



lets assume our target to 'y'.

Here x_0 is a matrix (1×3) .

w_1 is a matrix (3×4) .

so $x_1 = f(w_0 x_0)$.

$\Rightarrow x_1$ is $(1 \times 3) \cdot (3 \times 4) \Rightarrow$ ~~(1×4)~~

$\Rightarrow x_i$ is (1×4) matrix.

lets say our activation function is $f(x)$.

Then;

$$x_1 = f(w_1 x_0) + b_0 \rightarrow (1)$$

$$x_2 = f(w_2 x_1) + b_1 \rightarrow (2)$$

$$x_3 = f(w_3 x_2) + b_2 \rightarrow (3)$$

I ignore b_0, b_1, b_2 because they are constants and will be added after back propagation step.

Let's assume the error function to be: $\frac{1}{2} (y - x_3)^2$

$$E = \frac{1}{2} (y - x_3)^2$$

$$\text{or } E = \frac{1}{2} (x_3 - y)^2$$

Now, let's see how this error was affected by w_3 . For that we need the derivative with respect to w_3 .

$$\frac{\partial E}{\partial w_3} = \frac{\partial}{\partial w_3} \left(\frac{1}{2} (x_3 - y)^2 \right)$$

$$= (x_3 - y) \cdot \frac{\partial x_3}{\partial w_3}$$

$$= (x_3 - y) \cdot \frac{\partial (f(w_3 x_2))}{\partial w_3}$$

$$= [(\kappa_3 - y) \cdot f'(\omega_3 \kappa_2)] \cdot \frac{\partial(\omega_3 \kappa_2)}{\partial \omega_3}$$

$$= ((\kappa_3 - y) \cdot f'(\omega_3 \kappa_2)) \cdot \kappa_2$$

$$\boxed{\frac{\partial E}{\partial \omega_3} = \cdot \kappa_3 \cdot \kappa_2}$$

→ (4)

$$\text{or } \boxed{\kappa_3 = (\kappa_3 - y) \cdot f'(\omega_3 \kappa_2)}$$

→ (5)

Now lets do the backward pass for ω_2 .

$$\frac{\partial E}{\partial \omega_2} = (\kappa_3 - y) \cdot \frac{\partial \kappa_3}{\partial \omega_2}$$

$$= (\kappa_3 - y) \cdot \frac{\partial (f(\omega_3 \kappa_2))}{\partial \omega_2} = ((\kappa_3 - y) \cdot f'(\omega_3 \kappa_2)) \cdot \frac{\partial(\omega_3 \kappa_2)}{\partial \omega_2}$$

$$= ((\kappa_3 - y) \cdot f'(\omega_3 \kappa_2)) \cdot \omega_3 \cdot \frac{\partial \kappa_2}{\partial \omega_2}$$

From eqn (5):

$$= \cdot \kappa_3 \cdot \omega_3 \cdot \frac{\partial \kappa_2}{\partial \omega_2}$$

$$= \kappa_3 \omega_3 \cdot \frac{\partial f(\omega_2 \kappa_1)}{\partial \omega_2} = \kappa_3 \omega_3 \cdot f'(\omega_2 \kappa_1) \cdot \frac{\partial(\omega_2 \kappa_1)}{\partial \omega_2}$$

$$\Rightarrow \boxed{\frac{\partial E}{\partial w_2} = (k_3 \cdot w_3 \cdot f'(w_2 x_1)) \cdot x_1}$$

$$\boxed{\frac{\partial E}{\partial w_2} = k_2 \cdot x_1}$$

→ (6)

$$k_2 = (k_3 w_3 \cdot f'(w_2 x_1))$$

→ (7)

similarly for backpass $w.r.t$ w_1 :

$$\frac{\partial E}{\partial w_1} = (x_3 - y) \cdot \frac{\partial x_3}{\partial w_1}$$

$$\frac{\partial E}{\partial w_1} = (x_3 - y) \cdot \frac{\partial (f(w_3 x_2))}{\partial w_1} = (x_3 - y) \cdot f'(w_3 x_2) \cdot \frac{\partial (w_3 x_2)}{\partial w_1}$$

from (5):

$$\frac{\partial E}{\partial w_1} = k_3 \cdot w_3 \cdot \frac{\partial x_2}{\partial w_1}$$

