

Intelligent Systems

Exercise sheet 10

Rule Learning

Exercise 1¹ (8 points)

Give decision trees to represent the following boolean functions:

(a) $A \wedge \neg B$

(b) $A \vee [B \wedge C]$

Exercise 2¹ (10 points)

Consider the following set of training examples:

| Instance | Classification | a_1 | a_2 |
|----------|----------------|-------|-------|
| 1 | + | T | T |
| 2 | + | T | T |
| 3 | - | T | F |
| 4 | + | F | F |
| 5 | - | F | T |
| 6 | - | F | T |

Figure 1: Test set of training examples

¹ Exercise from T. Mitchell

- (a) What is the entropy of this collection of training examples with respect to the target function "Classification"?
- (b) What is the information gain of a_2 relative to these training examples?

Exercise 3¹ (13 points)

Experiment with the PlayTennis data

- (a) Implement the basic ID3 algorithm (for discrete attributes only).
- (b) Take a look at the training data in Figure 2, and then run your decision tree learning program on all the training examples to be sure it produces the correct decision tree for this data (it should produce the decision tree shown in Figure 3). Now try running it on a randomly chosen subset containing half of the examples for training, and using half for test. What are the training and test accuracies?

| Day | Outlook | Temperature | Humidity | Wind | PlayTennis |
|-----|----------|-------------|----------|--------|------------|
| D1 | Sunny | Hot | High | Weak | No |
| D2 | Sunny | Hot | High | Strong | No |
| D3 | Overcast | Hot | High | Weak | Yes |
| D4 | Rain | Mild | High | Weak | Yes |
| D5 | Rain | Cool | Normal | Weak | Yes |
| D6 | Rain | Cool | Normal | Strong | No |
| D7 | Overcast | Cool | Normal | Strong | Yes |
| D8 | Sunny | Mild | High | Weak | No |
| D9 | Sunny | Cool | Normal | Weak | Yes |
| D10 | Rain | Mild | Normal | Weak | Yes |
| D11 | Sunny | Mild | Normal | Strong | Yes |
| D12 | Overcast | Mild | High | Strong | Yes |
| D13 | Overcast | Hot | Normal | Weak | Yes |
| D14 | Rain | Mild | High | Strong | No |

Figure 2 - Training data

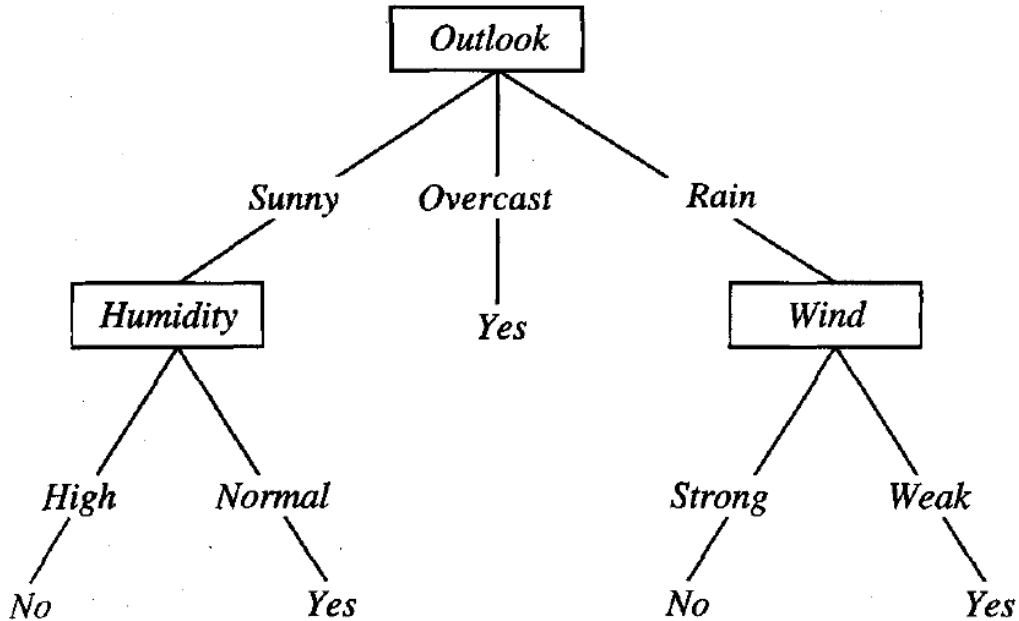


Figure 3 - PlayTennis decision tree

- (c) Answer whether or not the actions in the next three questions are possible or not. If possible, demonstrate it by running the decision tree learner, and turn in the parameter settings you used and your data set. If impossible, explain why. When answering the questions below, assume that (i) the target concept is the concept described by the decision tree of Figure 3; (ii) all training examples you add must be consistent with this target concept (meaning that the classification you include for the example must be the one with which the tree in Figure 3 would label it), and (iii) you are not allowed to include post pruning when learning decision trees.
- Is it possible to get ID3 to further elaborate the tree below the rightmost leaf in Figure 3 (and make no other changes to the tree), by adding a single new correct training example to the original fourteen examples?
 - Is it possible to get ID3 to learn an incorrect tree (i.e., a tree that is not equivalent to the target concept of Figure 3) by adding new correct training examples to the original fourteen?
 - Is it possible to produce some set of correct training examples that will get ID3 to include the attribute "Temperature" in the learned tree, even though the true target concept is independent of "Temperature"?