Group 6: Timm Behner, Philipp Bruckschen, Patrick Kaster, Markus Schwalb MA-INF 4111 - Intelligent Learning and Analysis Systems: Machine Learning Exercise Sheet 4

1. The Perceptron Algorithm

 $\vec{w}_i = \vec{w}_k$ as stated in the task formulation and assume the training examples $(\vec{x}_{i+1}, y_{i+1}), \ldots, (\vec{x}_k, y_k)$ to be already learned. Without loss of generality let

$$y_{i+1}, \ldots, y_j \in \{+1\}$$

and

$$y_{i+1}, \ldots, y_k \in \{-1\}$$
.

Then:

$$\vec{w}_{k} = \vec{w}_{i} + \underbrace{\begin{array}{c} =0, \text{since } \vec{w}_{k} = \vec{w}_{i} \\ \hline \left((\vec{x}_{i+1}, y_{i+1}) + \ldots + (\vec{x}_{j}, y_{j}) \right) + \left((\vec{x}_{j+1}, y_{j+1}) + \ldots + (\vec{x}_{k}, y_{k}) \right)} \\ \Rightarrow \vec{x}_{i+1} + \ldots + \vec{x}_{j} = \vec{x}_{j+1} + \ldots + \vec{x}_{k} \end{array}}$$
(1)

Assume \vec{u} to be a solution to the learning problem, i.e. a weight vector for a perceptron that classifies correctly:

$$\langle \vec{u}, \vec{x}_{i+t} \rangle = \begin{cases} \geq 0 & \text{if } t \in \{i+1, \dots, j\} \\ < 0 & \text{if } t \in \{j+1, \dots, k\} \end{cases}$$

Implying

$$\langle \vec{u}, \vec{x}_{i+1} + \ldots + \vec{x}_j \rangle \ge 0$$

 $\langle \vec{u}, \vec{x}_{i+1} + \ldots + \vec{x}_k \rangle < 0$

in contradiction to (1). Thus the perceptron learning algorithm will loop over the same data infinitely and will never converge.

2. Backpropagation with the Hyperbolic Tangent Function

Since $(tanh(x))' = 1 - tanh(x)^2$, the weight update rules become:

$$\begin{split} & \delta_k \leftarrow \left(1 - o_k^2\right) (t_k - o_k) \\ & \delta_h \leftarrow \left(1 - o_h^2\right) \sum_{k \in \text{outputs}} w_{h,k} \delta_k \\ & w_{i,j} \leftarrow w_{i,j} + \Delta w_{i,j}, \text{ where } \delta w_{i,j} = \eta \delta_i x_{i,j} \end{split}$$

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3. Backpropagation for the XOR Function

Structure The network has a single hidden layer with three nodes (two regular and one for bias value), a single output node and three input nodes (again two for input and one for a bias value). Thus the structure is the same as seen on slide 27 in Slides from 2014-11-14.

Termination Criterion The termination criterion is on either 100000 iterations or if none of the weights changed. These values can be changed with the command line options -m and -t respectively.

Example output See the file output.log

4. Distances

For $\delta(x,y)$ to be a metric, we have to show:

Non-negative $\delta(x,y) \geq 0$

Identity of indiscernibles $\delta(x,y) = 0$ iff x = y

Symmetry $\delta(x,y) = \delta(y,x)$

triangle inequality $\delta(x,y) \leq \delta(x,z) + \delta(z,y)$

(i) **Non-negative** because $|\cdot|$ is a norm.

Identity of indiscernibles Cleary the Hamming distance sums to zero iff x_r, x_s don't differ in more than one entry:

$$D_{\text{Hamming}} = 0 \Rightarrow x_r = x_s$$

Symmetry

$$D_{\text{Hamming}}(x_r, x_s) = \sum_{j=1}^{m} |x_{rj} - x_{sj}|$$

$$= \sum_{j=1}^{m} |x_{sj} - x_{rj}| = D_{\text{Hamming}}(x_s, x_r)$$

triangle inequality

$$\forall x_r, x_s, x_t$$
:

$$D_{\text{Hamming}}(x_r, x_t) = \sum_{j=1}^{m} |x_{rj} - x_{tj}|$$

$$= \sum_{j=1}^{m} |x_{rj} - x_{sj} + x_{sj} - x_{tj}|$$

$$\leq \sum_{j=1}^{m} (|x_{rj} - x_{sj}| + |x_{sj} - x_{tj}|)$$

$$= \sum_{j=1}^{m} |x_{rj} - x_{sj}| + \sum_{j=1}^{m} |x_{sj} - x_{tj}|$$

$$= D_{\text{Hamming}}(x_r, x_s) + D_{\text{Hamming}}(x_s, x_t)$$

(ii) **Non-negative** because $|\cdot|$ is a norm.

Identity of indiscernibles

$$D_{\Delta}(S,S) = |S \setminus S \cup S \setminus S| \tag{2}$$

$$=|\emptyset|=0\tag{3}$$

and Let $D_{\Delta}(A,B)=0$ from that follows

$$0 = |A \setminus B \cup B \setminus A| \tag{4}$$

$$= |A \setminus B| + |A \setminus B| \qquad \text{because } (A \setminus B) \cap (B \setminus A) = \emptyset$$
 (5)

and because $|\cdot|$ is positive follows $|A\setminus B|\geq 0$ and $|B\setminus A|\geq 0$ which implies $|A\setminus B|=0$ and $|B\setminus A|=0$. From that follows $A\subset B$ and $B\subset A$ which implies A=B.

Symmetry $D_{\Delta}(A,B) = |A \setminus B \cup B \setminus A| = |B \setminus A \cup A \setminus B| = D_{\Delta}(B,A)$ because \cup is commutative.

triangle inequality Consider sets A, B and C

$$D_{\Delta}(A,B) + D_{\Delta}(B,C) = |A| + |B| - 2|A \cap B| + |B| + |C| - 2|B \cap C|$$
(6)
$$= |A| + |C| + 2(-(|A \cap B| + |B \cap C|) + |B|)$$
(7)
$$\geq |A| + |C| + 2(-(|B| + |A \cap B \cap C|) + |B|)$$
(8)

$$= |A| + |C| - 2|A \cap B \cap C| \tag{9}$$

$$\geq |A| + |C| - |A \cap C| \tag{10}$$

$$=D_{\Delta}(A,C)\tag{11}$$

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