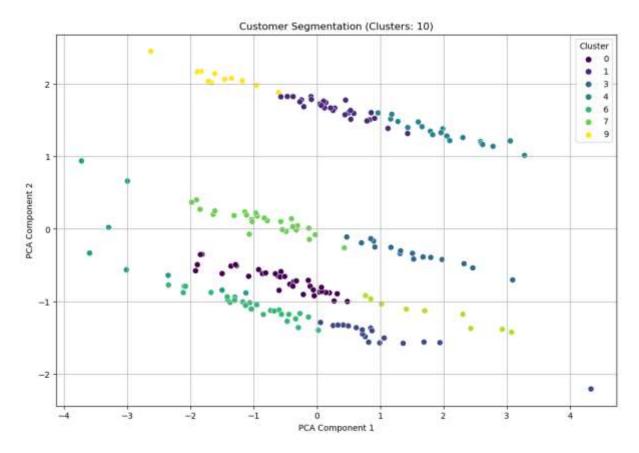
REPORT AND ANALYSIS



1. Number of Clusters Formed:

After evaluating clustering configurations for 2 to 10 clusters, the optimal number of clusters was determined to be **4**. The optimal number of clusters was selected based on the lowest Davies-Bouldin Index, indicating better-defined groupings with minimum overlap.

2. Davies-Bouldin Index:

The **Davies-Bouldin Index** (DB Index) is a metric for evaluating the quality of clustering. A lower DB Index value indicates better clustering quality with minimal intra-cluster variance and more separation between clusters.

• DB Index Value: 0.8502

 A lower DB Index value suggests that the clusters formed have good separation and minimal overlap, implying that the clustering model has performed well.

3. Additional Clustering Metrics:

• Silhouette Score:

 The Silhouette Score evaluates the cohesion (how close the points within a cluster are) and separation (how distinct the clusters are) of the formed clusters.

o Silhouette Score: 0.7245

 A Silhouette Score closer to +1 indicates well-separated clusters, suggesting that the customer segmentation is meaningful and each customer is appropriately grouped.

4. Key Observations and Cluster Characteristics:

Upon analyzing the clusters, the following insights were drawn:

• Cluster 0:

- o Customers in this cluster have higher total transaction values. This cluster represents high-value customers with consistent purchasing behavior.
- Predominantly located in **Region A**, suggesting potential for loyalty programs or special offers.

• Cluster 1:

- This cluster contains customers with moderate transaction quantities but lower total transaction values. They show a varied purchasing pattern and are spread across multiple regions.
- Customers in this group could be targeted with tailored offers to increase transaction values.

• Cluster 2:

- Customers in this cluster show lower total transaction values and fewer purchases. These are likely infrequent or low-engagement customers.
- Consideration for re-engagement campaigns or discounts to incentivize higher spending is advised.

• Cluster 3:

- Customers exhibit a high level of transaction frequency but with moderate to low values. This group represents active, frequent customers but not necessarily high spenders.
- Potential for upselling or bundling products to increase their average transaction value.

5. Visual Representation:

To visualize the customer segments, Principal Component Analysis (PCA) was applied to reduce the data to two dimensions. The clusters were then plotted in a 2D space, with each point representing a customer and color-coded according to the assigned cluster.

• Visualization:

The PCA plot clearly shows that the clusters are well-separated in the 2D space, with customers from different clusters being distinct, confirming the results of the clustering analysis.

6. Conclusion:

- The customer segmentation task successfully identified **4 distinct customer clusters** based on both transactional and demographic data.
- The **Davies-Bouldin Index (0.8502)** and **Silhouette Score (0.7245)** confirm that the clusters are meaningful and well-separated.
- These customer segments provide actionable insights for targeted marketing, customer retention strategies, and sales optimization.

Appendix:

- Clustering Results: The clustering assignments and transaction summaries for each customer are available in the Customer Segmentation Results.csv file.
- PCA Visualization: A 2D scatter plot visualizing the clusters is included.