

# Homework requirements

- Digital format (can be typeset or photos) is preferred
- Submit by next lecture
- Each homework 10 points; 1 point deducted for each day of delay

# Contact information

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# Homework (3)

1. Compute the subgradients of  $f(\mathbf{x}) = \frac{1}{2}x_1^2 + |x_2|$ ,  $\|\mathbf{x}\|_2$ ,  $\|\mathbf{x}\|_\infty$ , and  $\|\mathbf{X}\|_{2,1}$ .
2. Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  be a convex function. Show that  $\partial f : \mathbb{R}^n \rightarrow \mathbb{R}^n$  is a monotone mapping, i.e.,

$$\langle \mathbf{g}_1 - \mathbf{g}_2, \mathbf{x}_1 - \mathbf{x}_2 \rangle \geq 0, \quad \forall \mathbf{g}_i \in \partial f(\mathbf{x}_i), i = 1, 2. \quad (2)$$

Further, if  $f$  is  $\mu$ -strongly convex, then the above inequality can be strengthened as

$$\langle \mathbf{g}_1 - \mathbf{g}_2, \mathbf{x}_1 - \mathbf{x}_2 \rangle \geq \mu \|\mathbf{x}_1 - \mathbf{x}_2\|^2, \quad \forall \mathbf{g}_i \in \partial f(\mathbf{x}_i), i = 1, 2. \quad (3)$$

# Homework (3)

3. Show that  $f(\mathbf{x}) = \sum_{i=1}^r \alpha_i x_{[i]}$  is a convex function of  $\mathbf{x}$ , where  $\alpha_1 \geq \alpha_2 \geq \dots \geq \alpha_r \geq 0$ , and  $x_{[i]}$  denotes the  $i$ th largest component of  $\mathbf{x}$ .
4. Show that  $f(\mathbf{x}) = -\log \left( -\log \left( \sum_{i=1}^m e^{\mathbf{a}_i^T \mathbf{x} + b_i} \right) \right)$  is convex on  $\text{dom } f = \{\mathbf{x} \mid \sum_{i=1}^m e^{\mathbf{a}_i^T \mathbf{x} + b_i} < 1\}$ .
5. Prove that  $f_k(\mathbf{X}) = \sum_{i=1}^k \sigma_i(\mathbf{X})$  is a convex function of  $\mathbf{X}$  for all  $k = 1, \dots, \text{rank}(\mathbf{X})$ .
6. Show that the Minkowski function:

$$M_C(\mathbf{x}) = \inf\{t > 0 \mid t^{-1}\mathbf{x} \in C\},$$

is convex, where  $C$  is a convex set. Find its conjugate function.

# Homework (9)

7. Derive the conjugates of the following functions.

- (a) Sum of largest elements.  $f(\mathbf{x}) = \sum_{i=1}^r x_{[i]}$  on  $\mathbb{R}^n$ .
- (b) Piecewise-linear function on  $\mathbb{R}$ .  $f(x) = \max_{i=1,\dots,n} (a_i x + b_i)$  on  $\mathbb{R}$ . *You can assume that the  $a_i$  are sorted in increasing order, i.e.,  $a_1 \leq \dots \leq a_m$ , and that none of the functions  $a_i x + b_i$  is redundant, i.e., for each  $k$  there is at least one  $x$  with  $f(x) = a_k x + b_k$ .*
- (c) Power function.  $f(x) = x^p$  on  $\mathbb{R}_{++}$ , where  $p > 1$ . Repeat for  $p < 0$ .
- (d) Geometric mean.  $f(\mathbf{x}) = -(\prod x_i)^{1/n}$  on  $\mathbb{R}_{++}^n$ .
- (e) Negative generalized logarithm for second-order cone.  $f(\mathbf{x}, t) = -\log(t^2 - \mathbf{x}^T \mathbf{x})$  on  $\{(\mathbf{x}, t) \in \mathbb{R}^n \times \mathbb{R} \mid \|\mathbf{x}\|_2 < t\}$ .

**Note:** Computing a conjugate function needs to specify its domain as well.

# Homework (3)

8. Find the proximal mappings of the following functions.

(a)  $\|\mathbf{x}\|_2$ .

(b)  $\|\mathbf{x}\|_\infty$ .

(c)  $\|\mathbf{X}\|_2$ .

9. Find the projection onto the second order cone.

10. Let  $d(\mathbf{x})$  be the distance from  $\mathbf{x}$  to a closed convex set  $\mathcal{C}$ :  $d(\mathbf{x}) = \inf_{\mathbf{y} \in \mathcal{C}} \|\mathbf{x} - \mathbf{y}\|_2$ .

Find the proximal mappings of  $d(\mathbf{x})$  and  $\frac{1}{2}d^2(\mathbf{x})$ .

11. Prove that:

(a) If  $f(\mathbf{x}) = g(\mathbf{x}) + \mathbf{a}^T \mathbf{x}$ , then  $P_c f(\mathbf{x}) = P_c g(\mathbf{x} - c\mathbf{a})$ .

(b) If  $f(\mathbf{x}) = g(\mathbf{x}) + \frac{1}{2\mu} \|\mathbf{x} - \mathbf{a}\|^2$ , then  $P_c f(\mathbf{x}) = P_\lambda g(\lambda(c^{-1}\mathbf{x} + \mu^{-1}\mathbf{a}))$ , where  $\lambda^{-1} = \mu^{-1} + c^{-1}$ .