

# Homework 5

## Convex Optimization

Due April 16

1. Consider the following non-smooth function on the real line:

$$f(x) = \max((x+1)^2, (x-3)^2)$$

Describe the subdifferential  $\partial f(x)$  at every point  $x \in \mathbb{R}$ .

2. Prove or disprove: the subdifferential  $\partial f(x)$  of a convex function is a convex set at every  $x \in \mathbb{R}$ .

3. Recall the subdifferential for the nuclear norm for an  $n_1 \times n_2$  matrix  $\mathbf{X}$ .

$$\partial \|\mathbf{X}\|_* = \{\mathbf{U}\mathbf{V}^T + \mathbf{W} : \mathbf{U}\mathbf{W} = \mathbf{0}, \mathbf{W}\mathbf{V} = \mathbf{0}, \|\mathbf{W}\|_* \leq 1\}$$

In the expression above,  $\mathbf{X}$  has rank  $r$  and its SVD is  $\mathbf{X} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$ , where  $\mathbf{U}$  is  $n_1 \times r$ ,  $\mathbf{\Sigma}$  is  $r \times r$  and  $\mathbf{V}$  is  $n_2 \times r$ . Recall that

$$\begin{aligned} \mathbf{x}^+ &= \text{prox}_{t\|\cdot\|_*}(\mathbf{X}) \\ &= \arg \min_{\mathbf{Z}} \left( \|\mathbf{Z}\|_* + \frac{1}{2t} \|\mathbf{Z} - \mathbf{X}\|_F^2 \right) \end{aligned}$$

if and only if

$$\mathbf{X} - \mathbf{X}^+ \in t\partial \|\mathbf{X}^+\|_*$$

Show that we can compute the prox operator above by singular value thresholding:

$$\mathbf{X}^+ = \mathbf{U}\mathbf{\Sigma}^+\mathbf{V}^T, \text{ where } \mathbf{\Sigma}^+[i, i] = \max(\sigma_i t, 0).$$