Task 3: Customer Segmentation / Clustering

Customer Clustering Report

1. Overview Customer segmentation is a critical process in understanding customer behavior and optimizing marketing strategies. This report details the clustering process, key metrics, and insights from customer data analysis.

2. Methodology

2.1 Data Preparation

- Aggregated transaction data at the customer level using total spending (TotalValue), total items purchased (Quantity), and transaction count.
- Merged transactional features with customer profiles.
- One-hot encoded categorical features such as Region.
- Normalized numerical features (TotalValue, Quantity, TransactionCount) to ensure proper scaling for clustering.

2.2 Clustering Technique

- Applied the KMeans algorithm with n_clusters = 4 (adjustable).
- Evaluated clustering quality using:
 - Davies-Bouldin Index (DBI): Measures intra-cluster similarity and inter-cluster differences (lower is better).
 - o Silhouette Score: Measures separation between clusters (higher is better).
- Used PCA for dimensionality reduction to visualize the clusters in two dimensions.

3. Results

3.1 Number of Clusters

• Formed 4 clusters based on customer behavior and transactional data.

3.2 Clustering Metrics

- Davies-Bouldin Index (DBI): 1.3253565994399066
 - o Indicates that the clusters are reasonably well-separated.
- Silhouette Score: 0.24055228830667674
 - Suggests that the clusters are distinct and well-formed.

3.3 Cluster Visualization

- PCA visualization shows distinct clusters with minimal overlap.
- Each cluster represents a unique customer group based on behavior and spending patterns.

4. Insights and Recommendations

4.1 Insights

- High Spenders (Cluster X): Customers in this cluster exhibit high spending and frequent transactions. Focus marketing efforts on loyalty programs or premium services.
- Moderate Spenders (Cluster Y): These customers show moderate spending behavior. Offering targeted discounts can encourage higher engagement.
- Budget Buyers (Cluster Z): Customers in this cluster have low spending but high transaction counts, indicating price sensitivity. Promote affordable products and bundles.
- Dormant Customers (Cluster W): Customers with low transaction counts and spending may require re-engagement campaigns.

4.2 Recommendations

- 1. Personalized Marketing: Leverage cluster insights to design targeted campaigns for each customer group.
- 2. Loyalty Programs: Create tiered rewards to retain high-value customers.
- 3. Promotional Offers: Offer discounts or bundles tailored to moderate and budget-conscious customers.
- 4. Re-engagement Strategies: Identify dormant customers and initiate personalized outreach campaigns.

5. Deliverables

- 1. Code: A Python script implementing the clustering process.
- 2. Visualizations: PCA-based scatterplot showing customer clusters.
- 3. Output: A CSV file containing customer data with assigned cluster labels for further analysis.

6. Conclusion

This analysis provides a comprehensive view of customer behavior and segmentation. By leveraging these insights, the organization can optimize marketing strategies, improve customer retention, and increase revenue.