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Batch: A3

# **Practical 5**

## **Problem Statement:**

- a) Use clustering algorithms (K-Means and Hierarchical Clustering) to categorize customers based on their spending patterns.
- b) Visualize the formed customer segments and measure clustering effectiveness using the Silhouette Score.
- c) Conduct validation through cross-validation or alternative techniques to ensure clustering reliability.

#### **Dataset:**

Download the Mall Customer dataset from:

Mall Customers Dataset – Kaggle (<a href="https://www.kaggle.com/shwetabh123/mall-customers">https://www.kaggle.com/shwetabh123/mall-customers</a>)

This dataset captures demographic and behavioral information of mall visitors. It includes fields such as Customer ID, Gender, Age, Annual Income, and Spending Score (ranging from 1 to 100, reflecting customer spending behavior and engagement).

# **Objectives:**

- 1. Perform data preprocessing including encoding of categorical variables and normalization of numeric features.
- 2. Implement both K-Means and Agglomerative Hierarchical Clustering algorithms.
- 3. Generate visual representations of the resulting clusters and interpret customer groups.
- 4. Evaluate cluster quality using the Silhouette Score.
- 5. Assess the robustness of clustering through different initializations or subsets.

#### **Resources Used:**

- Software: Jupyter Notebook, Visual Studio Code
- Libraries: Pandas, NumPy, Scikit-learn, Seaborn, Matplotlib, Scipy

## **Theory:**

### 1. Clustering

Clustering is an unsupervised machine learning technique that groups data points with similar characteristics. It's commonly used for customer segmentation and uncovering patterns in data.

### 2. K-Means Clustering

K-Means is a centroid-based algorithm that partitions data into K groups. Each point is assigned to the cluster with the closest centroid, and centroids are updated to reduce intracluster variation.

### 3. Hierarchical Clustering

This approach creates a nested cluster structure using either a bottom-up (agglomerative) or top-down (divisive) strategy. The resulting clusters are represented using a dendrogram, which helps in choosing the appropriate number of groups.

## **Methodology:**

## 1. Data Preprocessing

- Load the dataset using Pandas
- Encode categorical variables like Gender using Label Encoding
- Normalize numerical features using MinMaxScaler or StandardScaler

## 2. Clustering Implementation

- Apply K-Means and determine the best number of clusters using the Elbow Method
- Perform Agglomerative Hierarchical Clustering and use a dendrogram to select cluster count

#### 3. Visualization

- Display clusters with scatter plots, particularly using Spending Score and Annual Income
- Plot the dendrogram for a visual understanding of hierarchical clustering

#### 4. Performance Evaluation

- Use the Silhouette Score to measure how well the data fits into the clusters
- Interpret cluster visualizations to identify high-value customer segments

#### 5. Validation

- Run clustering on various subsets or with different initializations
- Compare results to evaluate clustering stability across different runs

## **Conclusion:**

- Implemented both K-Means and Hierarchical Clustering to segment customers into meaningful groups
- Visualized customer clusters to identify potential high-value targets
- Used Silhouette Score to evaluate the clustering effectiveness
- Gained insights into customer behavior which can support strategic marketing
- Future enhancements can involve including more behavioral features or applying advanced methods like DBSCAN for improved segmentation