

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

$$\sigma'(z) = \frac{(-1)}{(1+e^{-z})^2} \cdot e^{-z} (-1)$$

$$= \frac{+e^{-z}}{(1+e^{-z})^2}$$

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

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

$$\sigma'(z) = \cancel{\frac{1}{1+e^{-z}}} \cdot \frac{e^{-z}}{1+e^{-z}}$$

$$\cancel{\sigma(z)} = \cancel{\sigma(z)}(\cancel{1-\sigma(z)})$$

$$\sigma'(z) = \sigma(z)(1-\sigma(z))$$

Hence proved

(2)

Correct cross entropy.

$$C = -[y \ln a + (1-y) \ln(1-a)]$$

 $y \rightarrow$ true or false label $a \rightarrow$ predicted activation

Correct cross entropy working

if $y=0$

$$C = -\ln(1-a) \Rightarrow \text{This means only as long as}$$

$\begin{aligned} & a < 1, \text{ increase} \\ & \cancel{a \rightarrow 1} \rightarrow \text{a large } \rightarrow \text{penalty} \end{aligned}$

$$C = -\ln a \Rightarrow \text{This means only as long as}$$

$\begin{aligned} & a > 1, \cancel{\text{small}} \\ & a \text{ small } \rightarrow \text{penalty} \end{aligned}$

wrong form

$$C = -[a \ln y + (1-a) \ln(1-y)]$$

if $y=1$

$$\ln(1-y) \rightarrow -\infty \quad C \rightarrow +\infty$$

$$\text{if } y=1 \quad \ln(y) \rightarrow -\infty \quad C \rightarrow +\infty$$

and $\ln(0)$ is undefined

so it doesn't make sense

(3)

$$C = -\frac{1}{n} \sum_{i=1}^n [y_i \ln a + (1-y_i) \ln(1-a)]$$

$$C = -\frac{1}{n} \sum_{i=1}^n [y_i \ln a + (1-y_i) \ln(1-a)]$$

↳ for one neuron

We have to find when cost w.r.t to predicted output is minimum.

$$\text{in the same } \frac{\partial C}{\partial a} = 0$$

$$0 = \frac{\partial}{\partial a} - [y_i \ln a + (1-y_i) \ln(1-a)]$$

$$= - \left[\frac{y_i}{a} + (-1) \frac{(1-y_i)}{1-a} \right] = - \left[\frac{y_i}{a} - \frac{1-y_i}{1-a} \right]$$

$$\frac{y_i}{a} = \frac{1-y_i}{1-a} \quad y_i - y_i/a = a - y_i a \\ \boxed{y_i = a}$$

$$\text{at } \frac{\partial C}{\partial a} = 0 \quad \boxed{y_i = a}$$

Hence cost is minimized when ~~predicted~~ \hat{a}
predicted equals actual

also $y = a$ mean

$$C = -\frac{1}{n} \sum_{i=1}^n [y_i \ln y + (1-y_i) \ln(1-y_i)]$$

↓

This also gives its minimum when the ~~at~~ $a = y$. and its maximum when a is far from y .

whole point is to find ~~at~~ Max and Min