

图 5.7 BP 网络及算法中的变量符号

设对于(x,y),神经网络的输出为y',x为d维向量($d\times 1$),y和y'为l维向量($l\times 1$),神经网络的中间层为q维

从输入到输出的函数为,其中f(x)为sigmoid函数

$$egin{aligned} lpha_h &= \sum_{i=1}^d v_{ih} x_i \ b_h &= f(lpha_h - \gamma_h) \ eta_j &= \sum_{h=1}^q w_{hj} b_h \ y' &= f(eta_j - heta_j) \end{aligned}$$

定义误差函数为

$$J = \frac{1}{2}(y'-y)^2$$

设学习率为 α ,基于梯度下降法,神经网络权重的更新公式为

$$v = v - \alpha \frac{\partial J}{\partial v}$$
 $w = w - \alpha \frac{\partial J}{\partial w}$

$$\frac{\partial J}{\partial v} = \frac{\partial J}{\partial y'} \frac{\partial y'}{\partial \beta} \frac{\partial \beta}{\partial b_h} \frac{\partial b_h}{\partial \alpha_h} \frac{\partial \alpha_h}{\partial v_{ih}}$$

$$\frac{\partial J}{\partial y'} = y' - y$$

$$\frac{\partial y'}{\partial \beta} = f'(\beta_j - \theta_j) = f(\beta_j - \theta_j)(1 - f(\beta_j - \theta_j)) = y'(1 - y')$$

$$\frac{\partial \beta}{\partial b_h} = \sum_{j=1}^l w_{hj}$$

$$\frac{\partial b_h}{\partial \alpha_h} = f'(\alpha_h - \gamma_h) = f(\alpha_h - \gamma_h)(1 - f(\alpha_h - \gamma_h)) = b_h(1 - b_h)$$

$$\frac{\partial \alpha_h}{\partial v_{ih}} = x_i$$

所以

$$v = v - lpha((y'-y)y'(1-y')\sum_{j=1}^{l}(w_{hj})b_h(1-b_h)x_i)$$