

# Assignment 1

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## Contents

<b>1</b>	<b>Question 1</b>	<b>1</b>
<b>2</b>	<b>Question 2</b>	<b>1</b>
<b>3</b>	<b>Question 3</b>	<b>2</b>
<b>4</b>	<b>Question 4</b>	<b>2</b>
<b>5</b>	<b>Question 5</b>	<b>2</b>
5.1	Question 5.1 . . . . .	2
<b>6</b>	<b>Question 6</b>	<b>3</b>

## 1 Question 1

The book *Introduction to Machine Learning* by Nils J. Nilsson can be found at <http://robotics.stanford.edu/people/nilsson/MLBOOK.pdf> and is 1.855 MB.

The book *A first encounter with Machine Learning* by Max Welling can be found at <http://citeseerx.ist.psu.edu/viewdoc/download?doi=10.1.1.441.6238&rep=rep1&type=pdf> and is 416 KB.

## 2 Question 2

We can represent some learning functions in machine learning as Boolean function. We can define a Boolean function as a function of the form

$$f(x_1, x_2, x_3, \dots, x_n) \quad (1)$$

Which maps a n-tuple of (0,1) values to 0,1. 0,1 can also be expressed as false,true.

There are 3 basic types of operations can be performed in Boolean functions. Firstly we have the "and" operation which uses the connective "." as in

$$x_1.z_2 \quad (2)$$

Which only returns true if  $x_1$  and  $x_2$  is true. Then there is the "or" operation represented as a "+"

$$x_1 + x_2 \quad (3)$$

Where it returns true if  $x_1$  is true or if  $x_2$  or both are true. Thirdly we have the negation operation indicated by a  $\bar{\phantom{x}}$  as in

$$\bar{x}_1 \quad (4)$$

This operation returns true if  $x_1$  is false and false if  $x_1$  is true. The  $\cdot$  and  $+$  operations are commutative

$$\begin{aligned} x_1 \cdot x_2 &= x_2 \cdot x_1 \\ x_1 + x_2 &= x_2 + x_1 \end{aligned} \quad (5)$$

and associative

$$\begin{aligned} x_1 \cdot x_2 (x_3) &= x_1 (x_2 \cdot x_3) \\ x_1 + x_2 + (x_3) &= x_1 + (x_2 + x_3) \end{aligned} \quad (6)$$

To commute between  $\cdot$  and  $+$  we use DeMorgan's laws

$$\begin{aligned} \overline{x_1 \cdot x_2} &= \bar{x}_1 + \bar{x}_2 \\ \overline{x_1 + x_2} &= \bar{x}_1 \cdot \bar{x}_2 \end{aligned} \quad (7)$$

Boolean functions can be broken down into various sub-classes. The first subclass is called *terms*. We can write these as  $k_1 k_2 k_3 \dots k_n$  where  $k_i$  are literals. An example would be the following term of size 4,  $x_2 \cdot x_3 \cdot \bar{x}_6 \cdot x_{12}$ . This is also called a conjunctive literal as in all the terms are separated by the and operation.

Secondly we have *clauses* or  $\cdot$ . A clause is a function where the literals are separated by the or function. As in  $k_1 + k_2 + \dots + k_n$ . An example would be  $x_1 + \bar{x}_5 + x_8$ .

Disjunctive normal functions are functions that can be written as a disjunction of terms.

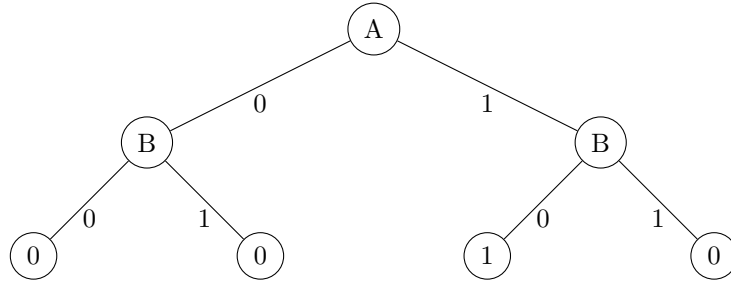
### 3 Question 3

### 4 Question 4

### 5 Question 5

#### 5.1 Question 5.1

$$A \wedge B$$



## 6 Question 6

Entropy is used to calculate a decision tree in the ID3 algorithm. For a collection of Samples ( $S$ ):

$$Entropy(S) = \sum_{i=1}^c -p_i \log_2 p_i \quad (8)$$

If we assume we only have 2 outcomes, then we let the number of positive outcomes be  $p$  and the number of negative outcomes be  $n$ . Then we can write Equation 8 as

$$Entropy(p, n) = -\frac{p}{p+n} \log_2 \frac{p}{p+n} - \frac{n}{p+n} \log_2 \frac{n}{p+n} \quad (9)$$

More specifically we compare the gain of different attributes of the sample set. Then create the tree based on the attributes with the greatest gain

$$Gain(S, A) \equiv Entropy(S) - \sum_{v \in Values(A)} \frac{|S_v|}{|S|} Entropy(S_v) \quad (10)$$

We need to determine  $Entropy(S)$

$$\begin{aligned} Entropy(p, n) &= -\frac{p}{p+n} \log_2 \frac{p}{p+n} - \frac{n}{p+n} \log_2 \frac{n}{p+n} \\ &= -\frac{3}{3+1} \log_2 \frac{3}{3+1} - \frac{1}{3+1} \log_2 \frac{1}{3+1} \\ &= \end{aligned} \quad (11)$$

Now we need to determine the gain for the 6 attributes to determine which one will be the root node.

For Sky we have sunny  $p_1 = 3, n_1 = 0$  and rainy  $p_1 = 0, n_1 = 1$

$$\begin{aligned} Gain(S, Sky) &= Entropy(S) - \sum_{v \in Values(A)} \frac{|S_v|}{|S|} Entropy(S_v) \\ &= ?? - \frac{1}{4} Entropy(S_{Sunny}) - Entropy(S_{Rainy}) \\ &= ?? - \frac{1}{4} Entropy(3, 0) - Entropy(0, 1) \\ &= ?? - \frac{3}{4} \left( -\frac{3}{3} \log_2 \frac{3}{3} - \frac{0}{3} \log_2 \frac{0}{3} \right) - \frac{1}{4} \left( -\frac{0}{1} \log_2 \frac{0}{1} - \frac{1}{1} \log_2 \frac{1}{1} \right) \\ &= \end{aligned} \quad (12)$$

## References