

Problem 4.1.9

1. $f(x) = |x|$, taking $\bar{x} = 0$, we have the following subgradient inequality

$$\begin{aligned} \langle \lambda, x - \bar{x} \rangle &\leq f(x) - f(\bar{x}) \Rightarrow \langle \lambda, x \rangle \leq f(x) \\ \Rightarrow \lambda x &\leq |x| \quad \forall x \in \mathbb{R} \Rightarrow \lambda \in [-1, 1] \Rightarrow \partial f(x) = [-1, 1], x=0 \\ \text{for } x < 0 \text{ we have } \partial f(x) &= \{-1\} \\ \text{for } x > 0 \text{ we have } \partial f(x) &= \{1\} \end{aligned}$$

2. $f(x) = \delta_{[0, +\infty)}$. Taking $C = [0, +\infty)$, $x_1, x_2 \in C$ and $\lambda \in [0, 1]$ where $x_1 \leq x_2$

$$\Rightarrow \lambda x_1 + (1-\lambda)x_2 \leq \lambda x_2 + (1-\lambda)x_2 = x_2$$

$$\lambda x_1 + (1-\lambda)x_2 \geq \lambda x_1 + (1-\lambda)x_1 = x_1$$

$\Rightarrow [0, +\infty)$ is a closed, convex set in \mathbb{R}

$$\Rightarrow \partial f(x) = \partial \delta_{[0, +\infty)}(x) = N_{[0, +\infty)}(x)$$

3. $f(x) = \begin{cases} -\sqrt{x} & \text{if } x \in [0, +\infty) \\ +\infty & \text{if } x \notin [0, +\infty) \end{cases} = -\sqrt{x} + \delta_{[0, +\infty)}$

For $x \in (0, +\infty)$ and $g(x) = -\sqrt{x} \Rightarrow \partial g(x) = \{-\frac{1}{2\sqrt{x}}\}$.

For $\bar{x} = 0$, $g(x) = -\sqrt{x}$, we have

$$\langle \lambda, x - \bar{x} \rangle \leq f(x) - f(\bar{x}) \Rightarrow \lambda x \leq -\sqrt{x} \quad \forall x \in (0, +\infty)$$

$$\Rightarrow \lambda \leq -\frac{1}{\sqrt{x}} \quad \forall x \in (0, +\infty) \Rightarrow \lambda \in (-\infty, 0]$$

From 2. we have that $\partial \delta_{[0, +\infty)}(x) = N_{[0, +\infty)}(x)$

\therefore as we know $\partial(f+g)(x) = \partial f(x) + \partial g(x)$

\Rightarrow we conclude that

$$\partial f(x \neq 0) = \left\{ -\frac{1}{2\sqrt{x}}, N_{[0, +\infty)}(x) \right\}$$

$$\partial f(x=0) = \left\{ \begin{matrix} (-\infty, 0] \\ N_{[0, +\infty)}(x) \end{matrix} \right\}$$

Problem 4.1.14

1. $f(x) = |x|$

$$f^*(y) = \sup_x \{xy - |x|\}$$

$$\text{if } y > 1 \quad \sup_x \{xy - |x|\} = +\infty$$

$$\text{if } -1 \leq y \leq 1 \quad \sup_x \{xy - |x|\} = 0$$

$$\text{if } y < -1 \quad \sup_x \{xy - |x|\} = +\infty$$

$$\Rightarrow f^*(y) = \begin{cases} 0, & |y| \leq 1 \\ +\infty, & |y| > 1 \end{cases}$$

2. $f(x) = \delta_{[0, +\infty)} = \begin{cases} 0, & x \in [0, +\infty) \\ +\infty, & x \notin [0, +\infty) \end{cases}$

We know from Problem 4.1.9 that the interval $[0, +\infty)$ is a closed, convex set. \Rightarrow we understand that the conjugate of the support function of $[0, +\infty)$ is the indicator function $\delta_{[0, +\infty)}$. Given that $S(C, \cdot)(x)$ is lower-semicontinuous, we have that $S^{**}(C, \cdot)(x) = S(C, \cdot)(x)$ by the biconjugate theorem. $\Rightarrow f^*(y) = S(C, \cdot)(y)$ where $C = [0, +\infty)$

