

## Homework 5

### Requirements:

1. Digital format (can be typeset or photos, ought to write clearly if written by hand), upload to <https://course.pku.edu.cn/>.
2. Submit by next class
3. A problem is not counted if nobody can work it out
4. Each homework 10 points; 1 point deducted for each week's delay

### Problems:

1. Use the trick of checking the convexity of 1D sections to prove that
  - (a)  $f(\mathbf{X}) = \text{tr}(\mathbf{X}^{-1})$  is convex on  $\text{dom } f = \mathbb{S}_{++}^n$ .
  - (b)  $f(\mathbf{X}) = (\det \mathbf{X})^{1/n}$  is concave on  $\text{dom } f = \mathbb{S}_{++}^n$ .
2. For strictly convex and differentiable  $f$ , prove the identity:

$$\langle \nabla f(\mathbf{x}) - \nabla f(\mathbf{y}), \mathbf{x} - \mathbf{z} \rangle = B_f(\mathbf{z}, \mathbf{x}) + B_f(\mathbf{x}, \mathbf{y}) - B_f(\mathbf{z}, \mathbf{y}).$$

3. Show that  $B_f(\mathbf{x}, \mathbf{y})$  is convex in  $\mathbf{x}$  but not necessarily in  $\mathbf{y}$ .
4. Let

$$\phi(\mathbf{x}) = \frac{\beta \|\mathcal{A}\|_2^2}{2} \|\mathbf{x} - \mathbf{u}\|^2 - \frac{\beta}{2} \|\mathcal{A}\mathbf{x} - \mathbf{v}\|^2,$$

where  $\mathbf{u}$  and  $\mathbf{v}$  are constant vectors. Show that  $B_\phi(\mathbf{x}, \mathbf{x}')$  actually does not depend on  $\mathbf{u}$  and  $\mathbf{v}$ .

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

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

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

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

7. 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  $\mathbf{x}$ .
8. Prove that  $f_k(\mathbf{X}) = \sum_{i=1}^k \sigma_i(\mathbf{X})$  is a convex function of  $\mathbf{X}$  for all  $k = 1, \dots, \text{rank}(\mathbf{X})$ .