## **Decision Trees**

## Applied ML in Engineering - Exercise 06

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Students are asked to implement a basic decision tree model from scratch using object-oriented programming in Python. The exemplary data set, also used in the lecture slides, is provided in the file  $decision\_tree\_dataset.txt$  where the first two columns provide the coordinates and the last column provides the targets (0 and 1 for the binary classification problem). Remember that the left child of a split will always carry all data points that are  $\leq$  the splitting threshold, and the right child will contain all data that are > the splitting condition.

## **Problem 1**

Implement various helper functions that will be required for the generation of a decision tree:

- def entropy(y) where y is a vector of labels (integers). Function returns the information entropy H(x).
- def information\_gain(y\_parent, index\_split) where y\_parent gives the labels of the parent node index\_split is a binary vector, 1 indicating samples assigned to the left child node, and 0 indicating samples assigned to the right child node. Returns the information gain that given split.
- def best\_split(X, y) where X is the data, and y is the distribution of corresponding labels at the parent node. Returns split\_dim, e.g. 0 for the first feature dimension, and split\_val, e.g. 1.5 for the decision rule  $x_1 \le 1.5$ , indicating the best split w.r.t. information gain given the current data and labels.
- def create\_split(X, split\_dim, split\_val where X is the data at the parent node and the splitting definition as in the previous point. Returns a boolean vector split\_idx that indicates assignment to the left child by 1 and to the right child by 0.

## **Problem 2**

Validate your implementation using the data set provided (or a self-generated data set) and manual computation of the entropy and information gain to check your code. Define the main for testing as if \_\_name\_\_ == \_\_main\_\_: . Implement a plotting routine for labeling the data points and showing the decision boundary (given by split\_dim and split\_val, and also returning the entropy and information gain values in the title of the figure.