

1. $\text{Set } A = \{\vec{a} \in \mathbb{R}^k \mid p(\vec{a}) = 1\} = \{\vec{a} \in \mathbb{R}^k \mid a_0 = 1\}$. If $\vec{x}_1, \vec{x}_2 \in A \Rightarrow \theta \vec{x}_1 + (1-\theta) \vec{x}_2 = \begin{bmatrix} \theta x_{01} + (1-\theta) x_{02} \\ \vdots \\ \theta x_{k1} + (1-\theta) x_{k2} \end{bmatrix} = \begin{bmatrix} 1 \\ \vdots \\ \theta x_{k1} + (1-\theta) x_{k2} \end{bmatrix}$ for $\theta \in \mathbb{R} \Rightarrow A$ is affine $\Rightarrow A$ is convex.

$\text{Set } B = \{\vec{a} \in \mathbb{R}^k \mid 1P(\vec{a}) \leq 1 \text{ for } \alpha \leq t \leq \beta\} = \{\vec{a} \in \mathbb{R}^k \mid \vec{c}^T \vec{a} \leq 1, -\vec{c}^T \vec{a} \leq -1, \vec{c} = \begin{bmatrix} 1 \\ \vdots \\ t+k-1 \end{bmatrix}, \alpha \leq t \leq \beta\}$

For a given t , B_t is a polyhedron \Rightarrow convex with respect to $\vec{a} \Rightarrow B$ is an intersection of infinite convex set $\Rightarrow B$ is convex with respect to \vec{a} .

$\Rightarrow \{\vec{a} \in \mathbb{R}^k \mid p(\vec{a}) = 1; 1P(\vec{a}) \leq 1 \text{ for } \alpha \leq t \leq \beta\}, P(t) = A + B_t, A \cap B_t = A \cap B$ is an intersection of 2 convex set \Rightarrow convex.

2. $\text{dom } f = \{(\vec{x}, t) \in \mathbb{R}^n \times \mathbb{R} \mid \|\vec{x}\| \leq t\}$ is a norm cone $\Rightarrow \text{dom } f$ is convex

$$f(\vec{x}, t) = -\log(t^2 - \vec{x}^T \vec{x}) = \mathbb{R} \log t - \log(t - \frac{\vec{x}^T \vec{x}}{t}) \quad \text{for } (\vec{x}, t) = t^2 - \vec{x}^T \vec{x}, \text{ dom } f = \{(\vec{x}, t) \mid t > 0\}, \text{ dom } h = \text{dom } f \Rightarrow f = g \circ h$$

$f_1(\vec{x}, t) = -\log t$ is convex and nonincreasing.

$f_2(\vec{x}, t) = \frac{1}{t} - \frac{\vec{x}^T \vec{x}}{t^2}$ is concave, $\frac{\vec{x}^T \vec{x}}{t^2}$ is quadratic-over-linear function

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$f_1(\vec{x}, t) = -\log t$ is convex and nonincreasing.

$f_2(\vec{x}, t) = \frac{1}{t} - \frac{\vec{x}^T \vec{x}}{t^2} = \frac{1}{t} + \frac{1}{t} - \frac{\vec{x}^T \vec{x}}{t^2}$ is concave, $\frac{\vec{x}^T \vec{x}}{t^2}$ is quadratic-over-linear function

$\Rightarrow f_2(\vec{x}, t)$ is sum of concave functions $\Rightarrow f_2(\vec{x}, t)$ is concave, and $t > \|\vec{x}\|, f_2(\vec{x}, t) > 0$

$\Rightarrow -\log(t - \frac{\vec{x}^T \vec{x}}{t})$ is the composition of a convex and nondecreasing function f_1 with a concave

function $f_2 \Rightarrow -\log(t - \frac{\vec{x}^T \vec{x}}{t})$ is convex

$\Rightarrow f(\vec{x}, t)$ is sum of two convex function $\Rightarrow f(\vec{x}, t)$ is convex.

