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## Machine Learning - Sheet 1

19.04.2018

Deadline: 26.04.2018 - 12:00

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### Task 1: Boolean Functions

(5 Points)

Give decision trees to represent the following Boolean functions:

- (a)  $A \wedge \neg B$
- (b)  $A \vee (B \wedge C)$
- (c)  $A \text{ xor } B$
- (d)  $(A \wedge B) \vee (C \wedge D)$
- (e)  $M$  of  $N$ : a function that depends on variables  $X_1, \dots, X_N$ , and returns **true** if and only if exactly  $M$  of the  $N$  variables are **true**.

### Task 2: Literature

(0 Points)

Read pages 55-60 of the book Machine Learning [1].

### Task 3: Entropy

(6 Points)

An information source emits four symbols  $A, B, C, D$ , which occur with probabilities  $p(A), \dots, p(D)$ .

1. Suppose that  $p(A) = 0.5$ ,  $p(B) = 0.3$ ,  $p(C) = 0.1$ ,  $p(D) = 0.1$ . Compute the entropy  $H(p)$ .
2. For which probabilities is the entropy  $H$  maximal? For which is it minimal?
3. Explain in your own words what the entropy is.
4. How is the entropy used for the decision tree construction?

### Task 4: ARFF Files

(0 Points)

Make yourself familiar with ARFF files (see <http://www.cs.waikato.ac.nz/ml/weka/arff.html>).

### Task 5: Information Gain

(9 Points)

Implement the following methods in Java or Python:

1. Implement the method `entropyOnSubset` that takes three arguments: a dataset  $D = [inst_0, \dots, inst_{N-1}]$ , a list of indices  $I = [i_0, i_1, \dots, i_{m-1}]$  that describe a subset of the dataset, and a class attribute  $C$ . The subset of the dataset is then defined as

$$S := \{inst_{i_j} | 0 \leq j < m\}.$$

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The entropy of  $S$  relative to the  $C$ -wise classification is then

$$H(S) := - \sum_{v \in \text{values}(C)} p_v \cdot \log_2(p_v),$$

where  $p_v$  is the proportion of  $S$  belonging to class  $v$ . The value  $H(S)$  is what the `entropyOnSubset`-method should return.

2. Let  $D$ ,  $I$ ,  $S$  and  $C$  be as above. Implement the method `informationGain` that takes  $D$ ,  $I$ ,  $C$  and an additional attribute  $A$  as arguments, and returns

$$\text{InformationGain}(S, A) := H(S) - \sum_{v \in \text{values}(A)} \frac{|S_v|}{|S|} \cdot H(S_v),$$

where  $S_v$  are those instances that take value  $v$  at attribute  $A$ .

3. Implement an ARFF file parser and test your codes from previous steps on the Weather dataset (`weather.nominal.arff`).

## References

- [1] Tom M. Mitchell. *Machine learning*. McGraw Hill series in computer science. McGraw-Hill, 1997.