

$$\# 1-1 \quad f(x, y, z) = x^2 + y^2 + 1.5z^2 + xy + yz + zx$$

$$(1) \frac{\partial f}{\partial x} = 2x + y + z$$

$$\frac{\partial f}{\partial y} = 2y + x + z$$

$$\frac{\partial f}{\partial z} = 3z + y + x$$

$$\frac{\partial^2 f}{\partial x^2} = 2, \quad \frac{\partial^2 f}{\partial y^2} = 2, \quad \frac{\partial^2 f}{\partial z^2} = 3$$

$$\frac{\partial^2 f}{\partial x \partial y} = 1, \quad \frac{\partial^2 f}{\partial y \partial z} = 1, \quad \frac{\partial^2 f}{\partial z \partial x} = 1$$

$$A = \begin{bmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 3 \end{bmatrix}$$

주대각선 소행렬식 정리

$$D_1 = 2 > 0$$

$$D_2 = 4 - 1 = 3 > 0$$

$$D_3 = \det(A) = 2(6-1) - 1(3-1) + 1(1-2) \\ = 7 > 0$$

D_1, D_2, D_3 모두 양수이므로 A 는 PD.

(2)

$$\text{adj}(A) = \begin{bmatrix} 5 & -2 & -1 \\ -2 & 5 & -1 \\ -1 & -1 & 3 \end{bmatrix}$$

$$A^{-1} = \frac{1}{\det(A)} \begin{bmatrix} 5 & -2 & -1 \\ -2 & 5 & -1 \\ -1 & -1 & 3 \end{bmatrix} = \frac{1}{7} \begin{bmatrix} 5 & -2 & -1 \\ -2 & 5 & -1 \\ -1 & -1 & 3 \end{bmatrix}$$

(3) \$xy\$ 중에 \$kxy\$가 있어 \$\det(A') = 0\$인 상황

$$A' = \begin{bmatrix} 2 & k & 1 \\ k & 2 & 1 \\ 1 & 1 & 3 \end{bmatrix} \quad \det(A') = -3k^2 + 2k + 8 = 0$$

$\det(A') = 0$ 이면 A' 는 non-invertible.

Non-invertible 같은 경우 $Ax = b$ 에서 어떤 b 에 대해서 3개가 존재하지 않거나 무수히 많이 존재함.
따라서 b 에 유일한 해라는 조건이 만족하지 않음.

2

$$B = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$M = B^T B = \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\det(M - \lambda I) = \begin{bmatrix} 1-\lambda & 1 & 0 \\ 1 & 1-\lambda & 0 \\ 0 & 0 & 1-\lambda \end{bmatrix}$$

$$\begin{aligned} & (1-\lambda)^2((1-\lambda)^2 - 1) \\ &= (1-\lambda)^2(1-\lambda+1)(1-\lambda-1) \\ &= (1-\lambda)^2(2-\lambda)(-\lambda) \\ &= -\lambda(1-\lambda)(2-\lambda) \end{aligned}$$

$$\lambda_1 = 2, \lambda_2 = 1, \lambda_3 = 0$$

$$\sigma_1 = \sqrt{2}, \sigma_2 = 1, \sigma_3 = 0$$

$$\text{i) } \lambda_1 = 2 \quad (M - 2I)W = 0$$

$$M - 2I = \begin{bmatrix} -1 & 1 & 0 \\ 1 & -1 & 0 \\ 0 & 0 & -1 \end{bmatrix} \begin{matrix} W_1 \\ W_2 \\ W_3 \end{matrix}$$

$$V_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$$

$-W_1 + W_2 = 0, \quad W_3 = 0$

$W_1 = W_2$

$$\text{ii) } \lambda_2 = 1 \quad (M - I)W = 0$$

$$M - I = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{matrix} W_1 \\ W_2 \\ W_3 \end{matrix}$$

$$V_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$V = \begin{bmatrix} \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & 0 & -\frac{1}{\sqrt{2}} \\ 0 & 1 & 0 \end{bmatrix}$$

