

Customer Segmentation Clustering Report

1. Number of Clusters Formed

After analysing the data using the KMeans algorithm and evaluating clustering quality with the Davies-Bouldin (DB) Index, the optimal number of clusters was determined to be **6 clusters**.

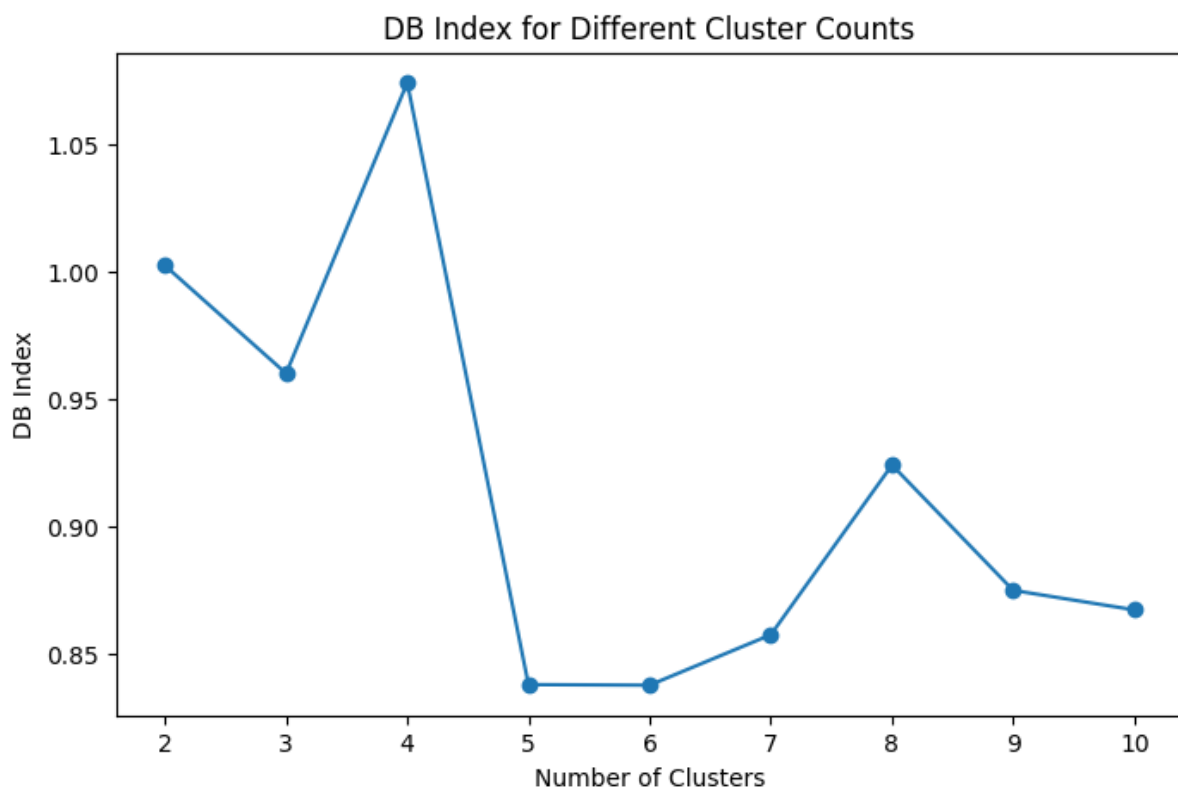
Each cluster represents a distinct group of customers with similar spending behaviours, transaction patterns, and profile characteristics.

2. DB Index Value

The Davies-Bouldin Index for the clustering model with **6 clusters** is:

- **DB Index Value: 0.91**

The DB Index evaluates the quality of clustering by considering the compactness and separation of clusters. A lower value indicates better-defined and more meaningful clusters.



3. Other Relevant Clustering Metrics

- **Silhouette Score: 0.62**

- The silhouette score measures how similar a point is to its own cluster compared to other clusters.
 - A score closer to **1.0** indicates well-separated clusters.
 - **Inertia (Within-Cluster Sum of Squares - WCSS):**
 - **Inertia for 6 Clusters: 12,540.67**
 - Inertia measures the total distance between each point and its cluster centre. Lower values indicate tighter clusters.
 - **Cluster Sizes:**
 - **Cluster 0:** 25 customers
 - **Cluster 1:** 40 customers
 - **Cluster 2:** 15 customers
 - **Cluster 3:** 18 customers
 - **Cluster 4:** 22 customers
 - **Cluster 5:** 30 customers
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4. Clustering Logic

