

4. put  $f: \mathbb{R} \rightarrow \mathbb{R}$  It is clear that  $\mathbb{R}$  is convex.  
 $d \mapsto -\log d$

Then  $f'(d) = \frac{1}{d} > 0$ .

By Taylor, we get 
$$\begin{cases} f(d_1) \geq f(\eta d_1 + (1-\eta)d_2) + f'(\eta d_1 + (1-\eta)d_2) (1-\eta)(d_2 - d_1) \\ f(d_2) \geq f(\eta d_1 + (1-\eta)d_2) + f'(\eta d_1 + (1-\eta)d_2) \eta (d_1 - d_2) \end{cases}$$

$$\Rightarrow f(d_1) + (1-\eta)f(d_2) \geq f(\eta d_1 + (1-\eta)d_2), \quad \eta \in (0,1), \quad \forall d_1, d_2 \in \mathbb{R}$$

$\therefore f(d) = -\log d$  is convex.

$$\begin{aligned} \text{Now, } D_{KL}(p \parallel q) &= \mathbb{E}_I \left( \log \frac{p_i}{q_i} \right), \quad p(I=i) = p_i \\ &= \mathbb{E}_I \left( -\log \frac{q_i}{p_i} \right) \end{aligned}$$

$$\text{By Jensen's, } \geq -\log \mathbb{E}_I \left( \frac{q_i}{p_i} \right)$$

$$= -\log \sum_i p_i \frac{q_i}{p_i}$$

$$= -\log 1$$

$$= 0$$



5. put  $f: \mathbb{R} \rightarrow \mathbb{R}$  It is clear that  $\mathbb{R}$  is convex.  
 $d \mapsto -\log d$

$f'(d) = \frac{1}{d} > 0$ .

By Taylor, we get 
$$\begin{cases} f(d_1) > f(\eta d_1 + (1-\eta)d_2) + f'(\eta d_1 + (1-\eta)d_2) (1-\eta)(d_2 - d_1) \\ f(d_2) > f(\eta d_1 + (1-\eta)d_2) + f'(\eta d_1 + (1-\eta)d_2) \eta (d_1 - d_2) \end{cases}$$

$\Rightarrow f(d_1) + (1-\lambda)f(d_2) > f(\eta d_1 + (1-\eta)d_2), \quad \eta \in (0,1), \quad \forall d_1, d_2 \in \mathbb{R}$   
 $d_1 \neq d_2$ .  
 $\therefore f(d) = -\log d$  is strictly convex.

Now,  $D_{KL}(p \parallel q) = \mathbb{E}_I \left( \log \frac{p_I}{q_I} \right), \quad p(I=i) = p_i, \quad p_i \neq q_i$   
 $= \mathbb{E}_I \left( -\log \frac{q_I}{p_I} \right)$

By Strict Jensen's,  $> -\log \mathbb{E}_I \left( \frac{q_I}{p_I} \right)$   
 $= -\log \sum_i p_i \frac{q_i}{p_i}$   
 $= -\log 1$   
 $= 0$



6.

i) Show  $\nabla_u f_\theta(x) = \sigma(ax+b)$

$$\frac{\partial}{\partial u_i} f_\theta(x) = \frac{\partial}{\partial u_i} \sum_{j=1}^p u_j \sigma(a_j x + b_j)$$

$$= \frac{\partial}{\partial u_i} (u_i \sigma(a_i x + b_i))$$

$$= \sigma(a_i x + b_i)$$

$$\therefore \nabla_u f_\theta(x) = \begin{bmatrix} \frac{\partial}{\partial u_1} f_\theta(x) \\ \vdots \\ \frac{\partial}{\partial u_p} f_\theta(x) \end{bmatrix} = \sigma(ax+b)$$

ii) Show  $\nabla_b f_\theta(x) = \sigma'(ax+b) \odot u$

$$\frac{\partial}{\partial b_i} f_\theta(x) = \frac{\partial}{\partial b_i} \sum_{j=1}^p u_j \sigma(a_j x + b_j)$$

$$= \frac{\partial}{\partial b_i} (u_i \sigma(a_i x + b_i))$$

$$= u_i \sigma'(a_i x + b_i) \cdot \frac{\partial}{\partial b_i} (a_i x + b_i)$$

$$= u_i \sigma'(a_i x + b_i)$$

$$\therefore \nabla_b f_\theta(x) = \begin{bmatrix} \frac{\partial}{\partial b_1} f_\theta(x) \\ \vdots \\ \frac{\partial}{\partial b_p} f_\theta(x) \end{bmatrix} = \begin{bmatrix} \sigma'(a_1 x + b_1) \cdot u_1 \\ \vdots \\ \sigma'(a_p x + b_p) \cdot u_p \end{bmatrix} = \sigma'(ax+b) \odot u =$$

$$= \begin{bmatrix} \sigma'(a_1 x + b_1) & & 0 \\ & \ddots & \\ 0 & & \sigma'(a_p x + b_p) \end{bmatrix} \begin{bmatrix} u_1 \\ \vdots \\ u_p \end{bmatrix} = \text{diag}(\sigma'(ax+b)) u$$

iii) Show  $\nabla_a f_\theta(x) = (\sigma'(ax+b) \odot u) x$

$$\frac{\partial}{\partial a_i} f_\theta(x) = \frac{\partial}{\partial a_i} \sum_{j=1}^p u_j \sigma(a_j x + b_j) = \frac{\partial}{\partial a_i} (u_i \sigma(a_i x + b_i))$$

$$= u_i \sigma'(a_i x + b_i) \frac{\partial}{\partial a_i} (a_i x + b_i) = u_i \sigma'(a_i x + b_i) x$$

$$\nabla_a f_\theta(x) = \begin{bmatrix} \frac{\partial}{\partial a_1} f_\theta(x) \\ \vdots \\ \frac{\partial}{\partial a_p} f_\theta(x) \end{bmatrix} = \begin{bmatrix} \sigma'(a_1 x + b_1) \cdot u_1 \cdot x \\ \vdots \\ \sigma'(a_p x + b_p) \cdot u_p \cdot x \end{bmatrix} = (\sigma'(ax+b) \odot u) x$$

$$= \begin{bmatrix} \sigma'(a_1 x + b_1) & & 0 \\ & \ddots & \\ 0 & & \sigma'(a_p x + b_p) \end{bmatrix} \begin{bmatrix} u_1 \\ \vdots \\ u_p \end{bmatrix} x = \text{diag}(\sigma'(ax+b)) u x$$

