

ECE 8823 (Convex Optimization), Spring 2017

Non-smooth Optimization Quiz

Friday April 21

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 $\partial f(\mathbf{x})$ of a convex function is a convex set at every $\mathbf{x} \in \mathbb{R}^N$.

3. Recall the subdifferential for the nuclear norm for an $N_1 \times N_2$ matrix \mathbf{X} .

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

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:

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