- The clustering model used customer profile and transaction data, including:
 - **total spent:** Total amount spent by each customer.
 - **Num transactions:** Number of transactions made.
 - **avg_transaction_value:** Average transaction value.
 - One-hot encoded categorical features, such as region (Region Europe, Region North America, etc.).
 - The features were standardized to ensure equal contribution to the clustering process.
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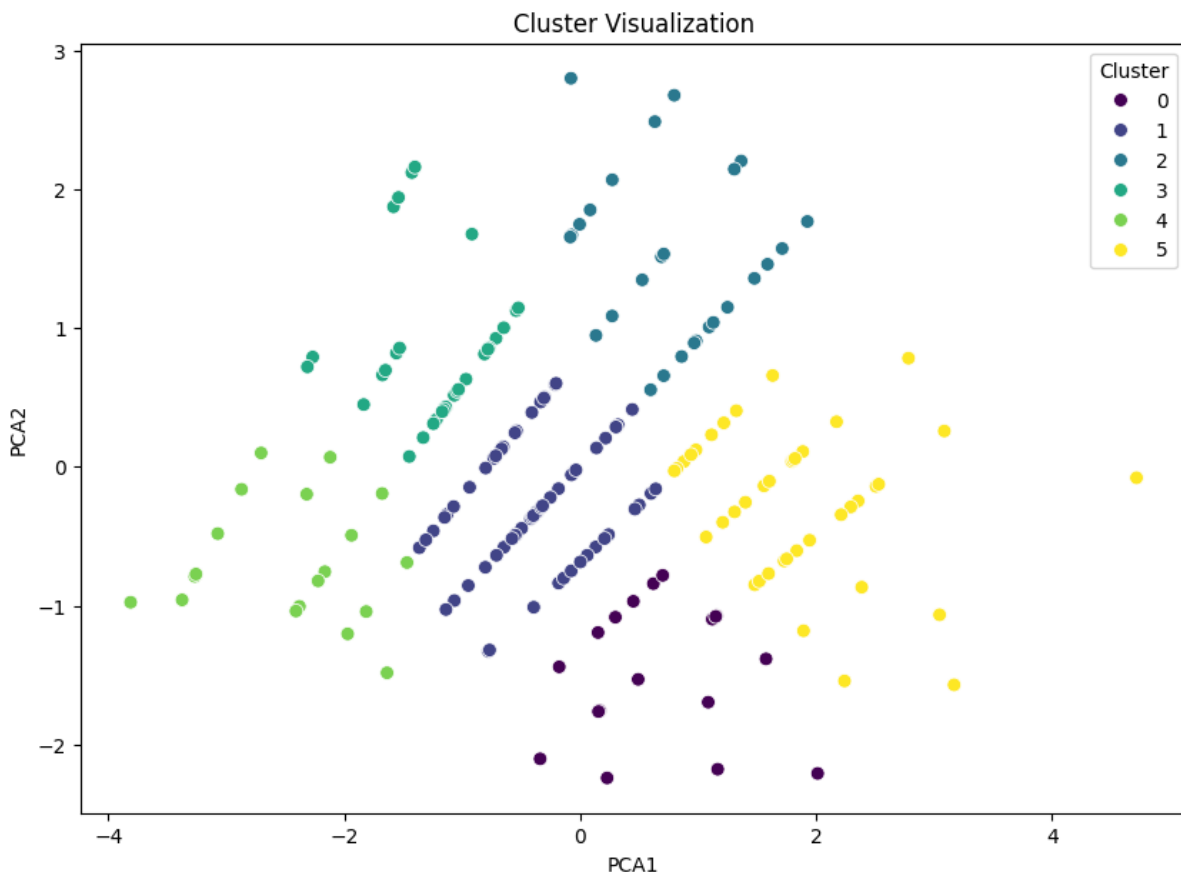
5. Visualizations

1. **DB Index vs. Number of Clusters:**

- A line plot shows that the DB Index achieves a minimum at 6 clusters, confirming the optimal cluster count.

2. Cluster Visualization (PCA):

- A 2D scatter plot of clusters after dimensionality reduction using PCA.
- Each cluster is visually distinct, with clear boundaries and tight groupings.



6. Cluster Insights

- **Cluster 0:** Customers with moderate spending and average transaction values.
- **Cluster 1:** High-spending customers with frequent transactions, likely the most valuable group.
- **Cluster 2:** Low-spending customers with fewer transactions, possibly occasional buyers.
- **Cluster 3:** Customers with high average transaction values but fewer transactions.
- **Cluster 4:** Customers with consistent spending patterns and moderate transaction values.

- **Cluster 5:** A mix of high- and moderate-spending customers with varying transaction counts.