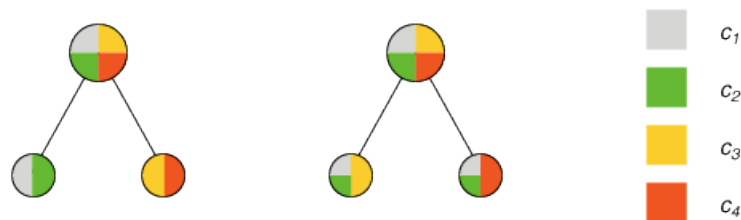


Exercise 5

Learning Goals

- Decision Trees
 - Impurity functions
 - The ID3 algorithm
-

- What is overfitting?
 - Check all correct statements.
 - ☐ A short training time leads to overfitting.
 - ☐ A smaller decision tree generalizes better to unseen examples than a larger decision tree.
 - ☐ The generalization capability of a decision tree depends on the training set.
 - ☐ By using entropy as an impurity measure, overfitting can no longer occur.
- For each of the following boolean functions: Draw a decision tree that represents the boolean function. All variables take values in $\{0, 1\}$. (You do not need to apply any algorithm for this exercise)
 - $x_1 \wedge \neg x_2$
 - $x_1 \text{ XOR } x_2$
 - $x_1 \vee (x_2 \wedge x_3)$
- Consider the following illustration of two possible splittings of an example set D with four classes $C = \{c_1, c_2, c_3, c_4\}$. Assume that in both splittings the resulting subsets each contain $\frac{1}{2}$ of the examples from D .



- Compute the drop in impurity $\Delta \iota$ for the left split and the right split, respectively, using the misclassification rate ι_{misclass} as well as the Gini impurity ι_{Gini} . (Note: This will yield 4 values: one for every combination of “split” and “impurity measure”.)
 - Which splitting is preferable under each impurity measure?
- Implement the impurity functions in section 2.1 of the jupyter notebook and implement the impurity reduction calculation in section 2.2 of the jupyter notebook.