

PROBLEMS 5: DECISION TREE AND ENSEMBLE MODELS

GOAL

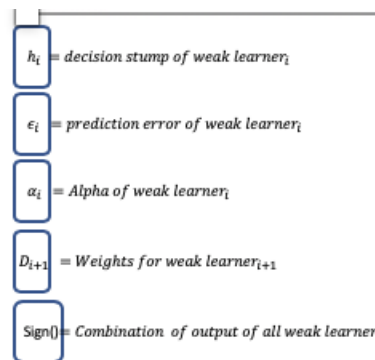
The goal of this practice is to understand how Decision Tree and Ensemble models are defined and work in supervised classification problems.

NEEDED CONCEPTS

- $Gini = \sum_{l=1}^K p_{jl}^2$
- $Gini\ Impurity = 1 - \sum_{l=1}^K p_{jl}^2$
- $Entropy = \sum_{l=1}^K *log_2 p_{jl}$
- $Information\ Gain = 1 - \sum_{l=1}^K *log_2 p_{jl}$
- $Weighted\ Gini\ impurity = 1 - \sum_{j=1}^L W_j * \sum_{l=1}^K *log_2 p_{jl}$

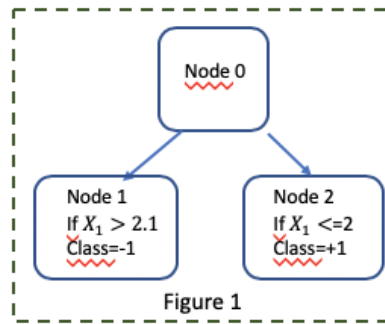
EXERCISE 1

1. **Decision Tree:** Consider a training dataset with two classes: the class $C_1 = \{\mathbf{x}_1 = (\frac{2}{3}), \mathbf{x}_2 = (\frac{2.1}{2}), \mathbf{x}_3 = (\frac{4.5}{6}), \mathbf{x}_6 = (\frac{5}{7}), \mathbf{x}_8 = (\frac{6}{5.5})\}$ with a label 1 and the class $C_2 = \{\mathbf{x}_4 = (\frac{4}{3.5}), \mathbf{x}_5 = (\frac{3.5}{1}), \mathbf{x}_7 = (\frac{5}{3}), \mathbf{x}_9 = (\frac{8}{6}), \mathbf{x}_{10} = (\frac{8}{2})\}$ with label -1.
 - (a) Plot the points. Is it feasible to find a linear “decision border” to classify both classes?
 - (b) Draw the node splits decision (or decision stumps) a decision tree in the scatter plot with the following hyperparameters: Minimum samples for a node split= 2 Minimum samples for a terminal node or leaf=2 Maximum depth of tree (vertical depth)=3
2. **Adaboost:** Using the Decision stump from previous exercise, develop an adaboost ensemble architecture with four weak learners.
 - (a) Considering the following pseudo-blocks, build an architecture for a 4-estimator (weak learners) adaboost ensemble:



- (b) First round:

Consider the decision stump of figure 1, build a table with the actual class, the weight, prediction, loss and weight*loss for every datapoint



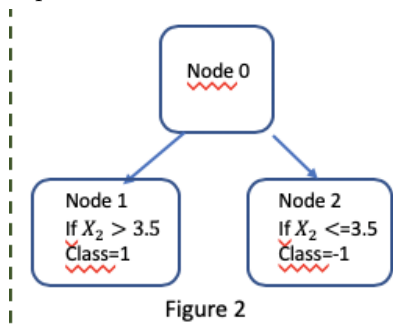
Calculate the error and alpha

Update the weights and normalized weights for every datapoint for next round

(c) Second round:

Plot the points which sizes should be aligned with NormD2(i) value

Consider the decision stump of figure 2, build a table with the actual class, the weight, prediction, loss and weight*loss for every datapoint



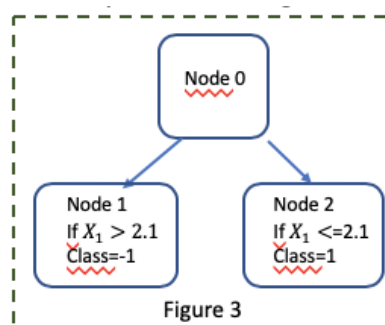
Calculate the error and alpha

Update the weights and normalized weights for every datapoint for next round

(d) Third round:

Plot the points which sizes should be aligned with NormD3(i) value

Consider the decision stump of figure 3, build a table with the actual class, the weight, prediction, loss and weight*loss for every datapoint



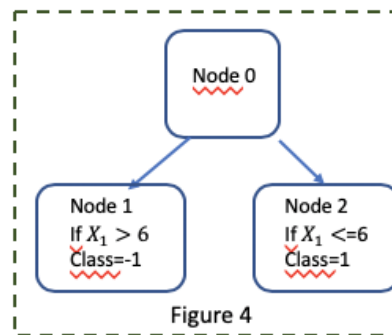
Calculate the error and alpha

Update the weights and normalized weights for every datapoint for next round

(e) Four round:

Plot the points which sizes should be aligned with NormD4(i) value

Consider the decision stump of figure 4, build a table with the actual class, the weight, prediction, loss and weight*loss for every datapoint



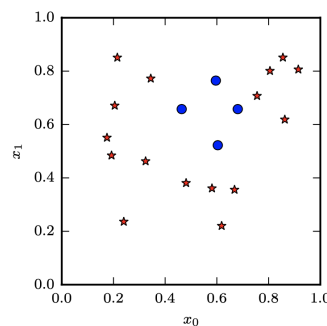
Calculate the error and alpha

Update the weights and normalized weights for every datapoint for next round

(f) Calculate the prediction for $C_1 = \{\mathbf{x}_1 = (\frac{2}{3})\}$:

(g) Draw the decision areas in the Adaboost classifier:

3. **CART:** Consider the following scatter plot of two classes (blue and red blobs) based on two variables x_0 and x_1 .



- Calculate the Gini and Gini impurity values of the current distributions
 - Consider $x_0=0.4$ as split value to create subnodes. Plot the selected split value and calculate the Gini and Gini impurity of the new subnodes. Which is the selected class in both nodes? Calculate the Weighted Gini for $x_0=0.4$.
 - Define a new split value in x_0 . Plot the selected split value and calculate the Gini and Gini impurity of the new subnodes. Which is the selected class in both nodes? Calculate the Weighted Gini for this new split.
 - Define a split value in x_1 . Plot the selected split value and calculate the Gini and Gini impurity of the new subnodes. Which is the selected class in both nodes? Calculate the Weighted Gini for this new split.
 - Calculate the Entropy and Information Gain in Node 0 and Node 4.
 - Plot the diagram of the full decision tree with node and subnodes.
 - Which is the predicted class for a new blob with value $\{\mathbf{x}_{new} = (\frac{0.8}{0.2})\}$?
4. (To be executed in Jupyter Notebook) In this exercise we will execute a decision tree with the Sklearn library for the Iris multiclassification. We will use the graphviz library to evaluate the node split.
5. (To be executed in Jupyter Notebook) In this exercise we will execute change the decision tree for an adaboost ensemble for the same dataset in the previous exercise.