

## Experiment: 6

**Aim: To demonstrate hierarchical clustering on given data set.**

### **Theory:**

Clustering is one of the most common exploratory data analysis techniques used to get an intuition about the structure of the data. It can be defined as the task of identifying subgroups in the data such that data points in the same subgroup (cluster) are very similar while data points in different clusters are very different. In other words, we try to find homogeneous subgroups within the data such that data points in each cluster are as similar as possible according to a similarity measure such as euclidean-based distance or correlation-based distance. The decision of which similarity measure to use is application-specific.

### **Hierarchical clustering:**

Hierarchical clustering, also known as Hierarchical cluster analysis or HCA, is an unsupervised clustering approach that includes forming groups with a top-to-bottom order.

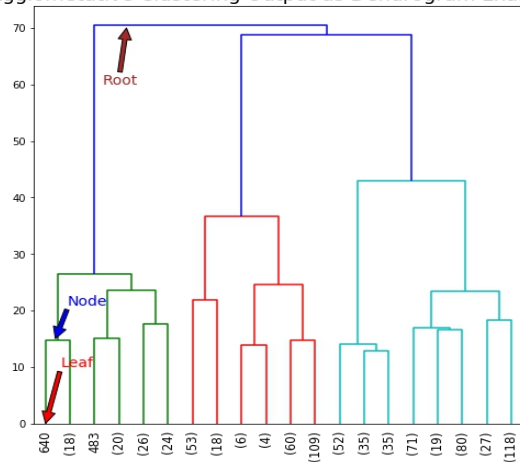
The hierarchical clustering technique has two approaches:

1. **Agglomerative:** Agglomerative is a **bottom-up** approach, in which the algorithm starts with taking all data points as single clusters and merging them until one cluster is left.
2. **Divisive:** Divisive algorithm is the reverse of the agglomerative algorithm as it is a **top-down approach**.

### **Agglomerative Hierarchical clustering:**

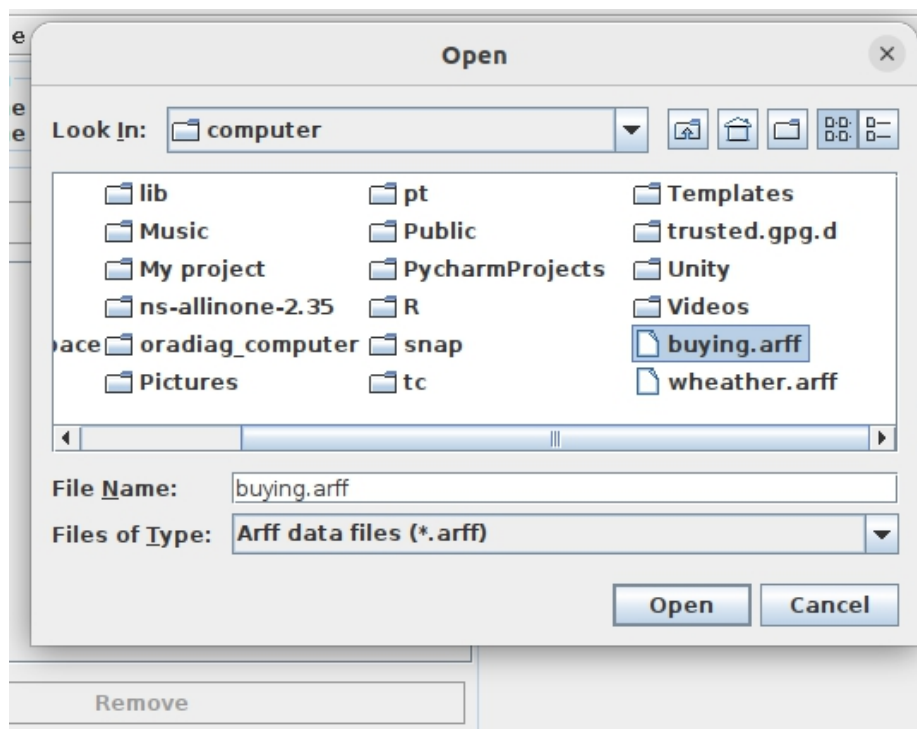
Agglomerative Clustering or bottom-up clustering essentially started from an **individual cluster** (each data point is considered as an individual cluster, also called **leaf**), then every cluster calculates their **distance** with each other. The two clusters with the shortest distance with each other would **merge** creating what we called **node**. Newly formed clusters once again calculating the member of their cluster distance with another cluster outside of their cluster. The process is repeated until all the data points assigned to one cluster called **root**. The result is a tree-based representation of the objects called **dendrogram**.

