

Machine Learning HW2 handwriting

1) $y = \sigma(x) = \frac{1}{1+e^{-x}} \Rightarrow \sigma(x) = (1+e^{-x})^{-1}$

a) prove $\frac{d\sigma(x)}{dx} \Rightarrow \sigma(x)(1-\sigma(x))$

"Chain Rule" $\frac{d\sigma(x)}{dx} \Rightarrow -(1+e^{-x})^{-2} * (-e^{-x}) \Rightarrow \frac{e^{-x}}{(1+e^{-x})^2} \Rightarrow$

Can be written as $\sigma(x) = \frac{1}{1+e^{-x}} \Rightarrow e^{-x} = \frac{1-\sigma(x)}{\sigma(x)}$ Substitute

$\frac{d\sigma(x)}{dx} \Rightarrow \frac{1-\sigma(x)}{\sigma(x)} \Rightarrow \frac{1-\sigma(x)}{\sigma(x)} * \sigma(x)^2 \Rightarrow (1-\sigma(x)) \sigma(x)$ ▲

b) prove $\sigma(-x) \Rightarrow 1-\sigma(x)$

$\sigma(-x) \Rightarrow \frac{1}{1+e^x} \xrightarrow{1+e^x} \frac{e^{-x}}{e^{-x}(1+e^x)} \Rightarrow \frac{e^{-x}}{1+e^{-x}} \Rightarrow 1 - \frac{1}{1+e^{-x}} \Rightarrow$
 $\underbrace{\frac{1}{1+e^{-x}}}_{\sigma(x)}$

" $1+e^x - 1 \Rightarrow e^x$ "

$\Rightarrow 1-\sigma(x)$ ▲

c) prove $x = \sigma^{-1}(y) = \ln\left(\frac{y}{1-y}\right)$

$y = \frac{1}{1+e^{-x}} \Rightarrow \frac{1}{y} \Rightarrow 1+e^{-x} \Rightarrow$

$e^{-x} \Rightarrow \frac{1}{y} - 1 \xrightarrow{1+y} \frac{1-y}{y} \Rightarrow$

$e^x \Rightarrow \frac{y}{1-y}$

$\ln(e^x) \Rightarrow x$

$\ln(e^x) \Rightarrow \ln\left(\frac{y}{1-y}\right) \Rightarrow x = \ln\left(\frac{y}{1-y}\right)$ ▲

$$2) \begin{cases} \hat{y}_n = \delta(W^T \phi_n) \\ E(w) = -\ln p(t|w) \Rightarrow -\sum_{n=1}^N \{t_n \ln y_n + (1-t_n) \ln(1-y_n)\} \\ \nabla E(w) = \sum_{n=1}^N (y_n - t_n) \phi_n \end{cases}$$

Derivative \Rightarrow

$$\frac{d}{dw} (t_n \ln y_n + (1-t_n) \ln(1-y_n))$$

(I) (II)

$$(I) \Rightarrow \frac{d}{dw} (t_n \ln y_n) \Rightarrow t_n \frac{d}{dw} (\ln y_n) \Rightarrow$$

$$t_n \frac{1}{y_n} \frac{dy_n}{dw}$$

Using result from "q1" $\frac{d\delta(x)}{dx} \Rightarrow \delta(x)(1-\delta(x))$ and substitute $y_n = \delta(W^T \phi_n)$

$$\frac{dy_n}{dw} \Rightarrow y_n(1-y_n)\phi_n \Rightarrow$$

$$\frac{d}{dw} (t_n \ln y_n) \Rightarrow t_n \frac{1}{y_n} \times y_n(1-y_n)\phi_n \Rightarrow t_n(1-y_n)\phi_n$$

$$(II) \frac{d}{dw} ((1-t_n) \ln(1-y_n)) \Rightarrow (1-t_n) \frac{d}{dw} (\ln(1-y_n)) \Rightarrow$$

$$-(1-t_n) \frac{1}{1-y_n} \times \frac{d(1-y_n)}{dw}$$

$$\frac{d(1-y_n)}{dw} \Rightarrow -y_n(1-y_n)\phi_n$$

$$\frac{d}{dw} ((1-t_n) \ln(1-y_n)) \Rightarrow (1-t_n) \frac{1}{(1-y_n)} \times -y_n(1-y_n)\phi_n \Rightarrow (1-t_n)y_n\phi_n$$

$$\Rightarrow -\sum_{n=1}^N t_n(1-y_n)\phi_n - (1-t_n)y_n\phi_n \Rightarrow$$

$$-\sum_{n=1}^N (t_n - t_n y_n - y_n + t_n y_n)\phi_n \Rightarrow -\sum_{n=1}^N (t_n - y_n)\phi_n$$

$$\sum_{n=1}^N (y_n - t_n)\phi_n$$

3)

$$\hat{y}_n = \sigma(w^T \phi_n)$$

$$L(w) \Rightarrow - \sum_{n=1}^N \{ y_n \ln(\hat{y}_n) + (1-y_n) \ln(1-\hat{y}_n) \}$$

"BCE" function

$$\nabla L(w) \Rightarrow \sum_{n=1}^N (\hat{y}_n - y_n) \phi_n$$

$\phi \Rightarrow$ input
 $w \Rightarrow$ weight
 $\sigma \Rightarrow$ sigmoid

$$2.1 \frac{d}{dw} \ln(\hat{y}) \Rightarrow \frac{1}{\hat{y}} \frac{d\hat{y}}{dw} \Rightarrow \frac{1}{\hat{y}} \cdot \hat{y}(1-\hat{y}) \phi \Rightarrow (1-\hat{y}) \phi_n$$

$$2.2 \frac{d}{dw} \ln(1-\hat{y}) \Rightarrow \frac{1}{1-\hat{y}} \frac{d(1-\hat{y})}{dw} \Rightarrow \frac{1}{1-\hat{y}} \cdot -\hat{y}(1-\hat{y}) \phi \Rightarrow -\hat{y} \phi_n$$

$$3.1 \frac{d\hat{y}}{dw} \Rightarrow \frac{d\sigma(w^T \phi_n)}{dw} \Rightarrow \hat{y}_n (1-\hat{y}_n) \phi_n$$

$$3.2 \frac{d(1-\hat{y})}{dw} \Rightarrow -\hat{y}_n (1-\hat{y}_n) \phi_n$$

$$- \sum_{n=1}^N \left[y_n (1-\hat{y}_n) \phi_n + (1-y_n) \hat{y}_n \phi_n \right]$$

$$\left[y_n \phi_n - \hat{y}_n y_n \phi_n - \hat{y}_n \phi_n + y_n \hat{y}_n \phi_n \right] \Rightarrow$$

$$- \sum_{n=1}^N (y_n - \hat{y}_n) \phi_n \Rightarrow \underbrace{\sum_{n=1}^N (\hat{y}_n - y_n) \phi_n}_{\nabla L(w)}$$