

1.

(a)

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

$$\frac{\partial g(x)}{\partial x} = \frac{\partial \left(\frac{1}{1 + e^{-x}} \right)}{\partial x}$$

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

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

$$= \frac{A}{(1 + e^{-x})} + \frac{B}{(1 + e^{-x})^2}$$

$$A + Ae^{-x} + B = e^{-x}$$

$$\begin{cases} A = 1 \\ A + B = 0 \end{cases} \Rightarrow A = 1 \quad B = -1$$

$$\frac{\partial g(x)}{\partial x} = \frac{1}{1 + e^{-x}} - \frac{1}{(1 + e^{-x})^2}$$

$$= g(x) - g(x)^2$$

$$= g(x) (1 - g(x)) \quad \text{得證} \quad \#$$

(b)

$$G(-x) = \frac{1}{1 + e^x}$$

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

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

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

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

$$= 1 - G(x), \text{ 得證 } \#$$

(c)

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

$$y + ye^{-x} = 1$$

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

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

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

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

$$-x = \ln\left(\frac{1-y}{y}\right)$$

$$x = \ln\left(\frac{y}{1-y}\right), \text{ 得證 } \#$$

2.

$$\nabla L(w) = - \sum_{n=1}^N \left\{ y_n \frac{\partial}{\partial w} \ln(\sigma(w^T \phi_n)) + (1-y_n) \frac{\partial}{\partial w} \ln(1-\sigma(w^T \phi_n)) \right\}$$

令 $x = w^T \phi_n$, 使用 chain rule

$$= - \sum_{n=1}^N \left\{ y_n \frac{\partial(\ln(\sigma(x)))}{\partial x} \frac{\partial x}{\partial w} + (1-y_n) \frac{\partial}{\partial x} \ln(1-\sigma(x)) \frac{\partial x}{\partial w} \right\}$$

$$= - \sum_{n=1}^N \left\{ y_n \frac{1}{\sigma(x) (1-\sigma(x))} \frac{\partial x}{\partial w} - (1-y_n) \frac{1}{1-\sigma(x)} \right\}$$

根據 1.(a) 的證明

$$\times (1-\sigma(x)) \sigma(x) \frac{\partial x}{\partial w} \}$$

$$= - \sum_{n=1}^N \left\{ y_n (1-\sigma(x)) - (1-y_n) \sigma(x) \right\} \frac{\partial x}{\partial w}$$

$$= - \sum_{n=1}^N \left\{ y_n - \sigma(x) y_n - \sigma(x) + y_n \sigma(x) \right\} \frac{\partial x}{\partial w}$$

$$= - \sum_{n=1}^N \left\{ y_n - \sigma(x) \right\} \frac{\partial x}{\partial w}$$

$$= - \sum_{n=1}^N \left\{ y_n - \sigma(w^T \phi_n) \right\} \frac{\partial w^T \phi_n}{\partial w}$$

$$= - \sum_{n=1}^N \left\{ y_n - \hat{y}_n \right\} \phi_n = \sum_{n=1}^N (\hat{y}_n - y_n) \phi_n, \text{ 得證 } \#$$

3.
(a)

使 $x_n = \frac{y_n + 1}{2}$, $\hat{x}_n = \frac{\hat{y}_n + 1}{2}$ 將 range 轉回 $\{0, 1\}$

$$L = - \sum_{n=1}^N \{x_n \ln(\hat{x}_n) + (1-x_n) \ln(1-\hat{x}_n)\}$$

$$= - \sum_{n=1}^N \left\{ \frac{y_n + 1}{2} \ln\left(\frac{\hat{y}_n + 1}{2}\right) + \left(1 - \frac{y_n + 1}{2}\right) \ln\left(1 - \frac{\hat{y}_n + 1}{2}\right) \right\}$$

$$= - \sum_{n=1}^N \left\{ \frac{y_n + 1}{2} \ln\left(\frac{\hat{y}_n + 1}{2}\right) + \frac{1 - y_n}{2} \ln\left(\frac{1 - \hat{y}_n}{2}\right) \right\}$$

$$\propto - \sum_{n=1}^N \{ (y_n + 1)(\ln(\hat{y}_n + 1) - \ln 2) + (1 - y_n)(\ln(1 - \hat{y}_n) - \ln 2) \}$$

• $\frac{1}{2}$ 只是一個倍率關係，像 learning rate
- 樣

$$= - \sum_{n=1}^N \{ (y_n + 1) \ln(\hat{y}_n + 1) + (1 - y_n) \ln(1 - \hat{y}_n) - 2 \ln 2 \}$$

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

• $2 \ln 2$ 是常數項，在求梯度時就會為 0
，故可消除 #

(b) 原本 range 為 $\{0, 1\}$ 的 \hat{y}_n

$$\hat{y}_n \rightarrow (\hat{y}_n - \frac{1}{2}) \cdot 2$$

修正後 range 為 $\{-1, 1\}$ 的 \hat{y}_n

$$\Rightarrow \frac{2}{1 + e^{-x}} - 1$$

$$\Rightarrow \frac{2 - (1 + e^{-x})}{1 + e^{-x}}$$

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

$$\Rightarrow \tanh\left(\frac{x}{2}\right)$$

$$\Rightarrow \tanh\left(\frac{w^T \phi_n}{2}\right) \quad \#$$