

Exercise 1.

a)  $\max_x L = w^T x + b - \lambda x^T x$

$$\frac{\partial L}{\partial x} = w - 2\lambda x \stackrel{!}{=} 0 \Rightarrow \boxed{x^* = \frac{w}{2\lambda}}$$

b)  $\max_x L = \log P(w_c | x) + \log P(x)$

$$= w^T x + \log \frac{1}{\sqrt{\det(\Sigma)}} + -\frac{1}{2}(x-\mu)^T \Sigma^{-1}(x-\mu)$$

$$\frac{\partial L}{\partial x} = w - \Sigma^{-1}(x-\mu) \stackrel{!}{=} 0$$

$$\boxed{x^* = \Sigma w + \mu}$$

c)  $\max_{z \in \mathbb{R}} L = \log P(w_c | g(z)) - \lambda \|z\|^2$

$$= w^T \underbrace{x}_{Az+c} + b$$

$$L = w^T A z + w^T c + b - \lambda z^T z$$

$$\frac{\partial L}{\partial z} = A^T w - 2\lambda z \stackrel{!}{=} 0 \Rightarrow \boxed{z^* = \frac{A^T w}{2\lambda}}$$

$$\boxed{x^* = A z^* + c = \frac{A A^T w}{2\lambda} + c}$$

d) if  $A A^T = I$ ,  $c = 0$ ,

the ~~proto~~ prototype  $x^*$  from regularizers and generator are the same.

when  $\Sigma w + \mu = \frac{w}{2\lambda}$

the prototype  $x^*$  from regularizers and experts are the same.

Exercice 2

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~~at~~ ~~Exercice~~  $f(x) = w^T x = \sum_i w_i x_i$

a) Sensitivity analysis :

$$\frac{\partial f}{\partial x_i} = w_i \Rightarrow R_i(x) = \left( \frac{\partial f}{\partial x_i} \right)^2 = w_i^2 \Rightarrow R = \begin{bmatrix} w_1^2 \\ w_2^2 \\ \vdots \\ w_n^2 \end{bmatrix}$$

$$\sum R_i(x) = \sum w_i^2$$

b) Taylor decomposition at root point  $\tilde{x} = 0$

because  $\tilde{x} = 0 \Rightarrow \tilde{x}_i = 0$

$$R_i(x) = \underbrace{\frac{\partial f}{\partial x_i}}_{= w_i} \bigg|_{x=\tilde{x}} \cdot \underbrace{(x_i - \tilde{x}_i)}_{= x_i}$$

$$= w_i x_i$$

~~or~~ or because  $f(tx) = w^T tx = t w^T x = t f(x)$   
 $f(0) = 0$

choose  $\tilde{x} = \lim_{t \rightarrow 0} tx$

$$R_i(x) = \frac{\partial f}{\partial x_i} \bigg|_{x=\tilde{x}} \cdot x_i = w_i x_i$$

c) ~~because~~ because  $f(tx) = w^T tx = t w^T x = t f(x) \Rightarrow f(0) = 0$   
 and  $x$  have no limitation on its domain of definition  
 so the root point ~~should~~ should be ~~at~~  $x=0$   
 which included in  $x$ 's domain of definition.

$$\Rightarrow R_i = \frac{\partial f}{\partial x_i} \bigg|_{x=\tilde{x}} (x_i - \tilde{x}_i)$$

$$= w_i x_i$$

### Exercise 3.

a) because we have LRP- $a_1 p_0$  have:

$$R_j = \sum_k \frac{a_j w_{jk}}{\sum_k a_j w_{jk}} R_k$$

in network (a) :

$$R_2 = \frac{a_3 \cdot 1}{\sum a_3 \cdot 1} y = y$$

$$R_1 = \frac{a_1 \cdot 1}{\sum a_1 \cdot 1} R_2 = y$$

in network (b) :

$$R_2 = \frac{a_4 \cdot 1}{\sum a_4 \cdot 1} y = y$$

$$\textcircled{a} R_1 = \frac{a_2 \cdot 1}{\sum a_2 \cdot 1} y = y$$

in network (c)

$$R_2 = \frac{a_3 \cdot 0.5}{a_3 \cdot 0.5} y = y$$

$$\textcircled{a} R_{11} = \frac{a_1 \cdot 1}{a_1 \cdot 1 + a_2 \cdot 1} R_2 = \frac{a_1}{a_1 + a_2} y$$

$$R_{12} = \frac{a_2 \cdot 1}{a_1 + a_2} R_2 = \frac{a_2}{a_1 + a_2} y$$

b) ~~The~~ 3 function lead to the same result that  $y = \min(a_1, a_2)$ . But the network c give the most intuitive explanation, because in the hidden layer, it ~~parties~~ describe 3 different kind of relationship between  $a_1$  and  $a_2$ . ~~the~~ ~~the~~

~~$a_1 > a_2$~~   $a_1 = a_2$ ,  $a_1 < a_2$ ,  $a_1 \geq a_2$  respectively.