```
# Step-1: Importing necessary libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.cluster import AgglomerativeClustering
import scipy.cluster.hierarchy as sch
# Step-2: Load the dataset
dataset = pd.read_csv("/content/CLUSTERING.csv")
dataset.describe()
\rightarrow
                    Age Annual Income (k$) Spending Score (1-100)
      count 200.000000
                                  200.000000
                                                            200.000000
                                                                          ıl.
              38.850000
                                   60.560000
                                                             50.200000
      mean
       std
              13.969007
                                   26.264721
                                                             25.823522
              18.000000
                                   15.000000
                                                              1.000000
       min
       25%
              28.750000
                                   41.500000
                                                             34.750000
       50%
              36.000000
                                   61.500000
                                                             50.000000
       75%
              49.000000
                                   78.000000
                                                             73 000000
              70.000000
                                   137.000000
                                                             99.000000
       max
# Step-3: Extracting the features
\ensuremath{\text{\#}} Here we are assuming that all columns are features
x = dataset.iloc[:, :].values
# Step-4: Build the Dendrogram
# This helps visualize the hierarchical clustering process
# We start with each data point as a single cluster and merge them step by step.
dendrogram = sch.dendrogram(sch.linkage(x, method="ward"))
plt.title("Dendrogram")
plt.xlabel("Data Points (or Customers)")
plt.ylabel("Euclidean Distance")
plt.show()
\rightarrow
                                          Dendrogram
         400
         350
         300
      Euclidean Distance
         250
         200
         150
```

```
# Step-5: Determine the number of clusters using the dendrogram (e.g., 5 clusters)
# Using Agglomerative Clustering to apply hierarchical clustering

HC = AgglomerativeClustering(n_clusters=5, linkage='ward')
predicted = HC.fit_predict(x)
```

Data Points (or Customers)

# Step-6: Visualize the Clusters

100

50

 $\ensuremath{\text{\#}}$  This plot shows clusters based on two selected features.

 $\overline{\Rightarrow}$ 

```
plt.figure(figsize=(10, 7))
plt.scatter(x[predicted == 0, 0], x[predicted == 0, 1], c='red', label='Cluster 1')
plt.scatter(x[predicted == 1, 0], x[predicted == 1, 1], c='blue', label='Cluster 2')
plt.scatter(x[predicted == 2, 0], x[predicted == 2, 1], c='green', label='Cluster 3')
plt.scatter(x[predicted == 3, 0], x[predicted == 3, 1], c='yellow', label='Cluster 4')
plt.scatter(x[predicted == 4, 0], x[predicted == 4, 1], c='orange', label='Cluster 5')
plt.title("Clusters of Customers")
plt.xlabel("Feature 1 (e.g., Annual Income)")
plt.ylabel("Feature 2 (e.g., Spending Score)")
plt.legend()
plt.show()
```

