, Silhouette Score: 0.3847, DB Index: 0.9246
Clusters: 6, Silhouette Score: 0.3937, DB Index: 0.8152
Clusters: 7, Silhouette Score: 0.3879, DB Index: 0.8450
Clusters: 8, Silhouette Score: 0.3738, DB Index: 0.8547
Clusters: 9, Silhouette Score: 0.3978, DB Index: 0.8094
```

```
Clusters: 10, Silhouette Score: 0.3839, DB Index: 0.8659
Best Silhouette Score: 0.4859 with Clusters: 2
best kmeans model = KMeans(n clusters=best num clusters,
random state=42)
customer profile['Cluster'] =
best kmeans model.fit predict(features scaled)
plt.figure(figsize=(10, 6))
plt.scatter(customer profile['TotalValue'],
customer profile['TransactionCount'], c=customer profile['Cluster'],
cmap='viridis', s=100)
plt.title('Customer Segmentation using K-Means Clustering')
plt.xlabel('Total Value (USD)')
plt.ylabel('Transaction Count')
plt.colorbar(label='Cluster')
plt.grid()
plt.show()
```



```
sns.pairplot(customer_profile, vars=['TotalValue', 'TransactionCount',
'UniqueProducts'], hue='Cluster', palette='viridis', diag_kind='kde',
markers=['o', 's', 'D', 'X', '^'])
```

```
plt.suptitle('Cluster Pairwise Relationships', y=1.02)
plt.show()
/usr/local/lib/python3.11/dist-packages/seaborn/axisgrid.py:1615:
UserWarning: The markers list has more values (5) than needed (2),
which may not be intended.
  func(x=x, y=y, **kwargs)
/usr/local/lib/python3.11/dist-packages/seaborn/axisgrid.py:1615:
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which may not be intended.
  func(x=x, y=y, **kwargs)
```



```
from mpl_toolkits.mplot3d import Axes3D

fig = plt.figure(figsize=(12, 8))
ax = fig.add_subplot(111, projection='3d')

scatter = ax.scatter(
    customer_profile['TotalValue'],
    customer_profile['TransactionCount'],
    customer_profile['UniqueProducts'],
    c=customer_profile['Cluster'],
    cmap='viridis',
    s=100
)
```

```
ax.set_title('3D Visualization of Clusters')
ax.set_xlabel('TotalValue')
ax.set_ylabel('TransactionCount')
ax.set_zlabel('UniqueProducts')
fig.colorbar(scatter, label='Cluster')
plt.show()
```

