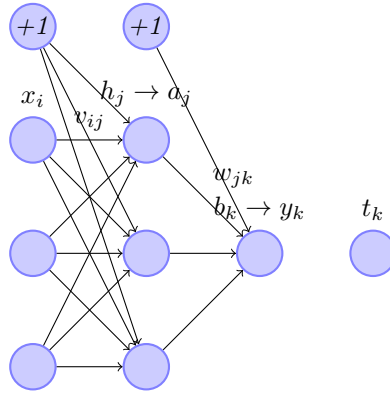


MLP

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1 Principles of Backpropagation



Suppose we have a simple binary classification problem that needs to be solved by MLP. The structure of the MLP is shown as above which has 3 input nodes for input layer, 3 hidden nodes for one hidden layer and 1 output node for output layer.

The activation function is sigmoid function: $g(x) = \frac{1}{1+e^{-x}}$ and $g'(x) = g(x)(1 - g(x))$.

Step 1: **Going forward**

$$h_j = \sum_i x_i \cdot v_{ij} \quad (1)$$

$$a_j = g(h_j) \quad (2)$$

$$b_k = \sum_j a_j \cdot w_{jk} \quad (3)$$

$$y_k = g(b_k) \quad (4)$$

Step 2: **Compute Error**

$$\begin{aligned}
E(\mathbf{v}, \mathbf{w}) &= \frac{1}{2} \sum_k (t_k - y_k)^2 \\
&= \frac{1}{2} \sum_k (t_k - g(b_k))^2 \\
&= \frac{1}{2} \sum_k (t_k - g(\sum_j a_j \cdot w_{jk}))^2 \\
&= \frac{1}{2} \sum_k (t_k - g(\sum_j g(h_j) \cdot w_{jk}))^2 \\
&= \frac{1}{2} \sum_k (t_k - g(\sum_j g(\sum_i x_i \cdot v_{ij}) \cdot w_{jk}))^2
\end{aligned} \tag{5}$$

Step 3: Compute the derivative of Error function respect to \mathbf{w} and \mathbf{v}

$$\begin{aligned}
\frac{\partial E}{\partial w_{jk}} &= \frac{\partial}{\partial w_{jk}} \left(\frac{1}{2} \sum_k (t_k - y_k)^2 \right) \\
&= \sum_k (t_k - y_k) \cdot \frac{\partial}{\partial w_{jk}} (t_k - y_k) \\
&= \sum_k (t_k - y_k) \cdot \frac{\partial}{\partial w_{jk}} (-g(b_k)) \\
&= \sum_k (t_k - y_k) \cdot -g'(b_k) \cdot \frac{\partial}{\partial w_{jk}} (\sum_j a_j \cdot w_{jk}) \\
&= \sum_k (y_k - t_k) \cdot y_k \cdot (1 - y_k) \cdot a_j
\end{aligned} \tag{6}$$

$$\begin{aligned}
\frac{\partial E}{\partial v_{ij}} &= \frac{\partial E}{\partial a_j} \cdot \frac{\partial a_j}{\partial v_{ij}} \\
&= \frac{\partial}{\partial a_j} \left(\frac{1}{2} \sum_k (t_k - g(\sum_j a_j \cdot w_{jk}))^2 \right) \cdot \frac{\partial}{\partial v_{ij}} (g(\sum_i x_i \cdot v_{ij})) \\
&= \left(\sum_k (y_k - t_k) \cdot y_k \cdot (1 - y_k) \cdot w_{jk} \right) \cdot a_j \cdot (1 - a_j) \cdot x_i
\end{aligned} \tag{7}$$

Step 4: Update weight \mathbf{w} and \mathbf{v}

$$w_{jk} = w_{jk} - \eta \frac{\partial E}{\partial w_{jk}} \tag{8}$$

$$v_{ij} = v_{ij} - \eta \frac{\partial E}{\partial v_{ij}} \tag{9}$$