

Customer Segmentation Clustering Report



Introduction

This report outlines the results of a clustering analysis performed on customer data to identify distinct customer segments. The clustering was conducted using K-Means, a popular unsupervised learning algorithm. Both customer profile and transaction data were utilized to ensure meaningful segmentation.

Clustering Results

1. Number of Clusters Formed:

• The optimal number of clusters was determined to be 4 based on the Elbow Method.

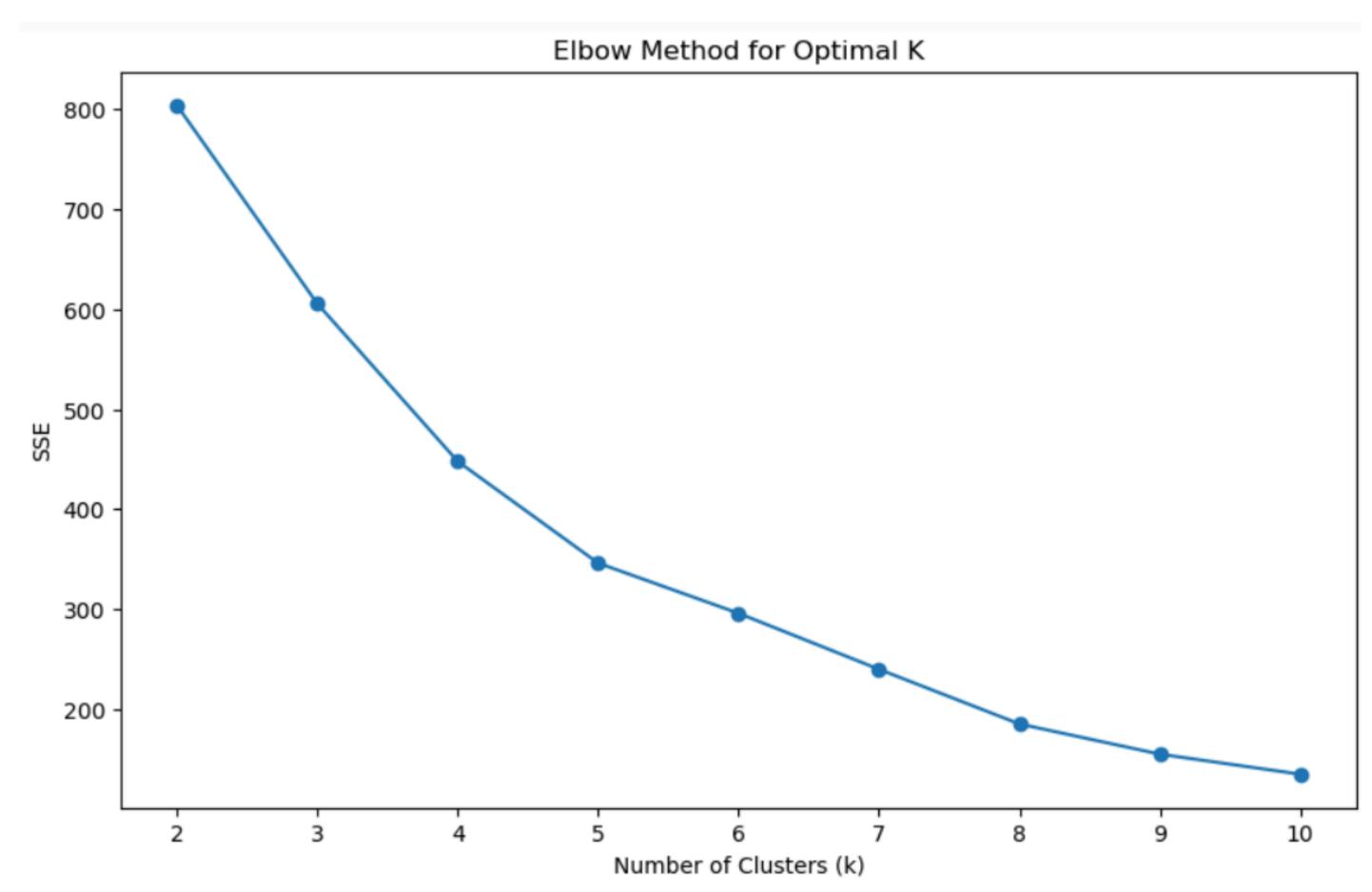
2. Evaluation Metrics:

