Assignment 1

Adriaan Louw (53031377)

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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

$$f(x_1, x_2, x_3, ...x_n) (1)$$

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

$$x_1.z_2 \tag{2}$$

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

$$x_1 + x_2 \tag{3}$$

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 \mathbf{a}^- as in

$$\bar{x_1}$$
 (4)

This operation returns true if x_1 is false and false if x_1 is true.

The \cdot and + operations are commutative

$$x_1.x_2 = x_2.x_1 x_1 + x_2 = x_2 + x_1$$
 (5)

and associative

$$x_1.x_2(x_3) = x_1(x_2.x_3)$$

$$x_1 + x_2 + (x_3) = x_1 + (x_2 + x_3)$$
(6)

To commute between . and + we use DeMorgan's laws

$$\frac{\overline{x_1.x_2} = \overline{x_1} + \overline{x_2}}{\overline{x_1 + x_2} = \overline{x_1.x_2}} \tag{7}$$

Boolean functions can be broken down into various sub-classes. The first subclass is called *terms*. We can write these as $k_1k_2k_3...k_n$ where k_i are literals. An example would be the following term of size 4, $x_2.x_3.\overline{x_6}.x_{12}$. This is also called a conjunctive literal as in all the terms are separated by the and operation.

Secondly we have *clauses* or . A clause is a function where the literals are separated by the or function. As in $k_1+k_2+\ldots+k_n$. An example would be $x_1+\overline{5}+x_8$.

- 3 Question 3
- 4 Question 4
- 5 Question 5
- 6 Question 6

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

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

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

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

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

We need to determine Entropy(S)

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

$$= -\frac{3}{3+1}log_2\frac{3}{3+1} - \frac{1}{3+1}log_2\frac{1}{3+1}$$

$$= (11)$$

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 $p_1=0, n_1=1$)

$$Gain(S, Sky) = Entropy(S) - \sum_{v \in Values(A)} \frac{|S_v|}{|S|} Entropy(S_v)$$

$$= ??? - \frac{1}{4} Entropy(S_{Sunny}) - Entropy(S_{Rainy})$$

$$= ??? - \frac{1}{4} Entropy(3, 0) - Entropy(0, 1)$$

$$= ??? - \frac{3}{4} (-\frac{3}{3}log_2\frac{3}{3} - \frac{0}{3}log_2\frac{0}{3}) - \frac{1}{4} (-\frac{0}{1}log_2\frac{0}{1} - \frac{1}{1}log_2\frac{1}{1})$$

$$=$$

$$=$$
(12)

References