

**AMS 229: Convex Optimization**  
Fall 2018

Midterm Exam

Name: \_\_\_\_\_ Student ID: \_\_\_\_\_

For this exam you only need a pen/pencil. No electronic device is allowed. Write your answers in the spaces provided. If you need more space, work on the other side of the page.

**Problem 1** (10 + 5 + 5 = 20 points)

A real  $n \times n$  matrix  $X$  is called doubly stochastic if its elements are non-negative, and all row sums and column sums equal unity, that is,

$$X_{ij} \geq 0, \quad \sum_{i=1}^n X_{ij} = \sum_{j=1}^n X_{ij} = 1, \quad \text{for all } i, j = 1, \dots, n.$$

In this problem, we investigate the set  $\mathcal{X} = \{X \in \mathbb{R}^{n \times n} \mid X \text{ is doubly stochastic}\}$ .

(a) Prove that  $\mathcal{X}$  is a convex set.

(b) Is  $\mathcal{X}$  a cone? Why/why not?

(c) Is  $\mathcal{X}$  a polytope (also known as polyhedron)? Why/why not?

**Problem 2**

(5 + 5 + 5 + 5 = 20 points)

Consider the following optimization problem, which we call problem (P). Here,  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\mathbf{b} \in \mathbb{R}^m$ ,  $\mathbf{c} \in \mathbb{R}^n$  are given.

$$\begin{aligned} & \underset{\mathbf{x} \in \mathbb{R}^n}{\text{minimize}} && \mathbf{c}^\top \mathbf{x} \\ & \text{subject to} && \mathbf{Ax} \preceq \mathbf{b}, \\ & && x_i \in \{0, 1\}, \quad i = 1, \dots, n. \end{aligned}$$

(a) Is problem (P) a convex optimization problem? Why/Why not?

(b) Consider the following modification of problem (P), which we call problem (Q).

$$\begin{aligned} & \underset{\mathbf{x} \in \mathbb{R}^n}{\text{minimize}} && \mathbf{c}^\top \mathbf{x} \\ & \text{subject to} && \mathbf{Ax} \preceq \mathbf{b}, \\ & && 0 \leq x_i \leq 1, \quad i = 1, \dots, n. \end{aligned}$$

Is problem (Q) convex? If “yes”, then what kind of convex optimization

problem is it? If “no”, then explain why.

- (c) Suppose the optimal value for problem (P) is  $p_{\text{opt}}$ , and the same for problem (Q) is  $q_{\text{opt}}$ . Write an inequality between  $p_{\text{opt}}$  and  $q_{\text{opt}}$ . Justify your answer.

- (d) Suppose we do not know how to solve problem (P). Instead, we solve problem (Q), and suppose we find that its minimizer  $\mathbf{x}_{\text{opt}}^q \in \{0, 1\}$ . Can we then conclude that  $p_{\text{opt}} = q_{\text{opt}} = \mathbf{c}^\top \mathbf{x}_{\text{opt}}^q$ , that is, problem (Q) solves problem (P)? Explain.

**Problem 3**

(5 × 2 = 10 points)

For each the following statements, ONLY ONE among the three options are correct. Choose the correct option for each. You DO NOT need to provide any explanation.

- (a) For a non-convex function  $f$ , its tri-conjugate  $f^{***}$  satisfies
- (i)  $f^{***} = f$ , the original non-convex function
  - (ii)  $f^{***} = f^*$ , which is a convex function
  - (iii)  $f^{***} = f^{**}$ , which is a convex function
- (b) At each point on the boundary of a compact convex set, a supporting hyperplane
- (i) may not exist
  - (ii) exists
  - (iii) exists and is unique
- (c) The negative entropy function  $f(\mathbf{x})$  given by
- $$f(\mathbf{x}) = \sum_{i=1}^n x_i \log x_i, \quad \text{dom}(f) = \{\mathbf{x} \in \mathbb{R}_{>0}^n \mid \mathbf{1}^\top \mathbf{x} = 1\},$$
- (i) is quasiconcave
  - (ii) is quasiconvex but not convex
  - (iii) both quasiconvex and convex
- (d) A function is concave if and only if its hypograph is
- (i) convex
  - (ii) open
  - (iii) closed
- (e) The dual cone of  $\mathbb{S}_+^n$  equals
- (i)  $\mathbb{S}_+^n$
  - (ii)  $\mathbb{S}_-^n$
  - (iii)  $\mathbb{S}^n$

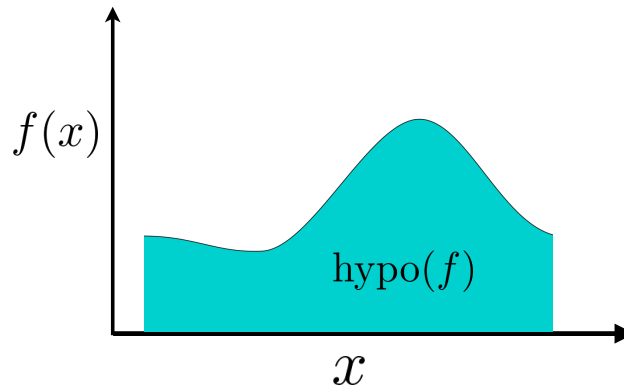
# Some useful information

- Convex conjugate or Legendre-Fenchel transform of function  $f(\mathbf{x})$  is

$$f^*(\mathbf{y}) = \sup_{\mathbf{x} \in \text{dom}(f)} (\mathbf{y}^\top \mathbf{x} - f(\mathbf{x})) .$$

$f^*$  is convex even if  $f$  is not.

- Suppose  $f$  is a twice differentiable function. Then  $f$  is convex (resp. concave) if and only if  $\text{dom}(f)$  is convex, and Hessian of  $f$  is positive (resp. negative) semidefinite everywhere in  $\text{dom}(f)$ .
- Hypograph of a function  $f$  is the set of points lying on or below its graph.



- Given a cone  $\mathcal{K}$ , its dual cone  $\mathcal{K}^*$  is given by

$$\mathcal{K}^* = \{y \mid \langle y, x \rangle \geq 0, \text{ for all } x \in \mathcal{K}\} .$$

$\mathcal{K}^*$  is convex even if  $\mathcal{K}$  is not.