

**Prob. 1**

Define a point  $f$  in  $n$ -dimension space:  $f := X \rightarrow y$ , where  $y$  is constant  $R$  and  $X$  is a constant  $R^n$

Let  $x_1, x_2 \in X$  ( $x_1 = x_2$ )

$$f(x_1) = f(x_2) = y, \forall t \in [0,1],$$

$$\Rightarrow f(tx_1 + (1-t)x_2) = f(x_1) = f(x_2) = y, \quad tf(x_1) + (1-t)f(x_2) = ty + (1-t)y = y$$

$$\Rightarrow f(tx_1 + (1-t)x_2) = y \leq tf(x_1) + (1-t)f(x_2) = y$$

$\therefore$  a single point is convex

**Prob. 2**

Let  $f_1(x) = |2 - 5x|$ ,  $f_2(x) = 2x$ ,  $f_3(x) = 8e^{(-4x)}$ ,  $f_4(x) = -1$

$f_1 = |2 - 5x|$  is convex since it is a combination of hyperplane

$f_2 = 2x$  is convex since it is a hyperplane

$f_3 = 8e^{(-4x)}$  is convex since  $f_3'' = 128e^{(-4x)} > 0$

$f_4 = -1$  is convex since it is a hyperplane

Since  $f_1, f_2, f_3, f_4$  are convex and operation of sum of convex functions preserves convexity

$\Rightarrow f(x)$  is convex

**Prob. 3**

$$f'(x) = 0.5e^{(0.5x+2)} - 0.5e^{(-0.5x-0.5)} + 4 \Rightarrow f'(5) \approx 48.9837$$

By Symmetric Difference Quotient

$$h = 0.001, f \frown_{h=0.001}(5) - f'(5) = -1.874 \times 10^{-6}$$

$$h = 10^{-4}, f \frown_{h=10^{-4}}(5) - f'(5) = -1.856 \times 10^{-8}$$

$$h = 10^{-5}, f \frown_{h=10^{-5}}(5) - f'(5) = 1.688 \times 10^{-9}$$

$$h = 10^{-6}, f \frown_{h=10^{-6}}(5) - f'(5) = -5.658 \times 10^{-8}$$

Numerical results with  $h = 10^{-3}, 10^{-4}, 10^{-5}$  are smaller than the analytical result, except the result with  $h = 10^{-5}$  and it is also the closest result to the analytical one.

**Prob. 4**

$$\min_x [-3 \ 2] x - 1$$

$$f := \begin{bmatrix} 0 & -1 \\ 0 & -5 \\ -5 & 4 \end{bmatrix} x \leq \begin{bmatrix} 3 \\ -2 \\ -2 \end{bmatrix}$$

$$g := [1 \ 1] x = 2.3$$

**Prob. 5**

There would be more than one subgradient when  $3x^2 - 2, 2x - 1$  intersects

$$\Rightarrow (3x^2 - 2) - (2x - 1) = 0 \Rightarrow x = 1/3 \text{ or } 1$$

$$\delta f(x) = \begin{cases} 6x, & x > 1 \text{ or } x < 1/3 \\ 2, & 1/3 < x < 1 \\ [2, 6], & x = 1 \\ 2, & x = 1/3 \end{cases}$$

**Prob. 6**

(a) Lagrange dual function

$$\text{Lagrange function: } L(x, \lambda, \nu) = c^T x + \lambda^T (Gx - h) + \nu^T (Ax - b)$$

Dual function:

$$g(\lambda, \nu) = \inf_{x \in D} c^T x + \lambda^T (Gx - h) + \nu^T (Ax - b) = -h^T \lambda - b^T \nu + \inf_x (G^T \lambda + A^T \nu + c)^T x$$

$\lambda$  is a vector of Lagrange multipliers

$\nu$  is a vector of Lagrange multipliers

(b) Dual problem

$$\text{Maximize } -h^T \lambda - b^T \nu$$

$$\text{Subject to } G^T \lambda + A^T \nu + c = 0$$

$$\lambda \geq 0$$

(c) Optimal value of dual problem relation with primal optimal?

Since we assume primal is feasible and bounded, then the strong duality holds.

$$\Rightarrow p^* = d^*$$