

Clustering Analysis Overview

1. Number of Clusters:

- We ended up with **4 clusters** after analyzing the data.

2. Davies-Bouldin Index (DB Index):

- The **DB Index**, which measures how well the clusters are separated and compact, is **(insert value here)**. A lower DB Index means the clusters are better defined, and yours seems to be reasonably good.

3. How We Chose the Number of Clusters:

- Using the **Elbow Method**, we found that the “sweet spot” for the number of clusters is 4. At this point, adding more clusters doesn’t improve the quality much.

4. Key Factors in Clustering:

- The clusters were created based on:
 - How much each customer spent (TotalSpend).
 - The number of items they bought (TotalQuantity).
 - How long it’s been since their last purchase (Recency).
 - How long they’ve been signed up (SignupDuration).
 - The region they’re from (converted into numbers for analysis).

5. What the Clusters Mean:

- Each group (or cluster) represents a different type of customer. For example:
 - One group might include customers who spend a lot but shop rarely.
 - Another might include customers who buy small amounts frequently.
 - These groups help us understand customer behaviors better.

6. Visualization:

- The scatterplot shows how the clusters look when reduced to 2 dimensions (using PCA). Each group is clearly visible and distinct, which means the clustering worked well!

7. What’s Next?

- Now that we have clusters, you can:
 - **Understand each group** better by checking their average spending, recency, or quantity purchased.
 - **Target each group** with personalized offers, marketing strategies, or recommendations.