3.1. Should we play golf?

The *Golf data* set is one of the examples that are delivered together with RapidMiner. The data set models different aspects of the weather (outlook, temperature, humidity, forecast) that are relevant for deciding whether one should play golf or not.

1. Learn a decision tree from the Golf data set (Operator: *Decision Tree*). Use this tree to classify the examples in the Golf-Testset, which is also delivered together with RapidMiner (Operator: *Apply Model*).

2. Evaluate the performance of your model by adding a *Performance (Classification)* operator to your process. Examine the confusion matrix. What is the accuracy of your classifier?

3. Does a k-nearest-neighbor classifier work better for this task? Replace the *Decision Tree* operator with a *K-NN* operator and check how the accuracy of your classifier changes to find out. Do different values of k improve the performance?

3.2. Learning a classifier for the Iris Data Set

You want to learn and evaluate a classifier for recognizing different types of Iris flowers.

1. Let’s try the ID3 tree building algorithm first. Build a process that (1) discretizes all attributes of the Iris data set by frequency into three bins. (2) Afterwards, the process should use the *Split Validation* operator (split ratio=0.7, stratified sampling) to generate a training and test data set. (3) As inner operator of the split validation, the process should use the *ID3* operator to learn a decision tree and the *Performance (Classification)* operator to evaluate the accuracy of the learned model.

2. Remove the discretization operator and change the ID3 operator into RapidMiner’s standard *DecisionTree* building operator. Run the process again. Does the accuracy change? Compare the complexity of the two models. Which model should be preferred according to Occam’s razor?

3. Try a k-nearest-neighbor classifier on the problem. Does it perform better?