

Clustering Report: Customer Segmentation

Objective

To segment customers based on their transaction behavior and profile information using clustering techniques and evaluate cluster quality using metrics like Davies-Bouldin Index (DB Index) and Silhouette Score.

Clustering Results

1. Number of Clusters Formed:

- The optimal number of clusters was determined to be 4, based on the elbow method.

2. Davies-Bouldin Index (DB Index):

- DB Index: 1.37
- A lower DB Index indicates better clustering. The value of 1.37 suggests moderate separation and compactness of clusters.

3. Silhouette Score:

- Silhouette Score: 0.22
 - A Silhouette Score closer to 0 indicates overlapping clusters or less-defined boundaries between clusters. The score of 0.22 suggests weakly defined clusters, potentially due to overlapping features or insufficient separation in the data.
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Cluster Characteristics

1. Cluster 1 (High Spenders):

- Customers with high total transaction value and frequent purchases.
- Strategy: Focus on loyalty programs and exclusive offers.

2. Cluster 2 (Moderate Spenders):

- Customers with average transaction value and diverse product interests.
- Strategy: Design cross-sell and upsell campaigns.

3. Cluster 3 (Occasional Buyers):

- Customers with low purchase frequency and minimal spending.
- Strategy: Use promotional offers to encourage higher engagement.

4. Cluster 4 (New Customers):

- Recently onboarded customers with little transaction data.
 - Strategy: Engage with onboarding campaigns and incentives.
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Evaluation Metrics

1. Davies-Bouldin Index (1.37):

- Indicates moderately well-separated clusters with some overlap.

2. Silhouette Score (0.22):

- Suggests weakly defined clusters, potentially requiring feature engineering or different clustering techniques.

3. Cluster Distribution:

- Review the number of customers in each cluster to ensure a balanced segmentation. Unbalanced clusters might require adjustments in feature scaling or algorithm parameters