

2 - Backpropagation with the Hyperbolic Tangent Function

Revise the backpropagation algorithm for two layers (see slide 34 of Lecture 2014-11-14) so that it operates on units using the hyperbolic tangent (squashing) function in place of the sigmoid function (see the remark on slide 28 of Lecture 2014-11-14). That is, the output o of a single unit is defined by

$$o = \tanh(\text{net}) = \tanh(\vec{w} \cdot \vec{x})$$

Give the weight update rule for output layer weights and hidden layer weights.

First we need to find a derivative of $o(x)$ function to use it in update procedure:

$$\begin{aligned} \frac{do}{dx} &= \frac{d}{dx} \left(\frac{e^{2x} - 1}{e^{2x} + 1} \right) = \frac{(e^{2x} + 1) \frac{d}{dx} (e^{2x} - 1) - (e^{2x} - 1) \frac{d}{dx} (e^{2x} + 1)}{(e^{2x} + 1)^2} = \\ &= \frac{2e^{2x}(e^{2x} + 1) - (e^{2x} - 1) * 2e^{2x}}{(e^{2x} + 1)^2} = \frac{2e^{2x}(e^{2x} + 1 - e^{2x} + 1)}{(e^{2x} + 1)^2} = \frac{4e^{2x}}{(e^{2x} + 1)^2} = \\ &= \left(\frac{2}{e^x + e^{-x}} \right)^2 = \text{sech}(x)^2. \end{aligned}$$

Having this derivative we can formulate update rules for output and hidden layer weights:

$$\begin{aligned} \text{a) } \delta_k &\leftarrow \frac{do}{dx} * (t_k - o_k) = \left(\frac{2}{e^{\text{net}} + e^{-\text{net}}} \right)^2 * (t_k - o_k) \\ \text{b) } \delta_h &\leftarrow \frac{do}{dx} * \sum_{k \in \text{outputs}} w_{h,k} \delta_k = \left(\frac{2}{e^{\text{net}} + e^{-\text{net}}} \right)^2 * \sum_{k \in \text{outputs}} w_{h,k} \delta_k \end{aligned}$$

Having corresponding δ value for each output and hidden weight we can formulate next weight update rule:

$$w_{i,j} \leftarrow w_{i,j} + n \delta_i x_{i,j}$$