# MLT LAB

# Q1. Pre-Pruning and Post-Pruning Techniques in Decision Trees

- 1. Load the Iris dataset using sklearn.datasets.load\_iris.
- 2. Split the dataset into training (70%) and testing (30%) sets.
- 3. Build a Decision Tree Classifier using sklearn.tree.DecisionTreeClassifier.

### (a) Implement Pre-Pruning:

- Use parameters such as max\_depth, min\_samples\_split, and min\_samples\_leaf to limit tree growth.
  - Train and evaluate the model accuracy on both training and testing sets.

# (b) Implement Post-Pruning:

- Use the parameter ccp\_alpha to prune the tree.
- Re-train and evaluate the model.

# (c) Display and Compare:

- Tree depth and number of nodes before and after pruning.
- Accuracy and confusion matrix for both models.

#### Q2. Naive Bayes on Iris Dataset

- 1. Load the Iris dataset using sklearn.datasets.load\_iris.
- 2. Split the dataset into training (70%) and testing (30%) sets.
- (a) Train a Gaussian Naive Bayes classifier (sklearn.naive\_bayes.GaussianNB). Evaluate accuracy and display the confusion matrix.

# (b) Compare with Another Naive Bayes Variant:

- Train a Multinomial Naive Bayes model on the same dataset.
- Compare the performance of both models using accuracy and F1-score.