

Connected layer from $i \rightarrow j$

$$\frac{\partial E}{\partial w_{jk}} = \underbrace{\frac{\partial E}{\partial o_k}}_{(1)} \underbrace{\frac{\partial o_k}{\partial s_k}}_{(2)} \underbrace{\frac{\partial s_k}{\partial w_{jk}}}_{(3)} \quad (9)$$

$$(1) \frac{\partial E}{\partial o_k} = -(y_k - o_k) \quad [5] \quad (10)$$

$$(2) \frac{\partial o_k}{\partial s_k} = o_k (1 - o_k) = 1 \quad (11)$$

The output is a sum,
i.e. there is no threshold

$$(3) s_k = \sum w_{jk} \times o_j = \frac{\partial s_k}{\partial w_{jk}} = \frac{\partial (w_{jk} \times o_j)}{\partial w_{jk}} = o_j \quad (12)$$

In conclusion:
$$\frac{\partial E}{\partial w_{jk}} = \underbrace{-(y_k - o_k)}_{\text{error}} \times \underbrace{1 \times o_j}_{\text{inputs}} \quad (13)$$

Jacobian

$$\text{As } g = \begin{bmatrix} \frac{\partial E}{\partial w_1} \\ \vdots \\ \frac{\partial E}{\partial w_m} \end{bmatrix} \text{ then } g_{ik} = \text{error} \begin{bmatrix} o_{1k} \\ o_{2k} \\ o_{3k} \end{bmatrix} \quad (14)$$

Connected layer from $i \rightarrow j$

$$\frac{\partial E}{\partial w_{ij}} = \underbrace{\frac{\partial E}{\partial s_j}}_{(1)} o_i \quad (15)$$

$$(1) \frac{\partial E}{\partial s_j} = - \left[o_j (1 - o_j) \underbrace{\sum_k (-\beta_k) w_{jk}}_{(2)} \right] \quad [5] (16)$$

$$(2) \beta_k = \underbrace{o_k (1 - o_k)}_{= 1 \text{ as centred}} [y_k - o_k] = 1 \times (y_k - o_k) \quad (17)$$

In conclusion: $\frac{\partial E}{\partial w_{ij}} = - \left[o_j (1 - o_j) \sum_k [-(y_k - o_k)] w_{jk} \right] o_i \quad (18)$

$$\frac{\partial E}{\partial w_{ij}} = \left[o_j (1 - o_j) \sum_k \underbrace{(y_k - o_k)}_{\text{Error}} \underbrace{w_{jk}}_{\text{weights}} \underbrace{o_i}_{\text{inputs}} \right] \quad (19)$$

Consequently: $g_{ij} = \left[\begin{array}{c} \frac{\partial E}{\partial w_{ij}} \\ \vdots \\ \frac{\partial E}{\partial w_{mj}} \end{array} \right] \quad (20)$