## **Hierarchical Clustering**

### Importing the libraries

```
In [3]: import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
```

### Importing the dataset

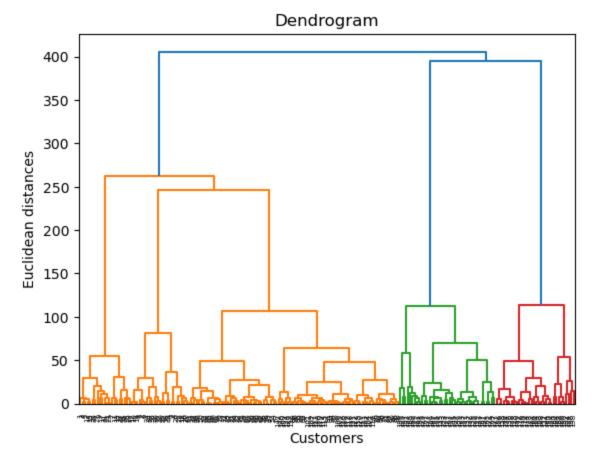
```
In [4]: df = pd.read_csv(r"C:\HP LAPTOP\EDUSNUG\FINAL VIDEOS\Python\PROJECT\Customers_spending.c
```

retrieve the values from columns 3 and 4 of the df (all rows) and return them as a NumPy array.

```
In [5]: x = df.iloc[:, [3, 4]].values
```

# Using the dendrogram to find the optimal number of clusters

```
In [6]: import scipy.cluster.hierarchy as sch
    dendrogram = sch.dendrogram(sch.linkage(x, method = 'ward'))
    plt.title('Dendrogram')
    plt.xlabel('Customers')
    plt.ylabel('Euclidean distances')
    plt.show()
```



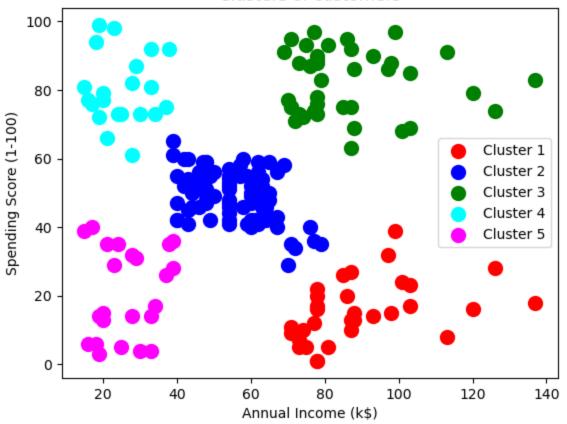
### Training the Hierarchical Clustering model on the dataset

```
In [7]: from sklearn.cluster import AgglomerativeClustering
hc = AgglomerativeClustering(n_clusters = 5, affinity = 'euclidean', linkage = 'ward')
y_hc = hc.fit_predict(x)
```

### Visualising the clusters

```
In [8]: plt.scatter(x[y_hc == 0, 0], x[y_hc == 0, 1], s = 100, c = 'red', label = 'Cluster 1')
    plt.scatter(x[y_hc == 1, 0], x[y_hc == 1, 1], s = 100, c = 'blue', label = 'Cluster 2')
    plt.scatter(x[y_hc == 2, 0], x[y_hc == 2, 1], s = 100, c = 'green', label = 'Cluster 3')
    plt.scatter(x[y_hc == 3, 0], x[y_hc == 3, 1], s = 100, c = 'cyan', label = 'Cluster 4')
    plt.scatter(x[y_hc == 4, 0], x[y_hc == 4, 1], s = 100, c = 'magenta', label = 'Cluster 5
    plt.title('Clusters of customers')
    plt.xlabel('Annual Income (k$)')
    plt.ylabel('Spending Score (1-100)')
    plt.legend()
    plt.show()
```

#### Clusters of customers



```
In []:
```