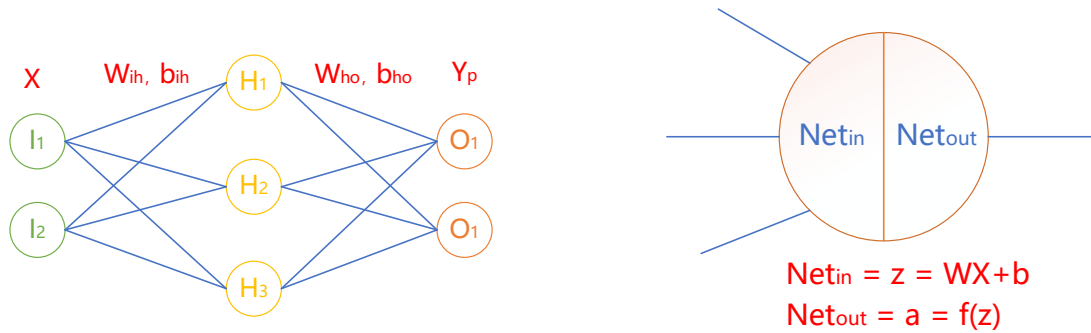


神经网络反向传播算法公式推导 (by: GiffordY)



学习率: α

$$\text{损失函数: } Loss = MSE = \sum_{i=1}^n \frac{1}{n} (y_{ti} - y_{pi})^2 = \sum_{i=1}^n \frac{1}{n} (y_{ti} - a_{oi})^2$$

$$\text{损失函数对 } a_o \text{ 求导: } \frac{\partial Loss}{\partial a_o} = -\frac{2}{n} [(y_{t1} - a_{o1}) + \dots + (y_{tm} - a_{on})] = -\frac{2}{n} (Y_t - a_o)$$

$$\text{激活函数: } f(z) = \text{sigmoid}(z) = \frac{1}{1 + e^{-z}}$$

$$\text{激活函数求导: } f'(z) = \frac{\partial a}{\partial z} = (-1) \times \frac{1}{(1 + e^{-z})^2} \times e^{-z} (-1) = f^2 \times \left(\frac{1}{f} - 1\right) = a(1 - a)$$

1、输出层→隐藏层:

$$(1) \text{ 损失函数对 } W_{ho} \text{ 的偏导: } \frac{\partial Loss}{\partial W_{ho}} = \frac{\partial Loss}{\partial a_o} \times \frac{\partial a_o}{\partial z_o} \times \frac{\partial z_o}{\partial W_{ho}}$$

$$\text{代入, 得: } \frac{\partial Loss}{\partial W_{ho}} = -\frac{2}{n} (Y_t - a_o) \times a_o (1 - a_o) \times a_h$$

$$\text{损失函数对 } b_{ho} \text{ 的偏导: } \frac{\partial Loss}{\partial b_{ho}} = \frac{\partial Loss}{\partial a_o} \times \frac{\partial a_o}{\partial z_o} \times \frac{\partial z_o}{\partial b_{ho}}$$

$$\text{代入, 得: } \frac{\partial Loss}{\partial b_{ho}} = -\frac{2}{n} (Y_t - a_o) \times a_o (1 - a_o)$$

$$(2) \text{ 更新权重 } W_{ho}: W_{ho} = W_{ho} - \alpha \times \frac{\partial Loss}{\partial W_{ho}}$$

$$b_{ho}: b_{ho} = b_{ho} - \alpha \times \frac{\partial Loss}{\partial b_{ho}}$$

2、隐藏层→输入层:

$$(1) \text{ 损失函数对 } W_{ih} \text{ 的偏导: } \frac{\partial Loss}{\partial W_{ih}} = \frac{\partial Loss}{\partial a_h} \times \frac{\partial a_h}{\partial z_h} \times \frac{\partial z_h}{\partial W_{ih}}$$

$$\text{其中: } \frac{\partial Loss}{\partial a_h} = \frac{\partial Loss}{\partial a_o} \times \frac{\partial a_o}{\partial z_o} \times \frac{\partial z_o}{\partial a_h} = -\frac{2}{n} (Y_t - a_o) \times a_o (1 - a_o) \bullet \times W_{ho}$$

$$\text{所以: } \frac{\partial Loss}{\partial W_{ih}} = -\frac{2}{n} (Y_t - a_o) \times a_o (1 - a_o) \bullet \times W_{ho} \times a_h (1 - a_h) \bullet \times X$$

$$\text{损失函数对 } b_{ih} \text{ 的偏导: } \frac{\partial Loss}{\partial b_{ih}} = \frac{\partial Loss}{\partial a_h} \times \frac{\partial a_h}{\partial z_h} \times \frac{\partial z_h}{\partial b_{ih}}$$

$$\text{代入, 得: } \frac{\partial Loss}{\partial b_{ih}} = -\frac{2}{n} (Y_t - a_o) \times a_o (1 - a_o) \bullet \times W_{ho} \times a_h (1 - a_h) = \frac{\partial Loss}{\partial a_h} \times a_h (1 - a_h)$$

$$(2) \quad \text{更新权重 } W_{ih}: W_{ih} = W_{ih} - \alpha \times \frac{\partial Loss}{\partial W_{ih}}$$

$$b_{ih}: b_{ih} = b_{ih} - \alpha \times \frac{\partial Loss}{\partial b_{ih}}$$

注：公式中“ $\cdot \times$ ”表示矩阵乘法，“ $\cdot \times$ ”后面的变量需要先变为它的转置矩阵，左乘还是右乘要先推一下矩阵的 shape。