Customer Segmentation Analysis Report

Objective

This analysis is based on customer transaction data. Therefore, the purpose of the report is to segment customers on-upon identification of groups that are significant from a business standpoint. The aim is to educate us with insight towards high-value customer groups and develop strategies for marketing efforts, promotional exercises, or customer retention.

Data Overview

Focusing on the two datasets that were used to conduct this analysis, they are listed below:

- 1. Customers Dataset (Contains customer's details like CustomerID and SignupDate)
- 2. Transactions Dataset (Contains transaction's details like CustomerID, TransactionID, TransactionDate, TotalValue (amount spend), and Quantity)

Approach

The process involving customer segmentation involves:

1. Data Preprocessing

- Date Handling: We had converted SignupDate and TransactionDate into datetime formats for purposes of calculation of the relevant metrics such as the days since signup and active purchase duration.
- Missing Data: Working with customers without any transactions, missing values were coded to zero for the metric derived itself.

2. Feature Engineering

- We calculated different metrics out of the raw data available, which described customer behavior:
- Days Active: Number of days since a customer signed up.
- Transaction Count: Total transactions by each customer.
- Total Spend: Total money spent by the customer overall.
- Average Transaction Value: The average value spent on each transaction (Total Spend / Transaction Count).
- Purchase Frequency: Transactions made monthly.
- Purchase Duration: Time between the first and last transaction.
- This category of metrics actually offers an insight into recency, frequency, and monetary values.

These metrics provided a comprehensive view of customer behavior, including recency (Days Active), frequency (Transaction Count and Purchase Frequency), and monetary value (Total Spend and Avg Transaction Value).

3. Data Scaling

The calculated features had varying scales (e.g., Days Active was in days, Total Spend in dollars). To ensure fair comparison during clustering, we used StandardScaler to normalize the data.

4. Clustering

To group customers into segments:

- 1. **Model Selection**: We chose the K-Means algorithm due to its simplicity and effectiveness for numeric data.
- 2. **Optimal Cluster Selection**: We evaluated clusters ranging from 2 to 10 using the following metrics:
 - Davies-Bouldin Index (DBI): Measures the compactness and separation of clusters. Lower values indicate better-defined clusters.
 - Silhouette Score: Measures how well samples are clustered. Higher values indicate better-defined clusters.
 - o Based on the results, the optimal number of clusters was determined to be 8.

5. Visualization

Two visualizations were created to interpret the clustering results:

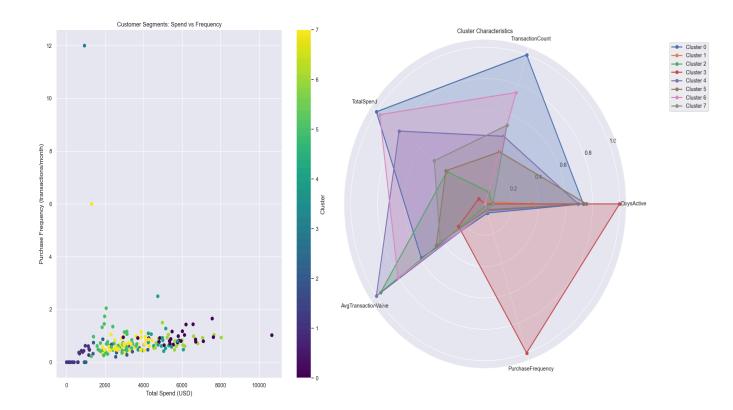
- 1. **Scatter Plot**: A 2D scatter plot of Total Spend vs. Purchase Frequency, colored by cluster labels, to observe customer distribution and segmentation.
- 2. **Radar Chart**: Showed normalized cluster centers for all features, highlighting differences in behavior between clusters.

Results

Clustering Evaluation

The evaluation metrics for the clustering model were as follows:

- Optimal Number of Clusters: 8
- **Davies-Bouldin Index**: 0.9116 (lower is better, indicating compact clusters with good separation).
- Silhouette Score: 0.2924 (indicating moderate cluster cohesion and separation).



Cluster Profiles

Below are the key characteristics of each customer cluster:

1. Cluster 0:

- o Size: 26 customers
- o **Behavior**: High spenders with moderate transaction counts and frequency.
- o Average Spend: \$5,861.11 per customer
- Average Transactions: 8.38
- o **Frequency**: 0.95 transactions/month
- Average Transaction Value: \$706.65
- Insights: These customers are valuable, making fewer but high-value purchases.

2. Cluster 1:

- o Size: 19 customers
- o **Behavior**: Low spenders with minimal transactions and low frequency.
- o Average Spend: \$627.26
- Average Transactions: 2.11
- **Frequency**: 0.23 transactions/month
- o Average Transaction Value: \$295.05
- Insights: These customers are infrequent buyers with low overall engagement.

3. Cluster 2:

- o Size: 19 customers
- Behavior: Moderate spenders with low transaction counts but high average transaction value.
- Average Spend: \$2,462.81

Average Transactions: 2.53

Frequency: 0.41 transactions/monthAverage Transaction Value: \$968.23

o **Insights**: High-value buyers with limited but significant transactions.

4. Cluster 3:

o Size: 1 customer

• **Behavior**: Unique customer with very high transaction frequency (12/month).

Average Spend: \$931.83Average Transactions: 2.0

Frequency: 12.00 transactions/month
Average Transaction Value: \$465.92

 Insights: Possibly an anomaly or a very specific customer with extremely frequent low-value purchases.

5. **Cluster 4**:

o Size: 19 customers

Behavior: High-value, moderate-frequency customers.

Average Spend: \$4,767.65Average Transactions: 4.89

Frequency: 0.80 transactions/monthAverage Transaction Value: \$996.64

o **Insights**: A valuable group with steady purchasing habits.

6. **Cluster 5**:

o Size: 48 customers

o **Behavior**: Average spenders with consistent but moderate purchasing.

Average Spend: \$2,512.05Average Transactions: 4.23

Frequency: 0.69 transactions/monthAverage Transaction Value: \$610.20

o **Insights**: Stable customers with potential for upselling.

7. Cluster 6:

o Size: 22 customers

Behavior: High spenders with moderate transaction counts and frequency.

Average Spend: \$5,696.99Average Transactions: 6.77

Frequency: 0.79 transactions/monthAverage Transaction Value: \$856.64

o **Insights**: Another high-value group with consistent engagement.

8. **Cluster 7**:

o Size: 46 customers

Behavior: Moderate spenders with steady purchasing habits.

Average Spend: \$3,075.35Average Transactions: 5.37

Frequency: 0.77 transactions/monthAverage Transaction Value: \$585.12

Insights: A middle-tier group with balanced behavior.

Conclusion

The segmentation identified 8 distinct customer groups with varying spending, frequency, and transaction behaviors. Key takeaways include:

- **Cluster 0, 4, and 6**: These are high-value customers who should be prioritized for retention strategies or premium offers.
- Cluster 1 and 3: Low-value or unique customers who may require re-engagement or customized offers.
- Clusters 5 and 7: Moderate spenders with potential for upselling or cross-selling opportunities.

This segmentation provides actionable insights that can help tailor marketing efforts and improve customer lifetime value.