$$\mathbf{w}_{1}^{(3)} = \mathbf{w}_{0}^{(3)} - \eta \vec{x}_{n}^{(2)} (\overrightarrow{\delta}^{(3)})^{T} = \mathbf{w}_{0}^{(3)} - \eta \begin{pmatrix} 1 \\ x_{1}^{(2)} \\ x_{2}^{(2)} \\ x_{3}^{(2)} \end{pmatrix} \delta_{1}^{(3)} = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} - 0.01 \begin{pmatrix} 42 \\ 7 * 42 \\ 7 * 42 \\ 7 * 42 \end{pmatrix}$$

$$= \begin{pmatrix} 0.58 \\ -1.94 \\ -1.94 \\ -1.94 \end{pmatrix}$$

t=2,对于第二个样本 $\vec{x}_2=(-1,-1)^T$ ,则第一层神经元的输入为:

$$\begin{pmatrix} s_1^{(1)} \\ s_2^{(1)} \end{pmatrix} = (\mathbf{w}^{(1)})^T \vec{\mathbf{x}}_n^{(0)} = \begin{pmatrix} -0.26 & -0.26 & -0.26 \\ -0.26 & -0.26 & -0.26 \end{pmatrix} \begin{pmatrix} 1 \\ -1 \\ -1 \end{pmatrix} = \begin{pmatrix} 0.26 \\ 0.26 \end{pmatrix}$$

$$\text{III: } \begin{pmatrix} x_1^{(1)} \\ x_2^{(1)} \end{pmatrix} = \begin{pmatrix} \max(0, s_1^{(1)}) \\ \max(0, s_2^{(1)}) \end{pmatrix} = \begin{pmatrix} 0.26 \\ 0.26 \end{pmatrix}$$

第二层神经元的输入为:

$$\begin{pmatrix} s_1^{(2)} \\ s_2^{(2)} \\ s_2^{(2)} \end{pmatrix} = (\mathbf{w}^{(2)})^T \begin{pmatrix} 1 \\ \chi_1^{(1)} \\ \chi_2^{(1)} \end{pmatrix} = \begin{pmatrix} 0.58 & -0.26 & -0.26 \\ 0.58 & -0.26 & -0.26 \\ 0.58 & -0.26 & -0.26 \end{pmatrix} \begin{pmatrix} 1 \\ 0.26 \\ 0.26 \end{pmatrix} = \begin{pmatrix} 0.44 \\ 0.44 \\ 0.44 \end{pmatrix}$$

则: 
$$\begin{pmatrix} x_1^{(2)} \\ x_2^{(2)} \\ x_3^{(2)} \end{pmatrix} = \begin{pmatrix} 0.44 \\ 0.44 \\ 0.44 \end{pmatrix}$$

则第三层的输入为:

$$\mathbf{s}_{1}^{(3)} = (\mathbf{w}^{(3)})^{T} \begin{pmatrix} 1 \\ \chi_{1}^{(2)} \\ \chi_{2}^{(2)} \\ \chi_{3}^{(2)} \end{pmatrix} = (0.58 -1.94 -1.94 -1.94) \begin{pmatrix} 1 \\ 0.44 \\ 0.44 \\ 0.44 \end{pmatrix} = -1.98$$

即输出 $\hat{y} = s_1^{(3)} = -1.98$ 

对于样本 $\vec{x}_2$ ,其标签为1,采用平方误差函数: $e_n=(y_n-\hat{y}_n)^2$ ,则:

$$\delta_1^{(3)} = -2(y_n - s_1^{(3)}) = -2(1 - (-1.98)) = -5.96$$

运用反向传播法,于是: