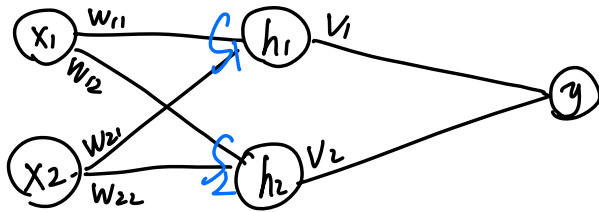


23-51-Q4

Q(a) MLP diagram \rightarrow ref topic 10 P71

input hidden output



(b) ① the net input to each hidden neuron

$$s_1 = w_{11}x_1 + w_{21}x_2 + b_1$$

$$s_2 = w_{12}x_1 + w_{22}x_2 + b_2$$

② σ activation function

$$h_1 = \sigma(s_1) = \frac{1}{1 + e^{-s_1}}$$

$$h_2 = \sigma(s_2) = \frac{1}{1 + e^{-s_2}}$$

③ output

$$y = v_1 h_1 + v_2 h_2$$

$$= \frac{v_1}{1 + e^{-(w_{11}x_1 + w_{21}x_2 + b_1)}} + \frac{v_2}{1 + e^{-(w_{12}x_1 + w_{22}x_2 + b_2)}}$$

(c) & design learning rule \rightarrow adjust $V = [v_1 \ v_2]^T$

Solution

① define the error function as Mean Square Error

$$E = \frac{1}{2} (t - y)^2$$

② compute gradient

$$y = v_1 h_1 + v_2 h_2$$

$$\frac{\partial y}{\partial v_j} = h_j$$

$$\frac{\partial E}{\partial v_j} = -(t - y) \frac{\partial y}{\partial v_j}$$

$$= -(t - y) h_j$$

③ Gradient Descent, η is the learning rate

$$\begin{aligned} v_j^{\text{new}} &= v_j^{\text{old}} - \eta \frac{\partial E}{\partial v_j} \\ &= v_j^{\text{old}} + \eta (t - y) h_j \end{aligned}$$

denote $e = t - y$

$$\Delta v_j = \eta e h_j$$

$$\text{So } V_j^{\text{new}} = V_j^{\text{old}} + \Delta V_j$$