Agglomerative Clustering Output as Dendrogram Example



### Steps to be followed:

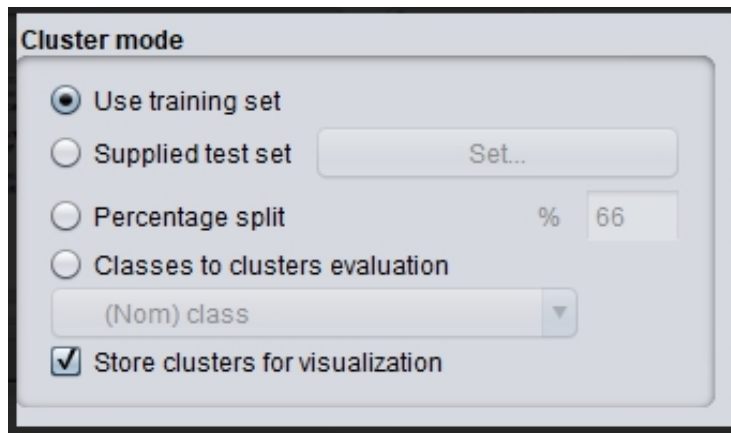
**Step 1:** Open the Weka explorer in the preprocessing interface and import the appropriate dataset;



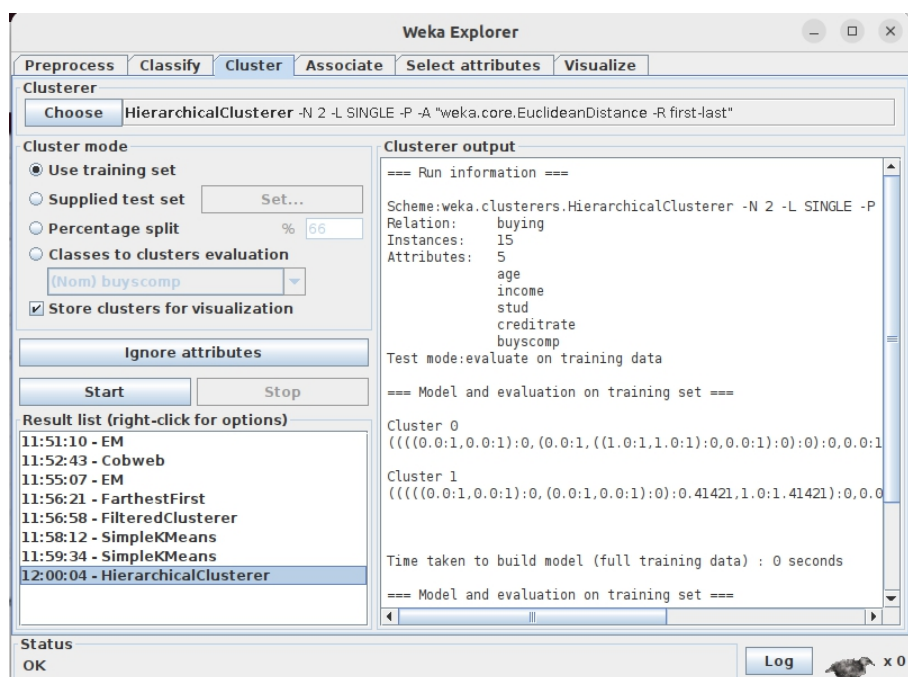
**Step 2:** To perform clustering, go to the explorer's 'cluster' tab and select the select button. As a result of this step, a dropdown list of available clustering algorithms displays; pick the Hierarchical algorithm.

