



Lab 4: Decision Trees and Random Forests

1. Load the breast_cancer data. Train a decision tree using the C4.5 algorithm and k-fold cross validation.
 - a. What is the mean accuracy and standard deviation achieved? What confidence do you have in that estimation for the accuracy?
 - b. Change the minimum number of instances to split a node and study the accuracy.
 - c. Compare the trees discovered with and without pruning.
 - d. Study the effect of using just binary splits.
 - e. Repeat the procedures with CART.
 - f. Compare the performance of C4.5, CART and DecisionStumps.
2. Splitting the same dataset into train and test datasets, train a random forest.
 - a. What is the accuracy achieved with default parameters?
 - b. Change the minimum number of trees to use and study the accuracy achieved.
 - c. Change the maximum tree size to use and study the accuracy achieved.
 - d. Is there any model in overfitting?
3. Load the credit data. Evaluate the accuracy on the training and testing datasets separately, study the overfitting of the different models trained.
 - a. Run C4.5, changing the number of instances to split a node.
 - b. Run RandomForest, changing the maximum tree size to combine in the ensembler.
4. Load the diabetes data. Study the role of each pre-processing technique when learning decision trees and random forests.
 - a. **discretization.**
 - b. **normalization.**
 - c. **feature selection.**