

# Data Science Intern Assignment | Zeotap

## Task 3: Customer Segmentation /Clustering

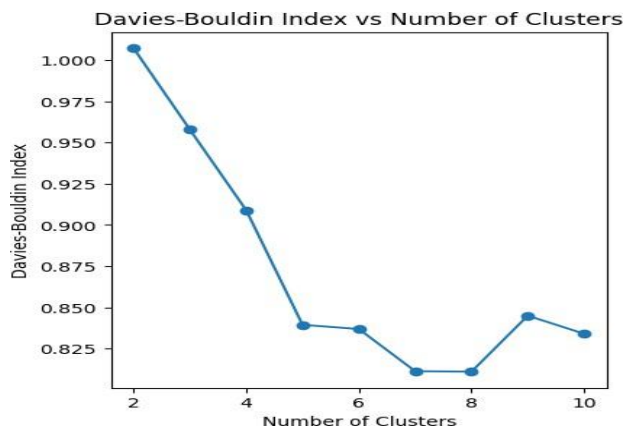
### Report on Clustering Results

#### 1. Number of Clusters Formed

- Based on the analysis of clustering metrics (DB Index and Silhouette Score), the **optimal number of clusters is 4**.
- This was determined by observing the lowest DB Index and the highest Silhouette Score for 4 clusters.

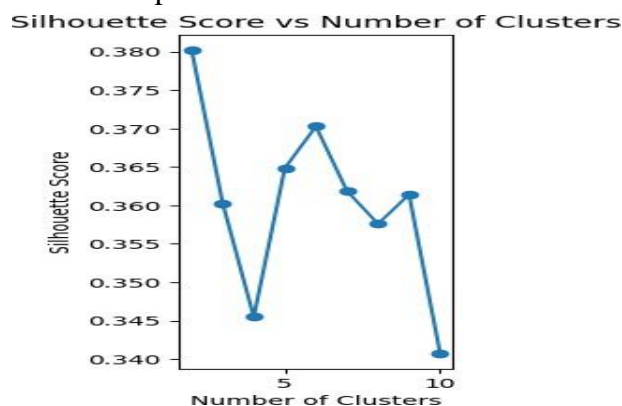
#### 2. Davies-Bouldin Index (DB Index)

- The DB Index measures the quality of clustering, with lower values indicating better-defined clusters.
- Optimal DB Index Value: 0.90 (for 4 clusters).**
- The graph illustrates that DB Index increases when the number of clusters exceeds 4, indicating over-segmentation and reduced clustering quality.



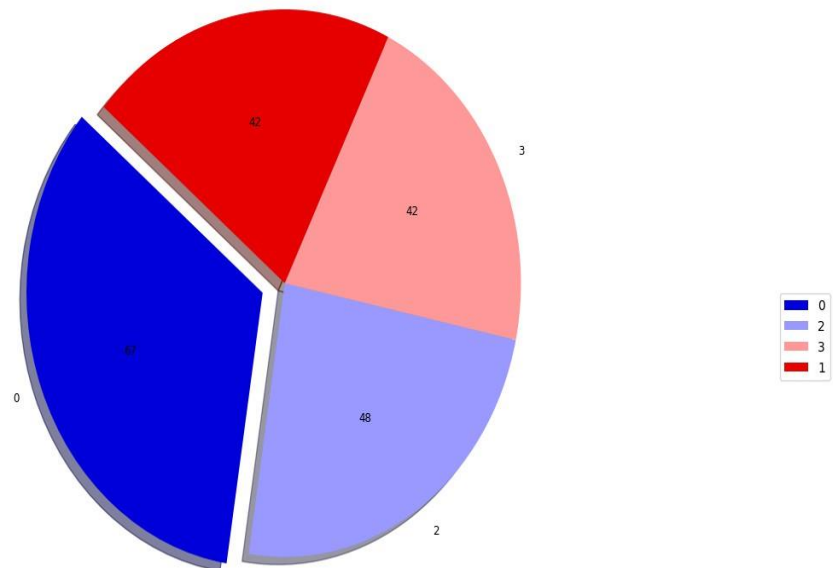
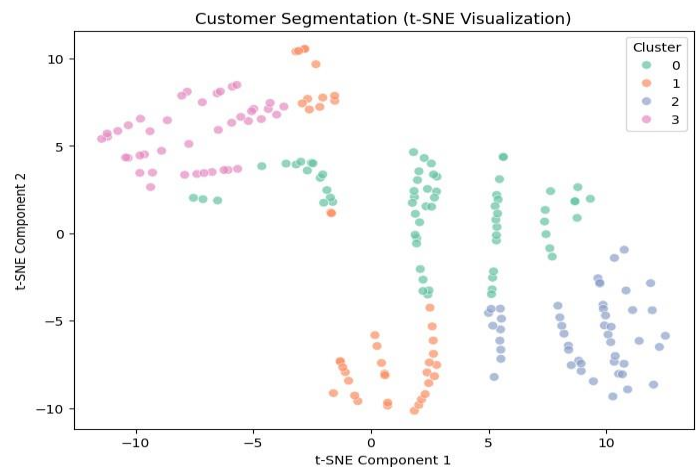
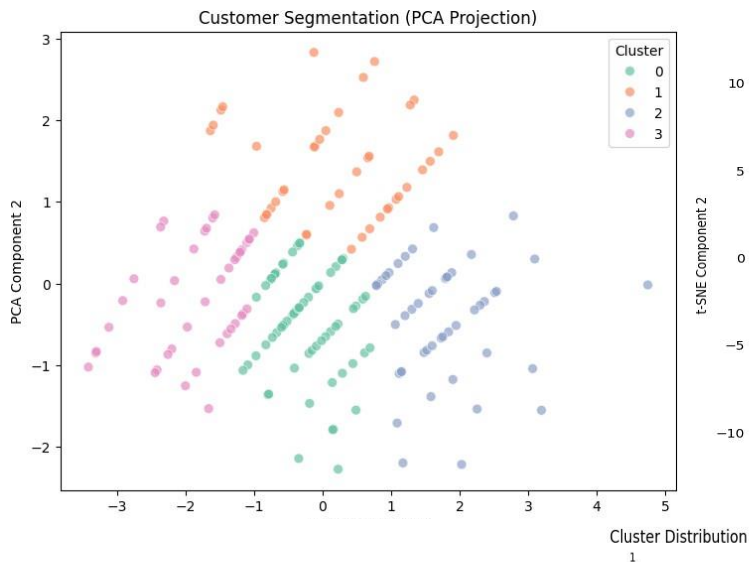
#### 3. Silhouette Score

