Vidyavardhini's College of Engineering and Technology

Department of Artificial Intelligence & Data Science

Name:	Vinith Shetty
Roll No:	52
Class/Sem:	TE/V
Experiment No.:	6
Title:	Using open-source tools Implement Clustering Algorithms.
Date of Performance:	
Date of Submission:	
Marks:	
Sign of Faculty:	

Aim: To implement k-means Algorithm on large dataset using Open source tool WEKA.

Objective: To make students well versed with open source tool like WEKA to implement k-means algorithm.

Theory:
- **The process of grouping a set of physical or abstract objects into classes of similar objects is called clustering.**
- **A cluster is a collection of data objects that are similar to one another within the same cluster and are dissimilar to the objects in other clusters.**
- **A cluster of data objects can be treated collectively as one group and so may be considered as a form of data compression.**
- **Cluster analysis is an important human activity. Cluster analysis has been widely used in numerous applications including market research, pattern recognition, data analysis and image processing.**
- **Clustering is also called data segmentation in some applications because clustering partitions large data sets into groups according to their similarity.**
- **Clustering can also be used in outlier detection where outliers may be more interesting than common case.**

WEKA contains "clusterers" for finding groups of similar instances in a dataset. The clustering schemes available in WEKA are k-means, Cobwebs, DBSCAN, OPTICS. Clusters can be visualized and compared to true clusters. Evaluation is based on log likelihood if clustering scheme produces a probability distribution.In 'preprocess' window click on 'open file...' button to select data file. Choosing Clustering scheme: In the 'clusterer' box click on 'choose' button. In the pull-down menu select WEKA Clustering, and select the cluster scheme 'simple K means'. Some implementations of K -means only allow numerical values for attributes ; therefore we do not need to use a filter.

Once the clustering algorithm is chosen, right click on algorithm, 'weak.gui.GenericObjectEditor' comes up to the screen. Set the value in 'numclusters' box to number of clusters required. The seed value is used in generating a random number, which is used for making the initial assignments of instances to clusters. Before we run the clustering algorithm, we need to select 'cluster mode'. Click on 'Classes to cluster evaluation' radio-button in 'Cluster mode' box. Click the start button to run the program. When the training set is complete, the 'Cluster' output area on the right panel of 'Cluster' window is filled with text describing the results of training and testing. A new entry appears in the 'Result list' box on the left of the result. Run information gives the information about the clustering scheme used, the relation name, the number of instances, number of attributes. The clustering model shows the centroid of each cluster and statistics on the number and percentage of instances assigned to different clusters. Cluster centroid is the mean vector of each cluster, so each dimension value and centroid represent the mean value for that dimension in the cluster. Thus, centroids can be used to characterize the cluster.
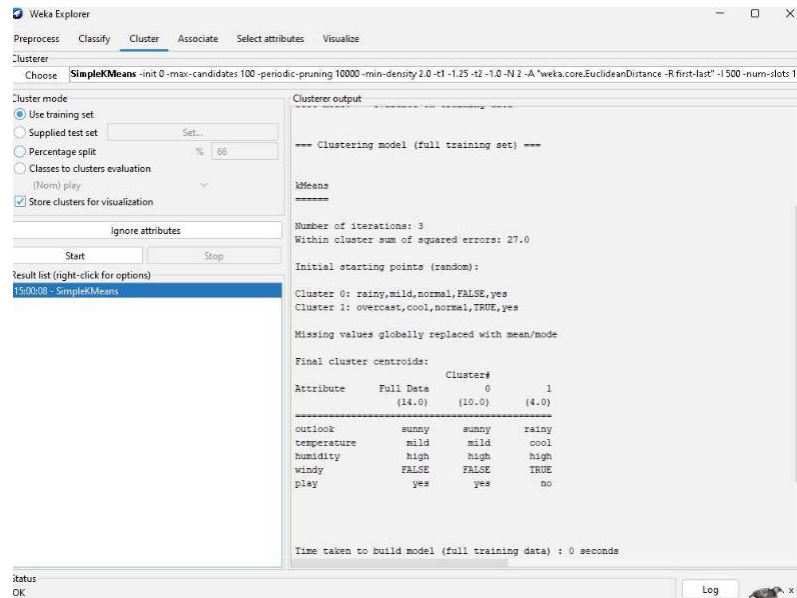
Another way of representation of results of clustering is through visualization. Right click on the entry in the 'Result list' and select ' Visualize cluster assignments' in the pull-down window. This brings up the Weka clustered visualize window. This window displays clusters in different colors for better visibility.

Output:

Conclusion:
What are the strengths and weaknesses of the K-Means algorithm?

Strengths of K-Means:
1. **Simplicity: Easy to understand and implement.**
2. **Scalability: Efficient on large datasets, especially with optimized implementations.**
3. **Speed: Fast convergence compared to other clustering algorithms.**
4. **Flexibility: Works well with distinct, well-separated clusters.**

Weaknesses of K-Means:
1. **Fixed number of clusters: Requires specifying the number of clusters (K) in advance.**
2. **Sensitive to initialization: Different initial centroids can lead to different results.**
3. **Assumes spherical clusters: Struggles with non-spherical, overlapping, or varying-sized clusters.**
4. **Outlier sensitivity: Easily affected by outliers, which can distort clusters.**
5. **Requires numerical data: Not suitable for categorical data without modification.**