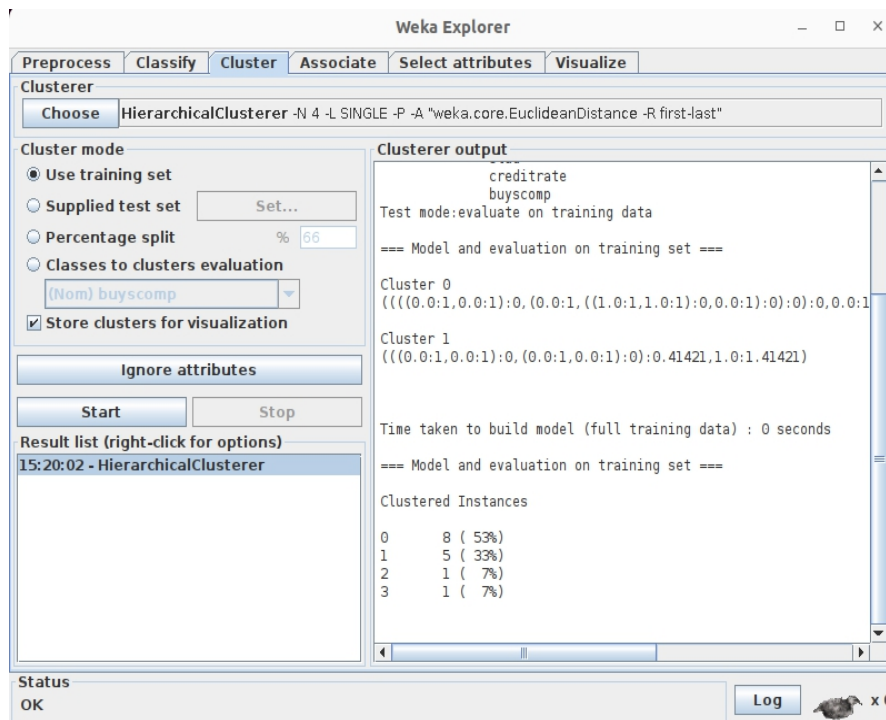
**Step 3:** Then press the text button to the right of the pick icon to bring up the popup window seen in the screenshots. In this window, we input three for the number of clusters and leave the seed value alone. The seed value is used to generate a random number that is used to allocate cluster instances to each other internally.

**Step 4:** One of the options has been selected. Before we execute the clustering method, we need to make sure they're in the 'cluster mode' panel. The option to employ a training set is chosen, after which the 'start' button is hit. The process and the resulting window are depicted in the screenshots below.

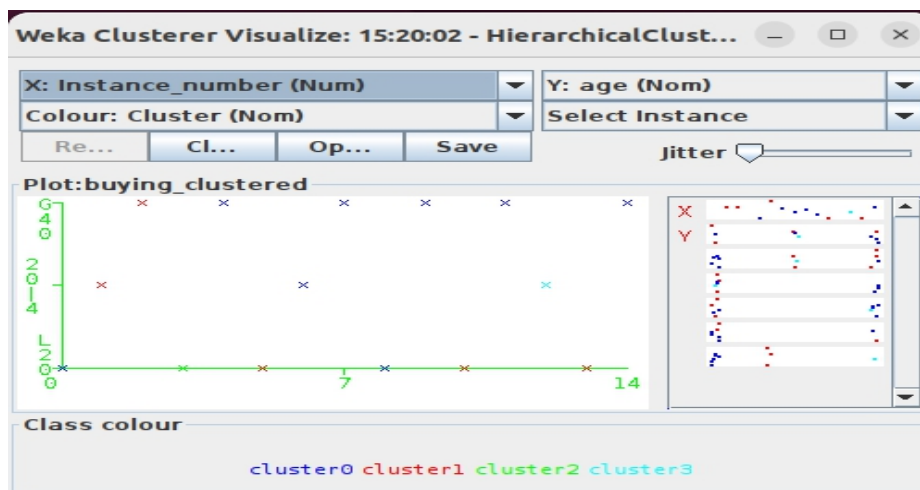


**Step 5:** The resulting window displays the centroid of each cluster, as well as data on the number and proportion of instances assigned to each cluster. A mean vector is used to represent each cluster centroid. A cluster can be described using this cluster.





**Step 6:** Visualizing the qualities of each cluster is another approach to grasp them. Right-click the result set on the result to do so. To visualize cluster assignments are selected from the list column.



**Conclusion:** Thus we have implemented hierarchical algorithm for clustering successfully.

**Viva Voce**

**Q.1** What is a Hierarchical Clustering Algorithm?

**Q.2** What are the various types of Hierarchical Clustering?

**Q.3 What is a dendrogram in Hierarchical Clustering Algorithm?**

**Q.4 Explain the different linkage methods used in the Hierarchical Clustering Algorithm.**