$W_2 = 0, \quad W_1 = 0, \quad W_3 \neq 0$

$$\text{iii) } \lambda_3 = 0 \quad M W = 0$$

$$M = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{matrix} W_1 \\ W_2 \\ W_3 \end{matrix}$$

$$V_3 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$$

$$W_1 + W_2 = 0, \quad W_3 = 0$$

$$W_1 = -W_2, \quad W_3 = 0$$

$$BB^T = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$$

$$U_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad U_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad U = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\Sigma = \begin{bmatrix} \sqrt{2} & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

$$B = U \Sigma V^T$$

$$U = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \Sigma = \begin{bmatrix} \sqrt{2} & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \quad V^T = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ 0 & 0 & 1 \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 \end{bmatrix}$$

3-1

$$(1). \quad C_1 = \{(x, y) \in \mathbb{R}^2 : x+y \leq 1\}$$

Suppose that $(x_1, y_1), (x_2, y_2) \in C_1$

$$x_1 + y_1 \leq 1, \quad x_2 + y_2 \leq 1$$

$$\theta \in [0, 1]$$

$$(x_\theta, y_\theta) = \theta(x_1, y_1) + (1-\theta)(x_2, y_2)$$

$$x_\theta = \theta x_1 + (1-\theta)x_2$$

$$y_\theta = \theta y_1 + (1-\theta)y_2$$

$$x_\theta + y_\theta = \theta(x_1 + y_1) + (1-\theta)(x_2 + y_2) \leq \theta \cdot 1 + (1-\theta) \cdot 1 = 1$$

$$\therefore (x_\theta, y_\theta) \in C_1, \quad C_1 \text{ is convex.}$$

$$C_2 = \{x \in \mathbb{R}^n : \|x\|_1 \leq 1\}$$

Suppose that $x_1, x_2 \in C_2$

$$\|x_1\|_1 \leq 1, \|x_2\|_1 \leq 1$$

$$\theta \in [0, 1], x_\theta = \theta x_1 + (1-\theta)x_2 \in C_2.$$

ℓ_1 -norm의 정의에 따르면

$$\begin{aligned}\|x_\theta\|_1 &= \|\theta x_1 + (1-\theta)x_2\|_1 \leq \|\theta x_1\|_1 + \|(1-\theta)x_2\|_1 \\ &= \theta \|x_1\|_1 + (1-\theta) \|x_2\|_1.\end{aligned}$$

우리가 살펴 보았던

$$\|x_\theta\|_1 \leq \theta \cdot 1 + (1-\theta) \cdot 1 = 1$$

$\therefore x_\theta \in C_2, C_2 \subseteq \text{Convex}$

$$C_3 = \{(x, y) \in \mathbb{R}^2 : y \geq e^x\}$$

Suppose that $(x_1, y_1), (x_2, y_2) \in C_3$

$$y_1 \geq e^{x_1}, y_2 \geq e^{x_2}$$

$$\theta \in [0, 1]$$

$$(x_\theta, y_\theta) = \theta(x_1, y_1) + (1-\theta)(x_2, y_2) \in C_2$$

$$x_\theta = \theta x_1 + (1-\theta)x_2, y_\theta = \theta y_1 + (1-\theta)y_2$$

$y \geq e^x$ 은 convex,

$$e^{x_\theta} = e^{\theta x_1 + (1-\theta)x_2} \leq \theta e^{x_1} + (1-\theta)e^{x_2}$$

$$y_\theta = \theta y_1 + (1-\theta)y_2 \geq \theta e^{x_1} + (1-\theta)e^{x_2} \geq e^{x_\theta}$$

$$y_\theta = \theta y_1 + (1-\theta)y_2 \geq \theta e^{x_1} + (1-\theta)e^{x_2} \geq e^{x_\theta}$$

$\therefore (x_\theta, y_\theta) \in C_3, C_3 \subseteq \text{Convex}.$

(2)

$$S = \{(x, t) \in \mathbb{R}^n \times \mathbb{R} : t \geq f(x)\}$$

Suppose that $(x_1, t_1), (x_2, t_2) \in S$
 $t_1 \geq f(x_1), t_2 \geq f(x_2)$

$$\theta \in [0, 1] \text{ or } \text{why } (x_\theta, t_\theta) = \theta(x_1, t_1) + (1-\theta)(x_2, t_2) \text{ is true.}$$

$$\rightarrow x_\theta = \theta x_1 + (1-\theta)x_2, t_\theta = \theta t_1 + (1-\theta)t_2$$

f is convex or

$$f(x_\theta) = f(\theta x_1 + (1-\theta)x_2) \leq \theta f(x_1) + (1-\theta)f(x_2)$$

or t_1, t_2 are also convex

$$t_\theta = \theta t_1 + (1-\theta)t_2 \geq \theta f(x_1) + (1-\theta)f(x_2) \geq f(x_\theta)$$

