

$$c) \ell(z, t) = - \sum_{i=1}^N t_i \ln(z_i) \quad , \quad t \in [0, 1]^{1 \times N}$$

$$D_z \ell(z, t) = \left[\frac{\partial \ell(z, t)}{\partial z_1}, \dots, \frac{\partial \ell(z, t)}{\partial z_N} \right]$$

$$\frac{\partial \ell(z, t)}{\partial z_i} = \frac{\partial (-t_i \ln(z_i))}{\partial z_i}$$

$$= -\frac{t_i}{z_i}$$

d) division by zero if $z_i \approx 0$

QUESTION 3:

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

$$\frac{\partial \tanh(x)}{\partial x} = \frac{\left(\frac{\partial}{\partial x} (e^x - e^{-x}) \right) (e^x + e^{-x}) - (e^x - e^{-x}) \left(\frac{\partial}{\partial x} (e^x + e^{-x}) \right)}{(e^x + e^{-x})^2}$$

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

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

$$= \underline{1 - \tanh^2 x}$$

QUESTION 4:

$$a) \text{softmax}(x) = \text{softmax}(x+c)$$

$$\text{softmax}_i(x+c) = \frac{e^{x_i+c}}{\sum_{j=1}^N e^{x_j+c}} = \frac{\cancel{e^c} e^{x_i}}{\cancel{e^c} \sum_{j=1}^N e^{x_j}} = \text{softmax}_i(x)$$

$$b) D_x \log(\sigma(x)) \in \mathbb{R}^{N \times N}$$

{log = ln}

$$\begin{aligned} \log \sigma_i(x) &= \log \frac{e^{x_i}}{\sum_{j=1}^N e^{x_j}} = \log e^{x_i} - \log \left(\sum_{j=1}^N e^{x_j} \right) \\ &= x_i - \log \left(\sum_{j=1}^N e^{x_j} \right) \end{aligned}$$

diagonal:

$$\frac{\partial}{\partial x_i} \log(\sigma_i(x)) = 1 - \frac{e^{x_i}}{\sum_{j=1}^N e^{x_j}} = \boxed{1 - \sigma_i(x)}$$

$$\star \frac{\partial \log(f(x))}{\partial x} = \frac{1}{f(x)} \cdot f'(x)$$

off-diagonal:

$$\begin{aligned} \frac{\partial}{\partial x_j} \log(\sigma_i(x)) &= \frac{\partial}{\partial x_j} \overset{0}{x_i} - \frac{\partial}{\partial x_j} \log \left(\sum_{k=1}^N e^{x_k} \right) \\ &= \frac{-e^{x_j}}{\sum_{k=1}^N e^{x_k}} = \boxed{-\sigma_j(x)} \end{aligned}$$

Combining:

$$\begin{bmatrix} 1 - \sigma_1(x) & -\sigma_2(x) & \dots & -\sigma_N(x) \\ -\sigma_1(x) & & & \\ \vdots & \ddots & & \\ -\sigma_1(x) & & & 1 - \sigma_N(x) \end{bmatrix}$$

c) $z = v \cdot D_x \log(\sigma(x))$

~~$z_1 =$~~ $v = [v_1, \dots, v_N]$

$$\begin{aligned} z_1 &= v_1(1 - \sigma_1(x)) - v_2 \sigma_1(x) - \dots - v_N \sigma_1(x) \\ &= v_1 - \sigma_1(x) \sum_{j=1}^N v_j \end{aligned}$$

$$z_i = v_i - \sigma_i(x) \sum_{j=1}^N v_j$$

d) with softmax: $l(z, t) = - \sum_{j=1}^N t_j \ln(z_j)$

with log-softmax: $l(z, t) = - \sum_{j=1}^N t_j z_j$

$$\frac{\partial l(z, t)}{\partial z_i} = - \frac{\partial}{\partial z_i} t_i z_i = -t_i$$

$$\nabla_z l(z, t) = [-t_1, \dots, -t_N] = \textcircled{-t}$$