# **Customer Segmentation Clustering**

#### Introduction:

This report summarizes the results of the customer segmentation analysis using clustering. The aim was to group customers based on both profile and transaction data to uncover patterns and behaviors.

### **Key Findings:**

- Optimal Number of Clusters: 2 clusters, based on the Silhouette Score.
- Davies-Bouldin Index: 0.72 (lower values indicate better cluster separation).
- **Silhouette Score**: **0.49** (higher values indicate better cluster cohesion and separation).

## **Methodology:**

- 1. **Clustering Algorithm**: **KMeans** clustering was applied with different numbers of clusters (from 2 to 10) to evaluate optimal segmentation.
- 2. Metrics:
  - Davies-Bouldin Index: Measures cluster separation. The lowest DB Index was 0.72 for 2 clusters.
  - Silhouette Score: Evaluates how well-separated the clusters are. The highest Silhouette Score of 0.49 occurred for 2 clusters.
- 3. **Elbow Method**: Supported the choice of **2 clusters** as optimal, as it showed minimal improvements beyond this point.

## **Conclusion:**

- The optimal segmentation resulted in 2 clusters, with a Silhouette Score of 0.49 and a Davies-Bouldin Index of 0.72, indicating well-separated and cohesive clusters.
- These results can be used for targeted marketing, customer behavior analysis, and product recommendations.