$\therefore (x_\theta, t_\theta) \in S$. $S \subseteq$ convex set

3-2

(1)-a.

h convex $\Leftrightarrow \forall x, y, \forall \theta \in [0, 1], h(\theta x + (1-\theta)y) \leq \theta h(x) + (1-\theta)h(y)$

$$h(x) = f(x) + g(x)$$

$$h(\theta x + (1-\theta)y) = f(\theta x + (1-\theta)y) + g(\theta x + (1-\theta)y)$$

f, g is convex or

$$f(\theta x + (1-\theta)y) \leq \theta f(x) + (1-\theta)f(y)$$

$$g(\theta x + (1-\theta)y) \leq \theta g(x) + (1-\theta)g(y)$$

$$h(\theta x + (1-\theta)y) \leq \theta f(x) + g(x) + (1-\theta)(f(y) + g(y))$$

$$h(\theta x + (1-\theta)y) \leq \theta h(x) + (1-\theta)h(y)$$

$\therefore f(x) + g(x)$ is convex

(1) - b.

$$h(x) = f(Ax+b) \text{ 은 } \text{두자.}$$

만약 $x, y, \theta \in [0, 1]$ 이면

$$h(\theta x + (1-\theta)y) = f(A(\theta x + (1-\theta)y) + b)$$

정의의 전제성

$$A(\theta x + (1-\theta)y) + b = \theta(Ax+b) + (1-\theta)(Ay+b)$$

따라서

$$h(\theta x + (1-\theta)y) + b = f(\theta(Ax+b) + (1-\theta)(Ay+b))$$

$f \rightarrow$ convex 이면

$$f(\theta u + (1-\theta)v) \leq \theta f(u) + (1-\theta)f(v)$$

$$u = Ax+b, v = Ay+b \text{ 은 } \text{정의}$$

$$h(\theta x + (1-\theta)y) \leq \theta h(x) + (1-\theta)h(y)$$

$$h(\theta x + (1-\theta)y) \leq \theta h(x) + (1-\theta)h(y)$$

$\therefore f(Ax+b)$ 은 convex.

$$(2) f(x) = \|Ax-b\|^2, \quad \text{만약 } x, y \in \mathbb{R}^d, \theta \in [0, 1]$$

$$u := Ax-b, \quad v := Ay-b$$

$$f(x) = \|u\|^2, \quad f(y) = \|v\|^2$$

$$f(\theta x + (1-\theta)y) = \|A(\theta x + (1-\theta)y) - b\|^2$$

정의성

$$A(\theta x + (1-\theta)y) - b = \theta(Ax-b) + (1-\theta)(Ay-b) = \theta u + (1-\theta)v$$

$$f(\theta x + (1-\theta)y) = \|\theta u + (1-\theta)v\|^2$$

$$\begin{aligned}\|\theta u + (1-\theta)v\|^2 &= (\theta u + (1-\theta)v)^T (\theta u + (1-\theta)v) \\ &= \theta^2 \|u\|^2 + (1-\theta)^2 \|v\|^2 + 2\theta(1-\theta) u^T v\end{aligned}$$

차이 계산

$$\begin{aligned}D &:= \theta \|u\|^2 + (1-\theta)\|v\|^2 - \|\theta u + (1-\theta)v\|^2 \\ &= \theta \|u\|^2 + (1-\theta)\|v\|^2 - [\theta^2 \|u\|^2 + (1-\theta)^2 \|v\|^2 + 2\theta(1-\theta) u^T v] \\ &= \theta(1-\theta) \|u\|^2 + \theta(1-\theta) \|v\|^2 - 2\theta(1-\theta) u^T v \\ &= \theta(1-\theta) (\|u\|^2 + \|v\|^2 - 2u^T v) \\ &= \theta(1-\theta) (u-v)^T (u-v) \\ &= \theta(1-\theta) \|u-v\|^2\end{aligned}$$