$\Rightarrow f(x,t)$ is sum of two convex function $\Rightarrow f(x,t)$ is convex *

PSD inside

3. $\text{dom } f = \{x \in \mathbb{R}^2 \mid x_1 > 0, x_1 - \frac{1}{x_2} > 0\}$, If $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} \in f \Rightarrow x_1, y_2 > 0, x_1 x_2 > 1, y_1 y_2 > 1$

\Rightarrow for $\theta \in [0,1]$, $\theta x_2 + (1-\theta)y_2 > 0$

$$(\theta x_1 + (1-\theta)y_1)(\theta x_2 + (1-\theta)y_2) \geq \theta^2 x_1 x_2 + (1-\theta)^2 y_1 y_2 > \theta^2 + (1-\theta)^2 = 2\theta^2 - 2\theta + 1 \geq 1$$

$$\Rightarrow \theta x_1 + (1-\theta)y_1 \geq \frac{1}{\theta x_2 + (1-\theta)y_2} > 0$$

$\Rightarrow \theta \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + (1-\theta) \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} \in f, \theta \in [0,1] \Rightarrow \text{dom } f \text{ is convex}$

$$f(\vec{x}) = \frac{1}{x_1 - \frac{1}{x_2}} = g(h(x_1, x_2)), g(x) = \frac{1}{x}, x > 0, h(\vec{x}) = x_1 - \frac{1}{x_2}, x_2 > 0, x_1 - \frac{1}{x_2} > 0$$

$\frac{1}{x}$ is convex and nonincreasing for $x > 0$, $h(\vec{x}) = x_1 + (-\frac{1}{x_2})$ is sum of two concave function \Rightarrow $h(\vec{x})$ is concave

$\Rightarrow f(\vec{x})$ is a composition of a convex and nondecreasing function $g(x) = \frac{1}{x}$ and a concave function $h(\vec{x}) = x_1 - \frac{1}{x_2}$

$\Rightarrow f(\vec{x})$ is a convex function *

② consider $g: \mathbb{R}^2 \rightarrow \mathbb{R}$, $g(\vec{x}) = \frac{1}{x_1 - \frac{1}{x_2}}, x_2 > 0, x_1 - \frac{1}{x_2} > 0, h: \mathbb{R}^2 \rightarrow \mathbb{R}^2, h(\vec{x}) = \begin{bmatrix} x_1 \\ x_2 - \frac{1}{x_1} \end{bmatrix}, x_1, x_2 - \frac{1}{x_1} > 0, x_1 - \frac{1}{x_2 - \frac{1}{x_1}} > 0$

$g(\vec{x})$ is convex by ①. Besides, $\frac{\partial g(\vec{x})}{\partial x_1} = -\frac{1}{(x_1 - \frac{1}{x_2})^2} < 0, \frac{\partial g(\vec{x})}{\partial x_2} = -\frac{1}{x_2^2(x_1 - \frac{1}{x_2})^2} < 0 \Rightarrow g(\vec{x})$ is convex and nonincreasing in each argument.

x_1 is concave, $x_2 - \frac{1}{x_1} = x_2 + (-\frac{1}{x_1})$ is sum of two concave function \Rightarrow is concave $\Rightarrow h_1, h_2$ are concave

$\Rightarrow f = g \circ h = \frac{1}{x_1 - \frac{1}{x_2 - \frac{1}{x_1}}}, x_1, x_2 - \frac{1}{x_1} > 0, x_1 - \frac{1}{x_2 - \frac{1}{x_1}} > 0$ is a composition of a convex and nonincreasing in each argument

function $g(\vec{x})$ and a concave fn h_1, h_2 function $h(\vec{x}) \Rightarrow f$ is convex *

Kennedy.
- Particle swarm optimization (PSO) by Eberhart &
- Ant colony optimization (ACO)