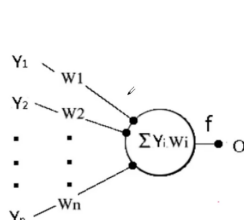


神经网络的实现

神经元基本计算公式



$$x_j = \sum_i Y_i W_{ji}$$

$$y_j = \frac{1}{1 + e^{-x_j}}$$

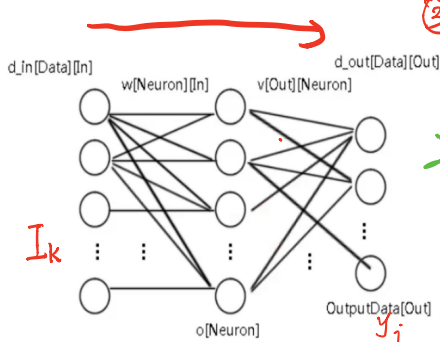
① $E = 1/2 \times \sum (y_j - d_j)^2$ 均方误差

(因为输出层的值与真实的值会存在误差, 我们可以用均方误差来衡量)

量预测值和真实值之间的误差)

神经网络的实现

输出层和隐藏层权值的调整方法 ② $y_j = \sum_i o_i v_{ji} \Rightarrow \frac{\partial y_j}{\partial v_{ji}} = o_i, \frac{\partial y_j}{\partial o_i} = v_{ji}$



③ $\frac{\partial E}{\partial y_j} = y_j - d$

$\therefore \frac{\partial E}{\partial v_{ji}} = \frac{\partial E}{\partial y_j} \frac{\partial y_j}{\partial v_{ji}} = (y_j - d) o_i$ ✓

2° $\frac{\partial E}{\partial o_i} = \sum_j \frac{\partial E}{\partial y_j} \frac{\partial y_j}{\partial o_i} = \sum_j (y_j - d) v_{ji}$

$x_i = \sum_k I_k w_{ki}$

$o_i = \frac{1}{1 + e^{-x_i}}$ (非线性函数)

1° $\frac{\partial E}{\partial w_{ki}} = \frac{\partial E}{\partial o_i} \frac{\partial o_i}{\partial w_{ki}}$

3° $\frac{\partial o_i}{\partial w_{ki}} = \frac{\partial o_i}{\partial x_i} \frac{\partial x_i}{\partial w_{ki}} = o_i (1 - o_i) I_k$

$\therefore \frac{\partial E}{\partial w_{ki}} = o_i (1 - o_i) I_k \sum_j (y_j - d) v_{ji}$ ✓

梯度下降法

(求偏导) (w, v)

调整权值

W, V: 权重由二维数组表示.