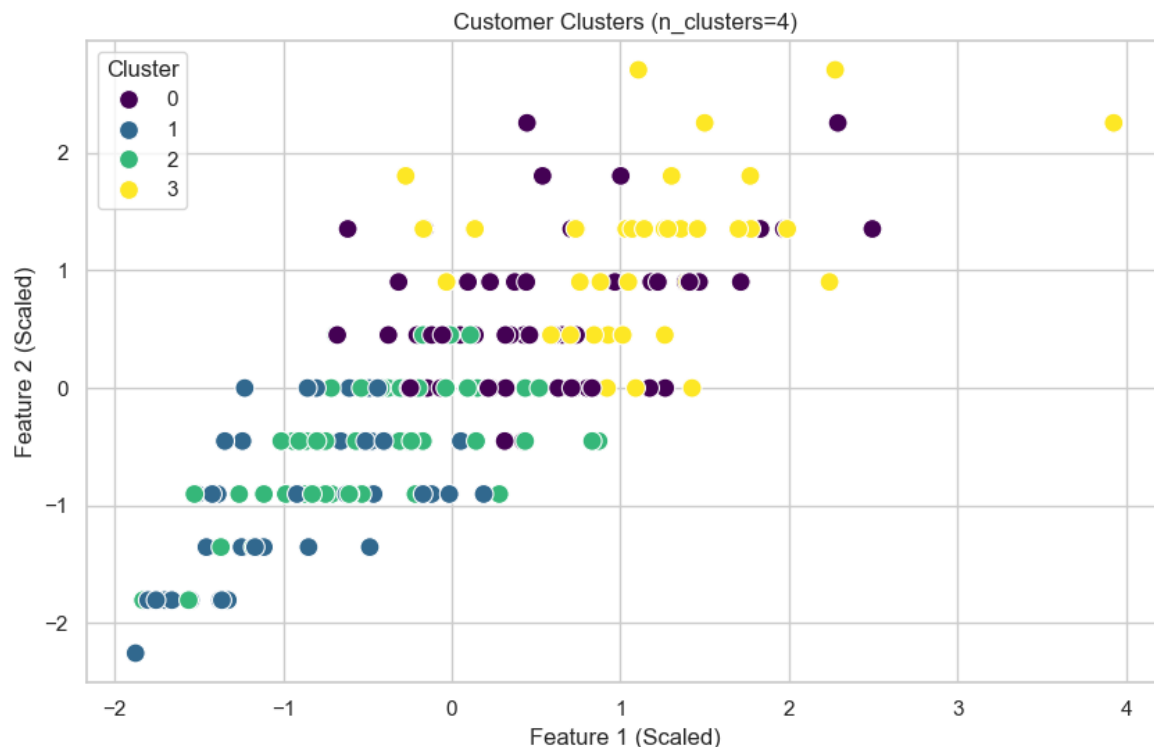


# Clustering Results Report



## 1. Number of Clusters Formed:

- After applying the KMeans clustering algorithm, we identified 4 distinct customer groups.
- These groups were formed based on customer spending behavior and transaction patterns.
- Each group represents a set of customers with similar characteristics in how they interact with the products and services.
- The segmentation allows for targeted strategies to enhance engagement with different customer groups.

## 2. Davies-Bouldin Index:

- The Davies-Bouldin Index (DB Index) for the clustering model is 0.9616.
- This index measures how well-separated the clusters are.
- A lower DB Index indicates better separation between clusters.
- The current value suggests that while the clusters are reasonably distinct, there is still room for improvement in achieving better separation.

## 3. Silhouette Score:

- The Silhouette Score is 0.3444.
- This score indicates how similar each customer is to its own cluster compared to other clusters.
- A higher score signifies that the clusters are well-defined and well-separated.

- A score of 0.3444 suggests that the clusters are somewhat separated, but not highly distinct.
- This indicates that the model has potential, but could benefit from further optimization or fine-tuning.

#### 4. Other relevant clustering metrics

- **Inertia (Within-Cluster Sum of Squares):** The inertia measures the compactness of the clusters, indicating how close the data points are to their cluster centres. A lower inertia value indicates tighter, more compact clusters, which is a desirable characteristic.
- **Cluster Size Distribution:** It's important to check if the clusters are of reasonable sizes. Ideally, clusters should not be too small (underrepresented) or too large, ensuring balanced segmentation.