jzx6ujdee

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1 Task 3: Customer Clustering

Step 1: Importing Libraries

```
[1]: import pandas as pd
from sklearn.cluster import KMeans
from sklearn.metrics import davies_bouldin_score
from sklearn.preprocessing import MinMaxScaler
import seaborn as sns
import matplotlib.pyplot as plt
```

Step 2: Loading Datasets

```
[2]: customers = pd.read_csv(r"C:\Users\JOGESH\Downloads\Customers.csv")
    transactions = pd.read_csv(r"C:\Users\JOGESH\Downloads\Transactions.csv")
    data = pd.merge(transactions, customers, on='CustomerID', how='inner')
```

Step 3: Feature Engineering

Step 4: Performing Clustering

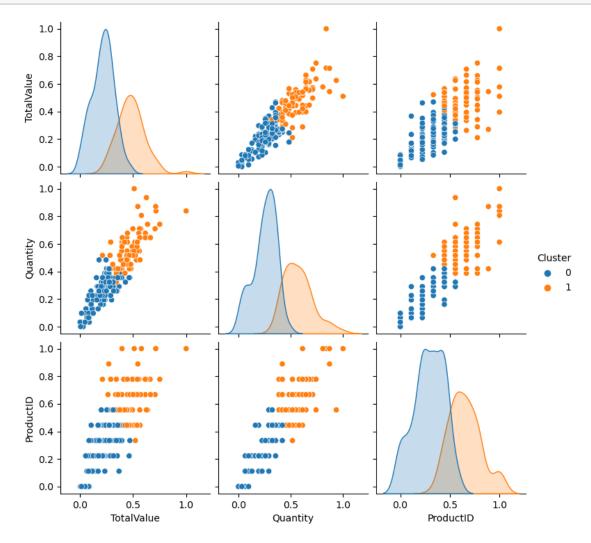
```
optimal_k = db_scores.index(min(db_scores)) + 2
print(f'Optimal number of clusters: {optimal_k}')
kmeans = KMeans(n_clusters=optimal_k, n_init='auto', random_state=42).

fit(customer_features.iloc[:, 1:])
customer_features['Cluster'] = kmeans.labels_
```

Optimal number of clusters: 2

Step 5: Visualizing Clusters

```
[6]: sns.pairplot(customer_features, hue='Cluster', diag_kind='kde') plt.show()
```



Step 6: Saving Clustering Results

```
[7]: customer_features[['CustomerID', 'Cluster']].

oto_csv('Keerthana_Kumbham_Clustering.csv', index=False)
```

Step 7: Reporting DB Index

```
[8]: print(f'Davies-Bouldin Index for {optimal_k} clusters: {min(db_scores)}')
```

Davies-Bouldin Index for 2 clusters: 0.3544759770719979