Problem 4.1.14

$$3. f(\lambda) = \begin{cases} -\sqrt{\lambda} & \text{if } \lambda \in [0, +\infty) \\ +\infty & \text{if } \lambda \notin [0, +\infty) \end{cases}$$

$$\text{let } g(\lambda) = -\sqrt{\lambda} \Rightarrow g^*(y) = \sup_{\lambda \in [0, +\infty)} \{\lambda y + \sqrt{\lambda}\}$$

$$\text{for } y \geq 0 \text{ we have } g^*(y) = +\infty$$

$$\text{Now looking at } \frac{d}{d\lambda} \{\lambda y + \sqrt{\lambda}\} = y + \frac{1}{2\sqrt{\lambda}} = 0 \Rightarrow \lambda = \frac{1}{4y^2} \quad y < 0$$

for $y < 0$ we have

$$g^*(y) = \frac{y}{4y^2} + \sqrt{\frac{1}{4y^2}} = \frac{1}{4y} - \frac{1}{2y} = -\frac{1}{4y}$$

$$\therefore g^*(y) = \begin{cases} -\frac{1}{4y}, & y < 0 \\ +\infty, & y \geq 0 \end{cases}$$

$$\Rightarrow f^*(y) = \begin{cases} -\frac{1}{4y}, & y < 0 \\ +\infty, & y \geq 0 \end{cases}$$

Problem 4.1.19

$q_A(x) = \frac{1}{2} x^T A x$ where A is real, positive definite square matrix

1. By lemma 4.1.8, we have that since $A \in P(n)$

$$\nabla^2 q_A(x) = A, \text{ where } A \text{ is a real, positive definite square matrix}$$

Given this fact about A , by Theorem 1.1.6 iii) as we have the hessian of $q_A(x)$ is positive definite for all $x \in X$, we have that q_A is strictly convex.

2. By lemma 4.1.18, we have that

$$\nabla q_A(x) = Ax, \text{ where } A \text{ is a real, positive definite square matrix}$$

$$\Rightarrow q_A^*(y) = \sup_x \langle xy - \frac{1}{2} x^T A x \rangle. \text{ let } f: x \mapsto xy - \frac{1}{2} x^T A x$$

$$\nabla f(x) = y - Ax = 0 \Rightarrow x = A^{-1}y$$

$$\text{Plug back into } f \Rightarrow y^T A^{-1}y - \frac{1}{2} (A^{-1}y)^T A (A^{-1}y)$$

$$= y^T A^{-1}y - \frac{1}{2} y^T A^{-1} A A^{-1}y = y^T A^{-1}y - \frac{1}{2} y^T A^{-1}y = \frac{1}{2} y^T A^{-1}y$$

Problem 4.1.19

3. We have that $q_A(\lambda) = \frac{1}{2} \lambda^T A \lambda$ is strictly convex and that $q_A^*(y) = \frac{1}{2} y^T A^{-1} y = q_{A^{-1}}(y)$

We know that if $A \preceq B \Leftrightarrow \lambda^T A \lambda \geq \lambda^T B \lambda \quad \forall \lambda$

$$\Rightarrow q_A(\lambda) = \frac{1}{2} \lambda^T A \lambda \geq \frac{1}{2} \lambda^T B \lambda = q_B(\lambda) \quad \forall \lambda$$

$$\Rightarrow q_A^*(y) = \frac{1}{2} y^T A^{-1} y \leq \frac{1}{2} y^T B^{-1} y = q_{B^{-1}}^*(y) \quad \forall y \quad \text{By lemma 4.1.12}$$

$$\Rightarrow A^{-1} \preceq B^{-1}$$

Now given $B^{-1} \preceq A^{-1} \Leftrightarrow y^T B^{-1} y \geq y^T A^{-1} y \quad \forall y$

$$\Rightarrow q_{B^{-1}}(y) = \frac{1}{2} y^T B^{-1} y \geq \frac{1}{2} y^T A^{-1} y = q_{A^{-1}}(y) \quad \forall y$$

$$\Rightarrow q_B(\lambda) = \frac{1}{2} \lambda^T B \lambda \leq \frac{1}{2} \lambda^T A \lambda = q_A(\lambda) \quad \forall \lambda \quad \text{By lemma 4.1.12}$$

$$\Rightarrow B \preceq A$$

□