

Convex

in 1 variable

• $f(x) = ax + b$

• $f(x) = x^2 + bx + c$

• $f(x) = |x|$

• $f(x) = -\ln(x)$ for $x > 0$

• $f(x) = \frac{1}{x}$ for $x > 0$ satisfying $Ax < b$

• $f(x) = e^x$

in n variables

• $f(x) = a^T x + b$

• $f(x) = x^T M x + c^T x$ where M is PSD

• $f(x) = \|x\|$ where $\| \cdot \|$ in any norm

• $\sum_{i=1}^m -\ln(b_i - a_i^T x)$ for x

判断凸优化的方法 (which of the following optimisation problem is convex)

①. objective

minimize convex ☒ maximize concave

②. feasible region

convex $\leq c$

concave $\geq c$

Addition: if $f_1(x)$ and $f_2(x)$ are convex, $a, b > 0$

$f(x) := a \cdot f_1(x) + b \cdot f_2(x)$ is also convex

Maximum: if $f_i(x)$ are convex, $i=1, 2$

$f(x) = \max_i f_i(x)$ is also convex

Composition: if $f(x)$ is convex, then

$g(y) := f(Ay + b)$ is also convex

Nonlinear Programming Exercises

Objective:

(1), minimise $\sum_{i=1}^m (y_i - b_0 + b_1 x_{i1} + \dots + b_n x_{in})^2$

- y_i, x_{i1}, x_{i2} are all affine
- $y_i - b_0 - x_{i1}b_1 - \dots - x_{in}b_n$ is affine
- $(y_i - b_0 - x_{i1}b_1 - \dots - x_{in}b_n)^2$ is affine

so it is convex

Constraints:

(1) - ①. $b_0, b_1, \dots, b_n \in \mathbb{R}$

• entire space \mathbb{R}^n

• so convex

(1) - ②. $b_j \geq -10, b_j \leq 10$

• linear constraints

• so convex

(1) - ③. $b_1 \geq 2b_2$

• $b_1 - 2b_2 \leq 0$

• linear constraint

• so convex

(1) - ④. $b_3 = b_4$

• $b_3 - b_4 = 0$

• linear constraint

• so convex

(1) - ⑤. $|b_j| \leq 10$

• Adding auxiliary variables γ_j

• $\gamma_j \geq b_j, \gamma_j \geq -b_j$

• $\gamma_1 + \gamma_2 + \dots + \gamma_{10} \leq 10$

• so convex

(1) - ⑥. At most 5 of the slopes should be non-zero

• not convex

$$11) - ⑦. \quad b_5 \leq 1 \quad \text{or} \quad b_5 \geq 2$$

• not convex

$$(2). \text{ minimise } \frac{1}{x_1} + \frac{2}{x_2} + |x_3|$$

• $\frac{1}{x_1}, \frac{2}{x_2}, |x_3|$ are all convex

• so it is convex

$$12) - ①. \quad \max \{x_1 + x_2, x_1 - x_3\} \geq c$$

• $\max \{x_1 + x_2, x_1 - x_3\}$ is convex

• but the form is $\text{convex} \leq c$

• So not convex

This problem is not convex

$$(3). \text{ maximise } x_1 - x_2^2$$

• x_1 is linear, hence convex and concave

• x_2^2 is convex, hence $-x_2^2$ is concave

• $x_1 - x_2^2$ is concave

• maximise concave

$$(3) - ①. \quad (2x_1 - x_2)^2 \leq x_1$$

• $(2x_1 - x_2)^2 - x_1 \leq 0$

• $2x_1 - x_2$ is linear, hence convex

• $(2x_1 - x_2)^2$ is convex

• $(2x_1 - x_2)^2 - x_1$ is convex

• $\text{convex} \leq c$

This problem is convex

$$(4). \text{ maximise } 3x_1 - 2x_2 + 5^2$$

• $3x_1$ linear

• $2x_2$ linear

• $3x_1 - 2x_2 + 5^2$ is linear

• maximise concave

$$(4) - ①. \quad x_1 + x_2 \leq 2$$

• linear

• $\text{concave} \geq c$

$$(4) - ②. \quad x_2 \geq x_1$$

• $x_2 - x_1 \geq 0$

• linear

• $\text{concave} \geq c$

$$(4) - (3). \quad x_1, x_2 \geq 0$$

$$\cdot \text{concave} \geq c$$

This problem is convex

$$(5). \text{ minimise } x_1$$

- x_1 is linear
- maximise concave

$$(5) - (1). \quad x_1 \cdot x_2 \geq 2 \quad (x_2 \geq 4)$$

- Since we know x_2 is positive
- $x_1 \geq 2/x_2$
- $x_1 - 2/x_2 \geq 0$
- x_1 linear
- $\frac{2}{x_2}$ convex, hence $-\frac{2}{x_2}$ concave
- $x_1 - \frac{2}{x_2}$ concave
- $\text{concave} \geq c$

This problem is convex.

$$(6). \text{ minimise } x^T \Sigma x$$

- x^T linear
- Σx linear
- $x^T \Sigma x$ convex
- minimise convex

$$(6) - (1). \quad e^T x = 1$$

- $e^T x = 1$: linear
- $\mu^T x \geq \mu$: linear

This problem is convex

(7).