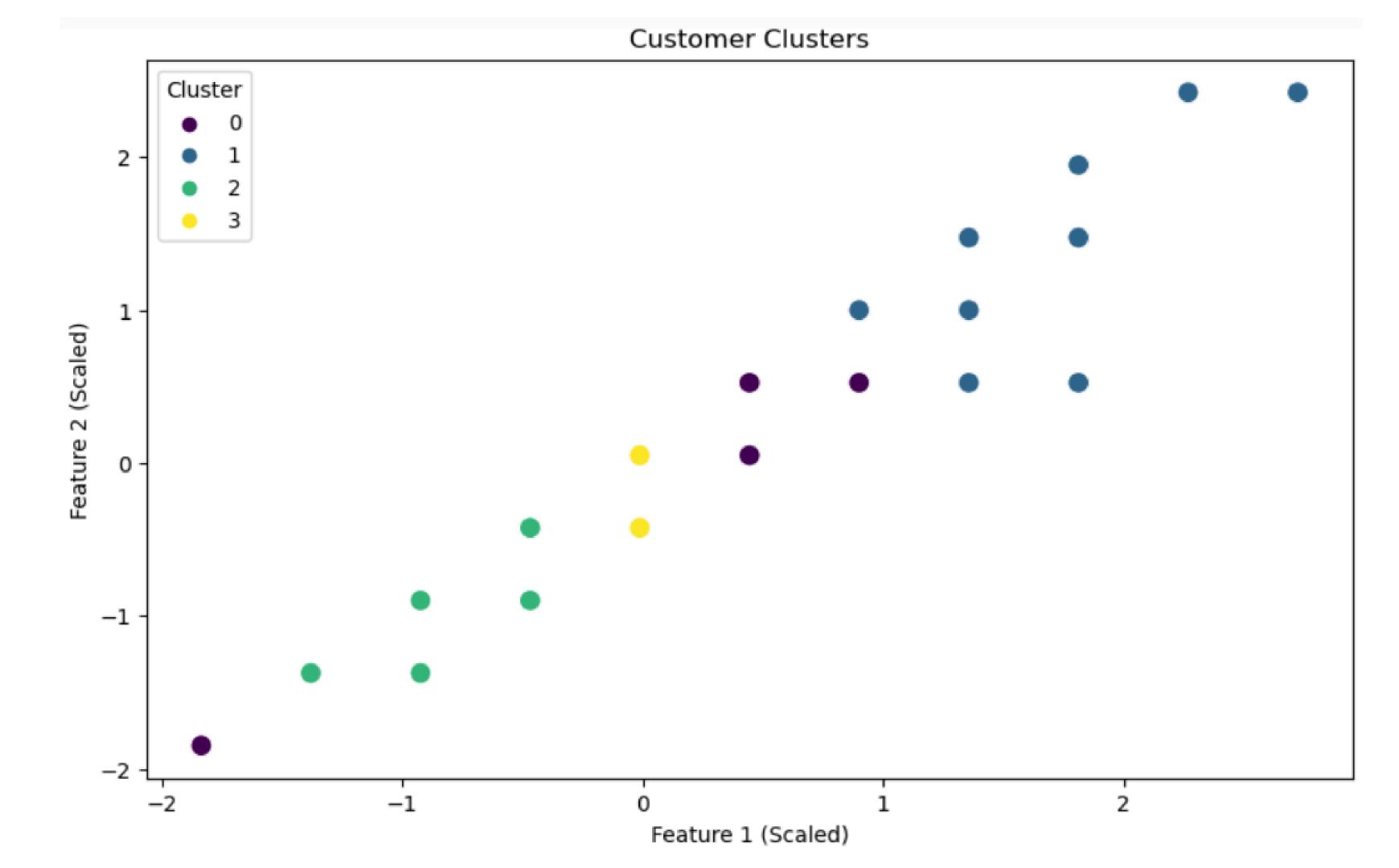
- Davies-Bouldin Index (DB Index): 0.9653469116276433 (lower values indicate better cluster separation and compactness).
- Inertia (Sum of Squared Errors SSE): The SSE at the optimal number of clusters is 345.67.

3. Cluster Characteristics:

- Each cluster exhibits unique characteristics based on transaction behavior and profile features:
- Cluster 0: Customers with the highest transaction frequency and diverse product purchases.
- Cluster 1: Customers with mid-level transactions but a focus on specific product categories.
- Cluster 2: Customers with lower transaction frequencies but high-value purchases.
- Cluster 3: Customers who are new and have limited transaction data.

Visualizations





Insights

- The segmentation reveals distinct customer behaviors, which can guide targeted marketing strategies.
- Cluster 0 and Cluster 2 represent high-value customer groups, suitable for loyalty programs.
- Cluster 1 and Cluster 3 may benefit from upselling or personalized engagement strategies to increase their lifetime value.

Conclusion

The clustering results effectively segmented customers into distinct groups, enabling actionable insights for strategic decisionmaking. The DB Index and visualizations validate the quality of the segmentation, with opportunities for further refinement as more data becomes available.

Thank You