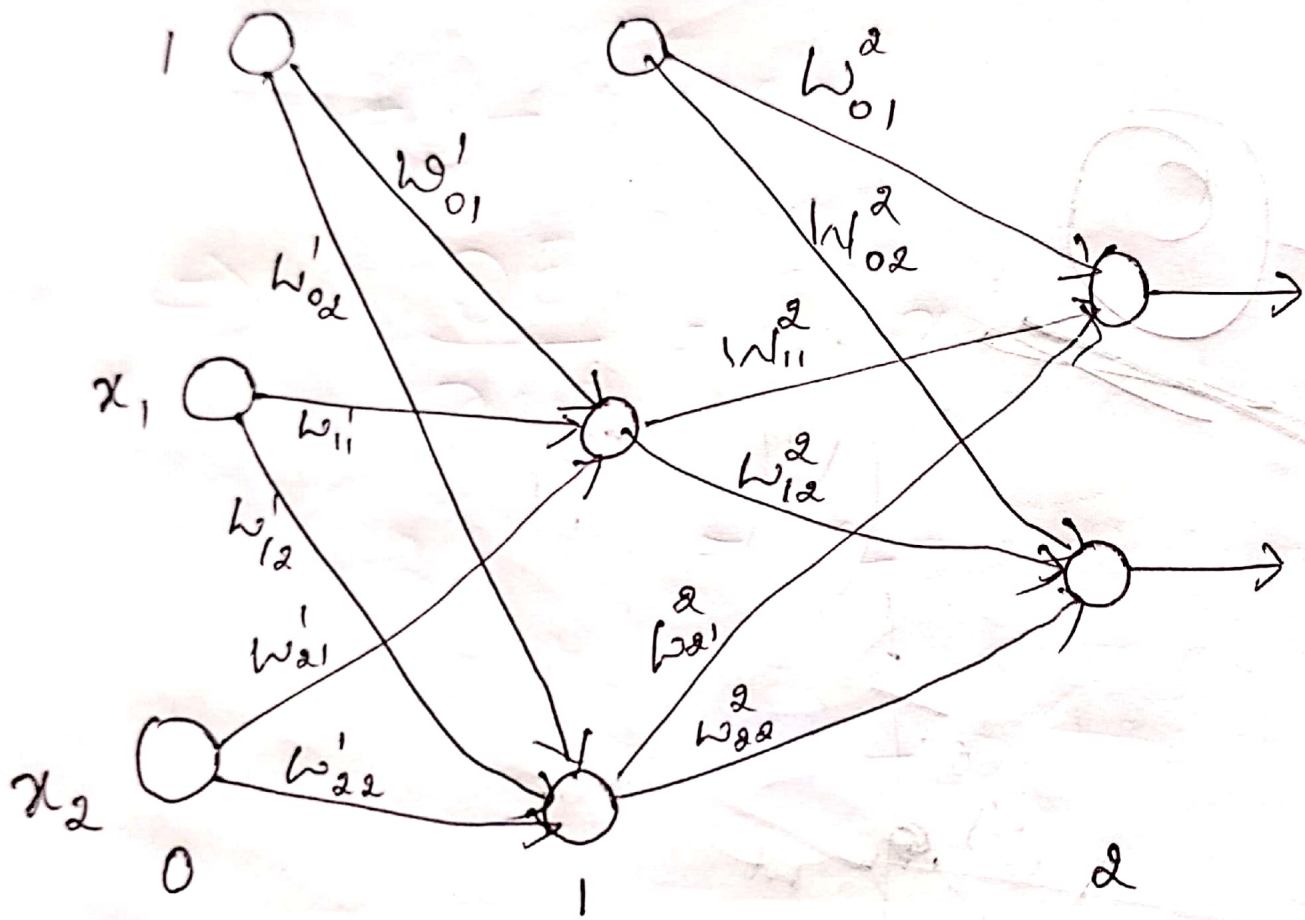


Backpropagation Learning Example-II

Feed Forward Pass



$$L^1 \quad x_i^0 \quad \theta_j^1 = \sum w_{ij}^1 x_i^0 \quad x_j^1 = \frac{1}{1 + e^{-\theta_j^1}}$$

$$\begin{bmatrix} 0.5 & 1.5 & 0.8 \\ 0.8 & 0.2 & -1.6 \end{bmatrix} \begin{bmatrix} 1 \\ 0.7 \\ 1.2 \end{bmatrix} = \begin{bmatrix} 2.31 \\ -9.8 \end{bmatrix} \Rightarrow \begin{bmatrix} 0.92 \\ 0.27 \end{bmatrix}$$

$$L^2 \quad x_i^1 \quad \theta_j^2 = \sum w_{ij}^2 x_i^1 \quad x_j^2 = \frac{1}{1 + e^{-\theta_j^2}}$$

$$\begin{bmatrix} 0.9 & -1.7 & 1.6 \\ 1.2 & 2.1 & -0.9 \end{bmatrix} \begin{bmatrix} 1 \\ 0.92 \\ -0.27 \end{bmatrix} = \begin{bmatrix} -0.232 \\ 3.057 \end{bmatrix} \Rightarrow \begin{bmatrix} 0.44 \\ 0.95 \end{bmatrix}$$