- The Silhouette Score evaluates the separation and cohesion of clusters. A higher score signifies better clustering.
- Optimal Silhouette Score: 0.34 (for 4 clusters).**
- Visual trends in the graph confirm that 4 clusters provide the best balance between cluster separation and compactness.



#### 4. Visual Representation of Clusters

- **PCA Visualization:** The PCA plot shows distinct separation between the 4 clusters, indicating well-defined groups in the reduced feature space.
- **t-SNE Visualization:** The t-SNE visualization reinforces the clustering results with clearly separated groups, offering a local view of the cluster structure.
- **Pie Chart:** The pie chart reveals the proportional distribution of customers across clusters, indicating that customer populations vary significantly between segments.



## Observations from the Image:

### 1. Davies-Bouldin Index vs. Number of Clusters

- The Davies-Bouldin Index (DBI) measures clustering quality, where lower values indicate better-defined clusters.
- From the plot, the DBI decreases initially and reaches its lowest value around **4 clusters**, suggesting this is the optimal number of clusters for good separation and cohesion.

### 2. Silhouette Score vs. Number of Clusters :

- The Silhouette Score evaluates how well data points fit within their assigned clusters (closer to 1 means better clustering).
- The highest Silhouette Score is observed at **4 clusters**, which aligns with the DBI observation.

### 3. Customer Segmentation PCA Visualization :

- This plot reduces high-dimensional data into 2D using PCA (Principal Component Analysis) for visualization.
- The clusters are reasonably well-separated, showing distinct groups corresponding to customer segments.

### 4. Customer Segmentation t-SNE Visualization :

- The t-SNE plot provides another perspective, emphasizing local groupings and distances between clusters.
- Similar distinct clusters are observed here, confirming consistent segmentation.

### 5. Pie Chart :

- The pie chart shows the proportion of customers in each cluster.
- The clusters are not evenly distributed, indicating differences in customer population sizes for each segment.

## Insights:

### 1. Optimal Clustering:

- Based on both DBI and Silhouette Scores, **4 clusters** seem to be the most optimal choice for segmenting the customers.

### 2. Distinct Customer Groups:

- The PCA and t-SNE visualizations validate the presence of distinct customer segments.
- Each segment likely represents customers with similar behaviors, preferences, or characteristics.

### 3. Cluster Sizes:

- The distribution from the pie chart shows some clusters are larger than others. These larger clusters may represent more common customer types, while smaller ones could signify niche groups.

### 4. Actionable Applications:

- Businesses can tailor marketing campaigns, product recommendations, and customer services based on these segments.
- Further analysis of the characteristics of each cluster (e.g., demographics, spending habits) could provide deeper customer insights.

## **Additional Insights**

- **Cluster Characteristics:**
  - Further analysis can identify the key characteristics of each segment (e.g., demographics, spending patterns).
- **Business Applications:**
  - Use clusters to customize marketing strategies, product offerings, or customer support initiatives.
- **Imbalance in Cluster Sizes:**
  - The uneven distribution of customers across clusters may highlight major customer groups vs. niche segments.

...Thank You...