

Task:

a) Show that the derivative for the hyperbolic tangent sigmoid:

$$a = \frac{e^n - e^{-n}}{e^n + e^{-n}}$$

is $1-a^2$

b) Write the backpropagation rule for the hyperbolic sigmoid transfer function

Solution:

a) The goal of this exercise is to show that there is a convenient way for computing the derivative of the hyperbolic tangent sigmoid similarly to the sigmoid function where the derivative was $a(1-a)$, $a=f(n)$.

$$f(n) = a = \frac{e^n - e^{-n}}{e^n + e^{-n}}$$

$$\begin{aligned} \frac{\partial f(n)}{\partial n} &= \frac{\partial}{\partial n} \left(\frac{e^n - e^{-n}}{e^n + e^{-n}} \right) = \frac{(e^n - e^{-n})(e^n + e^{-n}) - (e^n - e^{-n})'(e^n + e^{-n})}{(e^n + e^{-n})^2} = \\ &= \frac{(e^n - (-1)e^{-n})(e^n + e^{-n}) - (e^n - e^{-n})(e^n + (-1)e^{-n})}{(e^n + e^{-n})^2} = \frac{(e^n + e^{-n})^2 - (e^n - e^{-n})^2}{(e^n + e^{-n})^2} = \\ &= 1 - \frac{(e^n - e^{-n})^2}{(e^n + e^{-n})^2} = 1 - a^2 \end{aligned}$$

b) for output neurons:

$$\delta_i^\zeta = (d_i^\zeta - o_i^\zeta) \cdot f'(net_i^\zeta) = (d_i^\zeta - o_i^\zeta) \cdot (1 - o_i^{2\zeta})$$

$$\Delta w_{ji} = \eta \cdot \sum_{\zeta} \delta_i^\zeta \cdot o_j^\zeta = \eta \cdot \sum_{\zeta} (d_i^\zeta - o_i^\zeta) \cdot (1 - o_i^{2\zeta}) \cdot o_j^\zeta$$

for hidden neurons:

$$\delta_j^\zeta = f'(net_j^\zeta) \cdot \sum_i w_{ji} \cdot \delta_i^\zeta = (1 - o_j^{2\zeta}) \cdot \sum_i w_{ji} \cdot \delta_i^\zeta$$