# Homework 4: 19 March 2019 CS-526 Learning Theory

## Exercise 1 (from S. Boyd & J. Duchi)

For the following convex functions, explain how to calculate a subgradient at a given  $\mathbf{x}$ .

- 1.  $\forall \mathbf{x} \in \mathbb{R}^n : f(x) = \max_{1 \le i \le m} (\mathbf{a}_i^T \mathbf{x} + b_i)$ , where  $\forall i \in \{1, \dots, m\} : (\mathbf{a}_i, b_i) \in \mathbb{R}^n \times \mathbb{R}$ .
- 2.  $\forall \mathbf{x} \in \mathbb{R}^n : f(x) = \max_{1 \le i \le m} |\mathbf{a}_i^T \mathbf{x} + b_i|.$
- 3.  $\forall \mathbf{x} \in \mathbb{R}^n : f(\mathbf{x}) = \sup_{t \in [0,1]} p(t, \mathbf{x}), \text{ where } p(t) = x_1 + x_2 t + x_n t^{n-1}.$

## Exercise 2 (from S. Boyd & J. Duchi)

Convex functions that are not subdifferentiable. Verify that the following functions, defined on the interval  $[0, +\infty)$ , are convex, but not subdifferentiable at x = 0.

- 1. f(0) = 1, and f(x) = 0 for x > 0
- 2.  $f(x) = -\sqrt{x}$

#### Exercise 3

Theorem 14.11 in the textbook is a refined bound for Stochastic Gradient Descent (SGD) when the function f is strongly convex. The proof relies on Claim 14.10 which generalizes Lemma 13.5, i.e.,

Claim 1 If f is  $\lambda$ -strongly convex then for every  $\mathbf{w}$ ,  $\mathbf{u}$  and  $\mathbf{v} \in \partial f(\mathbf{w})$  we have

$$\langle \mathbf{w} - \mathbf{u}, \mathbf{v} \rangle \ge f(\mathbf{w}) - f(\mathbf{u}) + \frac{\lambda}{2} ||\mathbf{w} - \mathbf{u}||^2$$

Prove this claim.

#### Exercise 4

Let  $\pi_{\mathcal{C}}(\mathbf{x}) = \arg\min_{y \in \mathcal{C}} \|\mathbf{x} - \mathbf{y}\|^2$  denote the Euclidean projection of x onto a closed convex set  $\mathcal{C}$  of a Hilbert space H. Show that the projection is a 1-Lipschitz mapping, that is,

$$\|\pi_{\mathcal{C}}(\mathbf{x}_0) - \pi_{\mathcal{C}}(\mathbf{x}_1)\| \le \|\mathbf{x}_0 - \mathbf{x}_1\|,$$

for all vectors  $\mathbf{x}_0, \mathbf{x}_1 \in H$ . Show that the Lipschitz constant cannot be improved. *Hint:* First prove the following important property of the projection onto a closed convex.

**Lemma 1** If C is a non-empty closed convex subset of a Hilbert space H then

$$\forall (\mathbf{x}, \mathbf{y}) \in H \times C : \langle \mathbf{x} - \pi_C(\mathbf{x}), \mathbf{y} - \pi_C(\mathbf{x}) \rangle < 0.$$