$$t = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

Back propagation Learning: Output layer

$$E = \frac{1}{2} \sum_{j=1}^2 (x_j^2 - t_j)^2 \quad x_j^2 = \frac{1}{1 + e^{-\theta_j^2}} \quad \theta_j^2 = \sum_{i=0}^2 w_{ij}^2 x_i^1$$

$$\frac{\partial E}{\partial w_{ij}^2} = \frac{\partial E}{\partial x_j^2} \cdot \frac{\partial x_j^2}{\partial \theta_j^2} \cdot \frac{\partial \theta_j^2}{\partial w_{ij}^2} = (x_j^2 - t_j) x_j^2 (1 - x_j^2) x_i^1$$

Let $\delta_j^2 = x_j^2 (1 - x_j^2) (x_j^2 - t_j)$

$$\Rightarrow \frac{\partial E}{\partial w_{ij}^2} = \delta_j^2 x_i^1$$

$$\delta_j^2 = x_j^2 (1 - x_j^2) (x_j^2 - t_j)$$

$$\delta_1^2 = x_1^2 (1 - x_1^2) (x_1^2 - t_1)$$

$$= 0.44 (1 - 0.44) * (0.44 - 1)$$

$$= -0.138$$

$$\Rightarrow \frac{\partial E}{\partial W_{11}^2} = \delta_1^2 x_1^1$$

$$= (-0.138) (0.92)$$

$$= -0.126$$

$$W_{11}^2 \leftarrow W_{11}^2 - \eta (-0.126)$$

↳ increased

$$\begin{aligned}\delta_2^2 &= x_2^2 (1 - x_2^2) (x_2^2 - t_2) \\ &= 0.95 (1 - 0.95) * (0.95 - 0) \\ &= \underline{0.045} \checkmark\end{aligned}$$

$$\Rightarrow \frac{\partial E}{\partial w_{12}^2} = \delta_2^2 x_1^1 = 0.04$$

$$w_{12}^2 \leftarrow w_{12}^2 - \eta * 0.04$$

↓
Decreased

111y

$$\frac{\partial E}{\partial w_{21}^1} = \delta_1^2 x_2^1 = 0.037$$

$$\frac{\partial E}{\partial w_{22}^1} = \delta_2^2 x_2^1 = \underline{0.012}$$

$$\frac{\partial E}{\partial w_{01}^2} = \delta_1^2 x_0^1 = -1.38$$

$$\frac{\partial E}{\partial w_{02}^2} = \delta_2^2 x_0^1 = 0.045$$

Hidden Layer

Let set \rightarrow Backpropagated Error

$$\delta_i^{k+1} = x_i^{k+1} (1 - x_i^{k+1}) \sum_{j=1}^{M_{k+1}} \delta_j^{k+1} w_{ij}^{k+1}$$

\Downarrow

$$\delta_i^1 = x_i^1 (1 - x_i^1) \sum_{j=1}^2 \delta_j^2 w_{ij}^2$$

$$\frac{\partial E}{\partial w_{ij}^{k+1}} = \delta_i^{k+1} x_j^{k+1}$$

$$\delta_1' = x_1' (1 - x_1') [\delta_1^2 * w_{11}^2 + \delta_2^2 w_{12}^2]$$

$$= 0.92 * (1 - 0.92) [(-0.137 * (-1.7) + 0.045 * (2.1))]$$

$$= 0.024$$

$$\delta_2' = x_2' (1 - x_2') [\delta_1^2 * w_{21}^2 + \delta_2^2 w_{22}^2]$$

$$= 0.27 * (1 - 0.27) [(-0.137) * 1.6 + (0.045) * (-0.2)]$$

$$= -0.02$$

$$\frac{\partial E}{\partial w_{11}} = \delta_1^1 * x_1^0$$

$$= 0.024 * 0.7$$

$$= 0.017$$

$$\frac{\partial E}{\partial w_{12}} = \delta_1^1 * x_2^0$$

$$= -0.02 * 0.7$$

$$= -0.014$$

$$\frac{\partial E}{\partial w_{21}} = \delta_2^1 * x_1^0$$

$$= (0.024) * 1.2$$

$$= 0.0288$$

$$\begin{aligned}\frac{\partial E}{\partial w_{22}^1} &= \delta_2^1 * x_2^0 \\ &= (-0.02) * (1.2) \\ &= -0.024\end{aligned}$$

$$\begin{aligned}\frac{\partial E}{\partial w_{01}^1} &= \delta_1^1 * x_0^0 \\ &= 0.024 * 1 \\ &= 0.024\end{aligned}$$

$$\begin{aligned}\frac{\partial E}{\partial w_{22}^1} &= \delta_2^1 * x_0^0 \\ &= -0.02 * 1 \\ &= -0.02\end{aligned}$$

$$w_{ij}' \leftarrow w_{ij}' - \eta \frac{\partial E}{\partial w_{ij}'}$$