Convex Optimization for Machine Learning and Computer Vision

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Weekly Exercises 5

Room: 02.09.023 Wednesday, 27.11.2019, 12:15-14:00

Submission deadline: Monday, 25.11.2019, 16:15, Room 02.09.023

Convex conjugate and prox

(8+6 Points)

Exercise 1 (4 points). Given a $X \in \mathbb{R}^{m \times n}$, compute the subdifferential of the 1,2-norm, i.e.

$$\partial ||X||_{1,2} = \partial (\sum_{i=1}^{m} (\sum_{j=1}^{n} X_{i,j}^{2})^{1/2}).$$

Exercise 2 (4 Points). Consider following problems of convex conjugate:

• Let $f: \mathbb{R}^n \to \mathbb{R}$ be convex. Show that the convex conjugate of the perspective function $g: \mathbb{R}^n \times \mathbb{R} \to \mathbb{R} \cup \{\infty\}$

$$g(x,t) = \begin{cases} tf(x/t), & \text{if } t > 0\\ \infty, & \text{otherwise} \end{cases}$$

is given by

$$g^*(y,s) = \begin{cases} 0, & \text{if } f^*(y) \le -s \\ \infty, & \text{otherwise} \end{cases}$$

• Show that the biconjugate of the persepective function g is given by

$$g^{**}(x,t) = \begin{cases} tf(x/t), & \text{if } t > 0\\ \sigma_{\text{dom}(f^*)}(x), & \text{if } t = 0\\ \infty, & \text{if } t < 0 \end{cases}$$

where $\sigma_{\text{dom}(f^*)}(x) = \sup_{y \in \text{dom}(f^*)} \langle x, y \rangle$ is the support function of dom (f^*) .

Exercise 3 (6 Points). Let $A \in \mathbb{R}^{m \times n}$ be a linear operator and $f : \mathbb{R}^n \to \mathbb{R} \cup \{\infty\}$ a convex function. Then $Af : \mathbb{R}^m \to \mathbb{R} \cup \{\infty\}$ defined as

$$(Af)(u) := \begin{cases} \inf_{v \in \mathbb{R}^n, Av = u} f(v) & \text{if } \exists v \in \mathbb{R}^n \text{ s.t. } Av = u \\ \infty & \text{otherwise.} \end{cases}$$

is called the image of f under A.

- 1. Show that the convex conjugate $(Af)^*$ of Af is given as $f^* \circ A^{\top}$ where $(f^* \circ A^{\top})(v) := f^*(A^{\top}v)$.
- 2. Name the properties that we require for $A^{\top}f^* = (f \circ A)^*$ to hold. What theorem from the lecture applies here?
- 3. Give an example of a closed, convex and non-empty set C and a linear operator A s.t. $AC := \{Ax : x \in C\}$ is not closed.
- 4. Let f be closed, (convex) and proper. Argue that Af does not need to be closed.