$$= k_3 \cdot w_3 \cdot \frac{\partial (f(w_2 x_1))}{\partial w_1}$$

$$= (k_3 \omega_3 \cdot f'(\omega_2 x_1)) \cdot \frac{\partial (\omega_2 x_1)}{\partial \omega_1}.$$

from eqn (7)

$$= k_2 \cdot \omega_2 \cdot \frac{\partial x_1}{\partial \omega_1}.$$

$$= k_2 \cdot \omega_2 \cdot \frac{\partial f(\omega_1 x_0)}{\partial \omega_1}.$$

$$= k_2 \omega_2 \cdot f'(\omega_1 x_0) \cdot \frac{\partial (\omega_1 x_0)}{\partial \omega_1}.$$

$$\boxed{\frac{\partial E}{\partial \omega_1} = (k_2 \cdot \omega_2 \cdot f'(\omega_1 x_0)) \cdot x_0}$$

$$\Rightarrow \boxed{\frac{\partial E}{\partial \omega_1} = k_1 \cdot x_0} \quad \hookrightarrow (8)$$

$$\&\& \boxed{k_1 = k_2 \omega_2 \cdot f'(\omega_1 x_0)} \quad \hookrightarrow (9)$$

from eqns : (8), (6), (4) we see:

$$\frac{\partial E}{\partial \omega_3} = k_3 \cdot x_2 \quad ; \quad \frac{\partial E}{\partial \omega_2} = k_2 x_1 \quad ; \quad \frac{\partial E}{\partial \omega_1} = k_1 x_0.$$

from eqns : (9), (7), (5) we see :

$$k_3 = (x_3 - y) \cdot f'(w_3 x_2)$$

$$k_2 = k_3 w_3 \cdot f'(w_2 x_1) \dots$$

$$k_1 = k_2 w_2 \cdot f'(w_1 x_0)$$

So the trend is clear :

for the layer we calculate the " k_n "

$$k_n = (x_n - y) \cdot f'(w_n x_{n-1}) \rightarrow (10)$$

for any other layer " i " we calculate " k_i "

$$k_i = w_{i+1} \cdot k_{i+1} \cdot f'(w_i x_{i-1}) \rightarrow (11)$$

$$\therefore \frac{\partial E}{\partial w_i} = k_i \cdot x_{i-1} \rightarrow (12) \text{ (i where } (0 \leq i < n)).$$

CLARIFICATIONS:

① How do you use this?

→ In a function like gradient descent.

$$\frac{\partial E}{\partial w_i} = k_i x_i - 1$$

$$w_i = w_i - \alpha w_i \cdot \frac{\partial E}{\partial w_i}$$

$\alpha \rightarrow$ learning rate.

② How do I find derivative of matrix while calculating k_i ?

(Ans) you don't; suppose $x_0 = \begin{bmatrix} 1 & 0 & 1 \end{bmatrix}_{1 \times 3}$

$$w_1 = \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 0 \end{bmatrix}_{3 \times 2}$$

$$x_0 \cdot w_1 = \begin{bmatrix} 2 & 0 \end{bmatrix}$$

~~lets say $f(x) = \frac{1}{1+e^{-x}}$~~

lets say $f(x) = \frac{1}{1+e^{-x}}$ (sigmoid function) ..

$$f'(x) = f(x)(1-f(x)) \quad (\because \text{google})$$

now;

$$x_1 = f(w, x_0) = \left[\frac{1}{1+e^{-2}} \quad \frac{1}{1+e^0} \right] = \left[\frac{e^2}{1+e^2} \quad \frac{1}{2} \right]$$

now;

for $f'(w, x_0)$ we should expand $f'(x)$.

$$f'(x) = \frac{1}{(1+e^{-x})} \cdot \left(1 - \frac{1}{1+e^{-x}} \right)$$

$$\begin{aligned} f'(w, x_0) &= \left[\frac{1}{1+e^{-2}} \left(1 - \frac{1}{1+e^{-2}} \right) \quad \frac{1}{1+e^0} \left(1 - \frac{1}{1+e^0} \right) \right] \\ &= \left[\frac{1}{1+e^{-2}} \left(\frac{1+e^{-2}-1}{1+e^{-2}} \right) \quad \frac{1}{2} \times \frac{1}{2} \right] \end{aligned}$$

$$f'(w, x_0) = \left[\frac{e^{-2}}{(1+e^{-2})^2} \quad \frac{1}{4} \right]$$

(Here you go for computers its very easy).