

Question 2

① Given that,

$$V_j = \sum_i W_{ji} y_i$$

$$e_j = d_j - y_j$$

$$\text{error, } E = \frac{1}{2} \sum_j e_j^2$$

output from m layer neuron, $y_m = \phi(V_m)$

Here,

$$\begin{aligned} \delta_m &= \frac{\partial E}{\partial V_m} = - \frac{\partial E}{\partial y_m} \frac{\partial y_m}{\partial V_m} \\ &= - \frac{\partial E}{\partial y_p} \phi'(V_m) \end{aligned}$$

And,

$$E(n) = \frac{1}{2} \sum_{p \in C} e_p^2(n)$$

$$\Rightarrow - \frac{\partial E}{\partial y_m} = \sum_{p \in C} e_p \frac{\partial e_p}{\partial y_m}$$

$$= - \sum_{p \in C} e_p (1 - y_p) \phi(V_p) W_{pm}$$

$$= \sum_{p \in C} \delta_p W_{pm}$$

$$\frac{\partial e_p}{\partial y_m} = \frac{\partial e_p}{\partial y_p} \frac{\partial y_p}{\partial V_p} \frac{\partial V_p}{\partial y_m}$$

$$= (-1) \phi'(V_p) W_{pm}$$

$$\therefore \delta_m = \phi'(V_m) \sum_{p \in e} \delta_p W_{pm}$$

finally,

$$\delta_m = \begin{cases} e_p \phi'(V_m) & [\text{When } m \text{ is output neuron}] \\ \phi'(V_m) \sum_{p \in e} \delta_p W_{pm} & [\text{When } m \text{ is hidden neuron}] \end{cases}$$

So, this is our computed local induced gradient, δ_m

(Ans)