

$$1) \textcircled{a} \quad H(x) = \sum_x p(x) \log_2 \left(\frac{1}{p(x)} \right)$$

$$0 \leq p(x) \leq 1 \Rightarrow \frac{1}{p(x)} \geq 1 \Rightarrow \log_2 \frac{1}{p(x)} \geq \log_2 1 = 0$$

\downarrow
 non-negative
 ①

$\Rightarrow \log_2 \frac{1}{p(x)} \geq 0$
 (positive) and zero
 ②

$$\textcircled{1} \text{ and } \textcircled{2} \Rightarrow \boxed{\sum_x p(x) \log_2 \frac{1}{p(x)} \geq 0}$$

$$1) \textcircled{b} \quad \text{Take } Y(x) = \frac{q(x)}{p(x)}$$

Take $\phi(x) = \log_2(x) \rightarrow \text{concave}$

$$E[\phi(Y)] = \sum_x p(x) \log_2 Y(x) = \sum_x p(x) \log_2 \frac{q(x)}{p(x)} = - \sum_x p(x) \log_2 \frac{p(x)}{q(x)}$$

$$\phi[E(Y)] = \log_2 \sum_x p(x) Y(x) = \log_2 \sum_x \cancel{p(x)} \frac{q(x)}{\cancel{p(x)}} = \log_2 1 = 0$$

by Jensen's inequality (since $\log_2(x)$ is ~~convex~~ concave):

$$\phi(E[x]) \leq E[\phi(x)] \Rightarrow \boxed{0 \leq - \left(- \sum_x p(x) \log_2 \frac{p(x)}{q(x)} \right)} \\ = KL(p \parallel q)$$

$$1) \textcircled{c} \quad KL(p(x,y) \parallel p(x)p(y)) = \sum_{x,y} p(x,y) \log \frac{p(x,y)}{p(x)p(y)}$$

$$= \sum_{x,y} p(x,y) \left(\log \frac{p(x,y)}{p(x)} - \log p(y) \right)$$

$$= \underbrace{\sum_{x,y} p(x) p(y|x) \log(y|x)}_{-H(Y|X)} - \underbrace{\sum_y \log p(y) \sum_x p(x,y)}_{-H(Y)}$$

$$= -(-H(Y)) - H(Y|X) = \boxed{H(Y) - H(Y|X)} \\ = I(Y; X)$$

Mahsa Naserifar
CSC411 - Fall 18

2) ~~9~~ Hypothesis $\frac{1}{2} \left(\frac{1}{m} \sum_{i=1}^m (h_i(x) - t)^2 \right) \leq \frac{1}{3} \sum_{i=1}^3 \frac{1}{2} (h_i(x) - t)^2$

consider: $\psi(x) = h_1(x) - t$

$$\Rightarrow E[Y(x)] = \frac{1}{n} \sum_{i=1}^n (h_i - t) \Rightarrow \phi(E[Y]) = \frac{1}{2} \left(\frac{1}{n} \sum_{i=1}^n (h_i - t) \right)^2$$

consider: $\phi(y) = \frac{(h_i - t)^2}{2} \Rightarrow E[\phi(Y)] = \frac{1}{n} \sum_{i=1}^n \frac{(h_i - t)^2}{2}$

$\frac{y^2}{2}$ is convex

by the Jensen's inequality: $\Phi(E[Y]) \leq E[\Phi(Y)]$

$$\Rightarrow \underbrace{\frac{1}{2} \left(\frac{1}{n} \sum_{i=1}^n (h_i - t) \right)^2}_{L(\bar{h}, t)} < \frac{1}{n} \sum_{i=1}^n \underbrace{\frac{(h_i - t)^2}{2}}_{L(h_i, t)}$$

3) consider $w_b = \sum_{i=1}^N w_i \mathbb{I} \{h_t \neq 1\}$
and $w_g = \sum_{i=1}^N w_i \mathbb{I} \{h_t = 1\}$

and consider $w = \sum_{i=1}^N w_i$

$$\text{err}' = \frac{w_b'}{w'} = \frac{w_b e^{(-\alpha)x(-1)}}{w_b e^{(-\alpha)(-1)} w_g^{-\alpha}} = \frac{w_b e^{\alpha}}{w_b^{\alpha} e^{\alpha} + w_g^{-\alpha}}$$

$$= \frac{w_b \sqrt{\frac{1 - \text{err}}{\text{err}}}}{w_b \sqrt{\frac{1 - \text{err}}{\text{err}}} + w_g \sqrt{\frac{\text{err}}{1 - \text{err}}}}$$

$$\therefore \sqrt{\frac{1 - \text{err}}{\text{err}}} = \sqrt{\frac{1 - \frac{w_b}{w}}{\frac{w_b}{w}}} = \sqrt{\frac{\frac{w - w_b}{w}}{\frac{w_b}{w}}} = \sqrt{\frac{w - w_b}{w_b}}$$

$$\therefore \sqrt{\frac{err}{1-err}} = \sqrt{\frac{\frac{wb}{w}}{\frac{wg}{w}}} = \sqrt{\frac{wb}{wg}}$$

$$\therefore \text{err} = \frac{w_b \sqrt{\frac{w_g}{w_b}}}{w_b \sqrt{\frac{w_g}{w_b}} + w_g \sqrt{\frac{w_b}{w_g}}} = \frac{\sqrt{w_b w_g}}{2 \sqrt{w_b w_g}} = \frac{1}{2}$$

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