

Misc (Xiaowei Song)

Hölder's Inequality:

$$|u^T v| \leq (\sum |u_i|^p)^{\frac{1}{p}} (\sum |v_i|^q)^{\frac{1}{q}}, \text{ where } \frac{1}{p} + \frac{1}{q} = 1$$

For

$$\Psi(z) = \log \left(\sum_{i=1}^k e^{-z_i} \right), z \in R^k$$

1. Ψ is convex (5 points)

Proof:

$$\text{let } z_\lambda = \frac{1}{p}x + \frac{1}{q}y, \text{ where } x, y \in R^k, p, q \in R^+ \text{ and } \frac{1}{p} + \frac{1}{q} = 1, \lambda = \frac{1}{p}, 1 - \lambda = \frac{1}{q}, 0 < \lambda < 1$$

$$\begin{aligned} \Psi \left(\frac{1}{p}x + \frac{1}{q}y \right) &= \Psi(z_\lambda) = \log \left(\sum_{i=1}^k e^{-z_i} \right) \\ &= \log \left(\sum_{i=1}^k \exp \left[-\left(\frac{1}{p}x_i + \frac{1}{q}y_i \right) \right] \right) \\ &= \log \left(\sum_{i=1}^k \exp \left[-\left(\frac{1}{p}x_i \right) \right] \exp \left[-\left(\frac{1}{q}y_i \right) \right] \right) \\ &= \log \left(\sum_{i=1}^k [\exp(-x_i)]^{\frac{1}{p}} [\exp(-y_i)]^{\frac{1}{q}} \right) \\ &= \log (u^T v) \end{aligned}$$

where

$$u = \begin{bmatrix} [\exp(-x_1)]^{\frac{1}{p}} \\ [\exp(-x_2)]^{\frac{1}{p}} \\ \vdots \\ [\exp(-x_k)]^{\frac{1}{p}} \end{bmatrix}, v = \begin{bmatrix} [\exp(-y_1)]^{\frac{1}{q}} \\ [\exp(-y_2)]^{\frac{1}{q}} \\ \vdots \\ [\exp(-y_k)]^{\frac{1}{q}} \end{bmatrix}$$

thus from Hölder's Inequality, we can have

$$\begin{aligned} u^T v &= |u^T v| \leq \left(\sum |u_i|^p \right)^{\frac{1}{p}} \left(\sum |v_i|^q \right)^{\frac{1}{q}} \\ &= \left(\sum_{i=1}^k \left| [\exp(-x_i)]^{\frac{1}{p}} \right|^p \right)^{\frac{1}{p}} \left(\sum_{i=1}^k \left| [\exp(-x_i)]^{\frac{1}{p}} \right|^p \right)^{\frac{1}{q}} \\ &= \left(\sum_{i=1}^k \exp(-x_i) \right)^{\frac{1}{p}} \left(\sum_{i=1}^k \exp(-y_i) \right)^{\frac{1}{q}} \end{aligned}$$

(where $u^T v > 0$ since $u_i > 0, v_i > 0, i = 1, \dots, k$, thus $u^T v = |u^T v|$)
thus

$$\begin{aligned} \Psi \left(\frac{1}{p}x + \frac{1}{q}y \right) &= \Psi(z_\lambda) = \log (u^T v) \\ &\leq \log \left(\sum_{i=1}^k \exp(-x_i) \right)^{\frac{1}{p}} \left(\sum_{i=1}^k \exp(-y_i) \right)^{\frac{1}{q}} \\ &= \frac{1}{p} \log \left(\sum_{i=1}^k \exp(-x_i) \right) + \frac{1}{q} \log \left(\sum_{i=1}^k \exp(-y_i) \right) \\ &= \frac{1}{p} \Psi(x) + \frac{1}{q} \Psi(y) \end{aligned}$$

which proved convexity of function Ψ

2.

$$\Psi(y) - \Psi(z) \leq \sum_{i=1}^k (z_i - y_i) \frac{\exp(-y_i)}{\sum_{i=1}^k \exp(-y_i)}$$