$$\theta \in [0, 1] \Rightarrow \theta(1-\theta) \geq 0, \|u-v\|^2 \geq 0$$

$$D = \theta(1-\theta) \|u-v\|^2 \geq 0$$

$$\|\theta u + (1-\theta)v\|^2 \leq \theta \|u\|^2 + (1-\theta) \|v\|^2$$

원래 변수로 돌아오면

$$f(\theta x + (1-\theta)y) \leq \theta f(x) + (1-\theta)f(y)$$

$$\therefore f(x) = \|Ax-b\|^2 \text{은 convex.}$$

#4-1

| X | Y | 0 | 1 |
|---|---|---------------|---------------|
| 0 | | $\frac{1}{3}$ | $\frac{1}{3}$ |
| 1 | | 0 | $\frac{1}{3}$ |

$$\begin{aligned}P(X=0) &= \frac{2}{3} & P(Y=0) &= \frac{1}{3} \\ P(X=1) &= \frac{1}{3} & P(Y=1) &= \frac{2}{3}\end{aligned}$$

$$\begin{aligned}(a) H(X) &= -\frac{2}{3} \log_2 \frac{2}{3} - \frac{1}{3} \log_2 \frac{1}{3} & H(Y) : \text{디지털} \\ &= -\frac{2}{3} (1 - \log_2 3) + \frac{1}{3} \log_2 3 & H(Y) \approx 0.9183 \\ &= \log_2 3 - \frac{2}{3} \approx 0.9183\end{aligned}$$

$$(b) H(X|Y)$$

$$\text{i)} Y=0 \rightarrow X=0 \text{ 有り}, H(X|Y=0)=0$$

$$\text{ii)} Y=1 \quad P(X=0|Y=1) = \frac{1}{3}/\frac{2}{3} = \frac{1}{2}, \quad H(X|Y=1) = 1$$

$$H(X|Y) = P(Y=0) \cdot 0 + P(Y=1) \cdot 1 = \frac{2}{3}$$

$$H(Y|X) \text{ 有り}.$$

$$H(Y|X) = \frac{2}{3}$$

$$(c) H(X,Y) = H(Y) + H(X|Y)$$

$$= (\log_2 3 - \frac{2}{3}) + \frac{2}{3}$$

$$= \log_2 3 \approx 1.5850$$

$$(d) H(Y) - H(Y|X) = \left(\log_2 3 - \frac{2}{3}\right) - \frac{2}{3}$$

$$= \log_2 3 - \frac{4}{3} \approx 0.2516$$

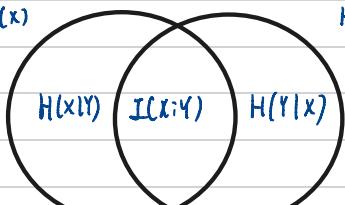
$$(e) I(X;Y) = H(X) - H(X|Y)$$

$$= \left(\log_2 3 - \frac{2}{3}\right) - \frac{2}{3}$$

$$= \log_2 3 - \frac{4}{3} \approx 0.2516.$$

$$(f) H(x)$$

$$H(Y)$$



$$H(X,Y)$$

4-2

(a) $p = (0.9, 0.1)$, $q = (0.5, 0.5)$

$$D(p||q) = \sum_i q_i \ln \frac{p_i}{q_i} = 0.5 \ln \frac{0.9}{0.5} + 0.5 \ln \frac{0.9}{0.5} \approx 0.5108$$

$$D(q||p) = \sum_i p_i \ln \frac{p_i}{q_i} = 0.9 \ln \frac{0.9}{0.5} + 0.1 \ln \frac{0.1}{0.5} \approx 0.3681$$

(b)

$$D(p||q) = \sum_x p(x) \ln \frac{p(x)}{q(x)} = - \sum_x p(x) \ln \frac{q(x)}{p(x)} = - E_p \left[\ln \left(\frac{q(x)}{p(x)} \right) \right]$$

$\ln(\cdot)$ 凸 凹 函数.

$$E_p[\ln Z] \leq \ln(E_p[Z])$$

$$Z = \frac{\pi(x)}{p(x)} \geq 1.$$

$$E_p \left[\ln \left(\frac{q(x)}{p(x)} \right) \right] \leq \ln \left(E_p \left[\frac{q(x)}{p(x)} \right] \right) = \ln \left(\sum_x p(x) \frac{q(x)}{p(x)} \right)$$

$$= \ln \left(\sum_x q(x) \right)$$

$$= \ln 1 = 0$$

$$\therefore D(p||q) = - E_p \left[\ln \left(\frac{q(x)}{p(x)} \right) \right] \geq 0$$