

# MGMT 690 - Pset 1

Spring 2024

## Instructions:

- This pset is due on Sunday, March 24 at 11:59pm.
- Completed psets should be submitted to Gradescope.
- **Exercises** are for your own review only. They do not need to be submitted and will not be graded.
- **Complete all problems 1–3 and one of either 4 or 5.**

## Exercises

1. Let  $V$  be a Euclidean space and let

$$\|v\| := \sqrt{\langle v, v \rangle}.$$

Prove that this is a norm.

2. Let  $p \in \mathbb{R}$ ,  $p > 0$ . For  $x \in \mathbb{R}^n$ , define

$$\|x\|_p := \left( \sum_{i=1}^n |x_i|^p \right)^{1/p}.$$

Prove that this is *not* a norm for  $p \in (0, 1)$  and  $n \geq 2$ .

3. Prove that the affine image of a convex set is a convex set.
4. Let  $C \subseteq \mathbb{R}^n$  be a convex set. Let  $x \in \text{rint}(C)$  and  $y \in \text{cl}(C)$ . Prove that for all  $\theta \in [0, 1)$ , that  $(1 - \theta)x + \theta y \in \text{rint}(C)$ .

## Problems

1. [25 pts] Given  $A \in \mathbb{S}^n$  and  $B \in \mathbb{S}^m$ , the Kronecker product  $A \otimes B$  is the  $\mathbb{S}^{mn}$  matrix given in block form as

$$A \otimes B = \begin{pmatrix} A_{1,1}B & \dots & A_{1,n}B \\ \vdots & \ddots & \vdots \\ A_{n,1}B & \dots & A_{n,n}B \end{pmatrix}$$

Suppose  $A \in \mathbb{S}_+^n$  and  $B \in \mathbb{S}_+^m$ . Show that  $A \otimes B \succeq 0$ .

2. [25 pts] Given a symmetric matrix  $A \in \mathbb{S}^n$ , let  $\text{Inertia}(A) := (n_-, n_0, n_+)$  denote the number of negative eigenvalues, number of zero eigenvalues, and number of positive eigenvalues of  $A$ . Prove that for any invertible  $P \in \mathbb{R}^{n \times n}$ , that

$$\text{Inertia}(A) = \text{Inertia}(P^T A P).$$

3. [25 pts] Prove that
- [5pts] the nonnegative orthant is self-dual,
  - [10pts] the second-order cone is self-dual, and
  - [10pts] the semidefinite cone is self-dual.
4. [25 pts] In sparse recovery, the goal is to recover a sparse vector  $x^* \in \mathbb{R}^n$  given linear measurements  $(A, b) \in \mathbb{R}^{m \times n} \times \mathbb{R}^m$  where  $b = Ax^*$ . A convex-optimization approach to this problem is to output the optimizer of

$$\min_{x \in \mathbb{R}^n} \{\|x\|_1 : Ax = b\}.$$

This problem gives a necessary and sufficient condition for when this convex-optimization approach correctly recovers  $x^*$ .

We say that a vector is  $k$ -sparse if it has at most  $k$  nonzero entries. Given a subset  $S \subseteq [n]$  and a vector  $x \in \mathbb{R}^n$ , let  $x_S$  denote the restriction of  $x$  onto the set  $S$ . Let  $S^c$  denote the complement of  $S$ . For a vector  $x \in \mathbb{R}^n$ , let  $\text{sign}(x)$  denote the  $\{-1, 0, 1\}$ -valued vector giving the individual signs of the coordinates of  $x$ .

- (a) [10pts] The *descent cone* of a convex-optimization problem at a feasible solution  $\bar{x}$  is defined as

$$\left\{ \delta \in \mathbb{R}^n : \begin{array}{l} \forall \epsilon > 0 \text{ small enough :} \\ \bar{x} + \epsilon \delta \text{ is feasible} \\ \text{obj. value at } \bar{x} + \epsilon \delta \leq \text{obj. value at } \bar{x} \end{array} \right\}$$

Show that for this problem, the descent cone at the optimal solution  $x^*$  is

$$\left\{ \delta \in \mathbb{R}^n : \begin{array}{l} \delta \in \ker(A) \\ \langle \text{sign}(x^*), \delta_{S^*} \rangle + \|\delta_{(S^*)^c}\|_1 \leq 0 \end{array} \right\}$$

where  $S^*$  is the support of  $x^*$ .

- (b) [10pts] The matrix  $A$  is said to satisfy the *nullspace property at order  $k$*  if for all sets  $S \subseteq [n]$  with  $|S| \leq k$  and for all  $\delta \in \ker(A) \setminus \{0\}$ , we have

$$\|\delta_S\|_1 < \|\delta_{S^c}\|_1.$$

Show that the descent cone at  $x^*$  is trivial, i.e., equal to  $\{0\}$ , if  $A$  satisfies the nullspace property at order  $k$  and  $x^*$  is  $k$ -sparse.

- (c) [5pts] Show that if  $A$  does not satisfy the nullspace property at order  $k$ , then there exists a  $k$ -sparse  $x^*$  for which the convex-optimization approach may fail to recover  $x^*$ . That is, for which the descent cone at  $x^*$  is nontrivial.
5. [25 pts] Given a permutation  $\sigma$  of  $[n]$ , we can associate  $\sigma$  with the  $n \times n$  permutation matrix

$$(X^\sigma)_{i,j} = \begin{cases} 1 & \text{if } \sigma(i) = j \\ 0 & \text{else} \end{cases}.$$

Let  $\mathcal{P}_n$  denote the set of  $n!$  permutation matrices of size  $n \times n$ . Prove that  $\text{conv}(\mathcal{P}_n) = \text{DS}_n$ , the set of doubly stochastic matrices:

$$\text{DS}_n := \left\{ X \in \mathbb{R}^{n \times n} : \begin{array}{l} X \geq 0 \\ X^\top \mathbf{1}_n = \mathbf{1}_n \\ X \mathbf{1}_n = \mathbf{1}_n \end{array} \right\}.$$

*Hint:* Use Hall's marriage theorem to prove that the support of any doubly stochastic matrix contains a permutation matrix.