

Problem Set 2

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1. Show the dual norm: $\|\mathbf{y}\|_{\infty}^* = \sup_{\|\mathbf{x}\|_{\infty} \leq 1} \mathbf{y}^{\top} \mathbf{x} = \|\mathbf{y}\|_1$.
2. Let $f(\mathbf{x}) := \frac{1}{2} \mathbf{x}^{\top} \mathbf{M} \mathbf{x}$ with \mathbf{M} symmetric, then show that $\nabla f(\mathbf{x}) = \mathbf{M} \mathbf{x}$, $\mathbf{H}_f(\mathbf{x}) = \mathbf{M}$.
3. Show if $f_i(\mathbf{x})$ is convex, then so is $g(\mathbf{x}) := \max_i f_i(\mathbf{x})$.
4. **(EX 3.2) (Optimality conditions)** Consider the problem of minimizing $\mathbf{c}^{\top} \mathbf{x}$ over a polyhedron P . Prove the following:
 - (a) A feasible solution \mathbf{x} is optimal if and only if $\mathbf{c}^{\top} \mathbf{d} \geq 0$ for every feasible direction \mathbf{d} at \mathbf{x} .
 - (b) A feasible solution \mathbf{x} is the unique optimal if and only if $\mathbf{c}^{\top} \mathbf{d} > 0$ for every nonzero feasible direction \mathbf{d} at \mathbf{x} .
5. **(EX 3.7) (Optimality conditions)** Consider a feasible solution \mathbf{x} to a standard form problem, and let $Z = \{i \mid x_i = 0\}$. Show that \mathbf{x} is an optimal solution if and only if the linear programming problem

$$\begin{aligned}
 \text{Min} \quad & \mathbf{c}^{\top} \mathbf{d} \\
 \text{s.t.} \quad & \mathbf{A} \mathbf{d} = \mathbf{0} \\
 & d_i \geq 0, \quad i \in Z
 \end{aligned}$$

has an optimal cost of zero. (In this sense, deciding optimality is equivalent to solving a new linear programming problem.)

6. **(EX 3.4)** Consider the problem of the standard form polyhedron $P = \{\mathbf{x} \in \Re^n \mid \mathbf{A} \mathbf{x} = \mathbf{b}, \mathbf{D} \mathbf{x} \leq \mathbf{f}, \mathbf{E} \mathbf{x} \leq \mathbf{g}\}$. Let \mathbf{x}^* be an element of P that satisfies $\mathbf{D} \mathbf{x}^* = \mathbf{f}, \mathbf{E} \mathbf{x}^* < \mathbf{g}$. Show that the set of feasible directions at the point \mathbf{x}^* is the set

$$\{\mathbf{d} \in \Re^n \mid \mathbf{A} \mathbf{d} = \mathbf{0}, \mathbf{D} \mathbf{d} \leq \mathbf{0}\}.$$

7. **(EX 4.6) (Duality in Chebychev approximation)** Let \mathbf{A} be a $m \times n$ matrix and let \mathbf{b} be a vector in \Re^m . We consider the problem of minimizing $\|\mathbf{A} \mathbf{x} - \mathbf{b}\|_{\infty}$ over all $\mathbf{x} \in \Re^n$. Here $\|\cdot\|_{\infty}$ is the vector norm defined by $\|\mathbf{y}\|_{\infty} = \max_i |y_i|$. Let v be the value of the optimal cost.

- (a) Let \mathbf{p} be any vector in \Re^m that satisfies $\sum_{i=1}^m |p_i| = 1$ and $\mathbf{p}^\top \mathbf{A} = \mathbf{0}$. Show that $\mathbf{p}^\top \mathbf{b} \leq v$.
- (b) In order to obtain the best possible lower bound of the form consider in part (a), we form the linear programming problem

$$\begin{aligned} \text{Max} \quad & \mathbf{p}^\top \mathbf{b} \\ \text{s.t.} \quad & \mathbf{p}^\top \mathbf{A} = \mathbf{0} \\ & \sum_{i=1}^m |p_i| \leq 1. \end{aligned} \tag{1}$$

Show that the optimal cost in this problem is equal to v .

8. **(EX 4.7) (Duality in piecewise linear convex optimization)** Consider the problem of minimizing $\max_{i=1, \dots, m} (\mathbf{a}_i^\top \mathbf{x} - b_i)$ over all $\mathbf{x} \in \Re^n$. Let v be the value of optimal cost, assumed finite. Let \mathbf{A} be the matrix with rows $\mathbf{a}_1, \dots, \mathbf{a}_m$, and let \mathbf{b} be the vector with components b_1, \dots, b_m .

- (a) Consider any vector $\mathbf{p} \in \Re^m$ that satisfies $\mathbf{p}^\top \mathbf{A} = \mathbf{0}$, $\mathbf{p} \geq \mathbf{0}$, and $\sum_{i=1}^m p_i = 1$. Show that $-\mathbf{p}^\top \mathbf{b} \leq v$.
- (b) In order to obtain the best possible lower bound of the form considered in part (a), we form the linear programming problem

$$\begin{aligned} \text{Max} \quad & -\mathbf{p}^\top \mathbf{b} \\ \text{s.t.} \quad & \mathbf{p}^\top \mathbf{A} = \mathbf{0} \\ & \mathbf{p}^\top \mathbf{e} = 1 \\ & \mathbf{p} \geq \mathbf{0}, \end{aligned} \tag{2}$$

where \mathbf{e} is the vector with all components equal to 1. Show that the optimal cost in the problem is equal to v .

9. **(EX 4.10) (Saddle points of the Lagrangean)** Consider the standard form problem of minimizing $\mathbf{c}^\top \mathbf{x}$ subject to $\mathbf{Ax} = \mathbf{b}$ and $\mathbf{x} \geq \mathbf{0}$. We define the *Lagrangean* by

$$L(\mathbf{x}, \mathbf{p}) = \mathbf{c}^\top \mathbf{x} + \mathbf{p}^\top (\mathbf{b} - \mathbf{Ax}).$$

Consider the following “game”: player 1 choose some $\mathbf{x} \geq \mathbf{0}$, and player 2 choose some \mathbf{p} ; then, player 1 pays to player 2 the amount $L(\mathbf{x}, \mathbf{p})$. Player 1 would like to minimize $L(\mathbf{x}, \mathbf{p})$, while player 2 would like to maximize it.

A pair $(\mathbf{x}^*, \mathbf{p}^*)$, with $\mathbf{x}^* \geq \mathbf{0}$, is called an *equilibrium* point (or a *saddle point*, or a *Nash equilibrium*) if

$$L(\mathbf{x}^*, \mathbf{p}) \leq L(\mathbf{x}^*, \mathbf{p}^*) \leq L(\mathbf{x}, \mathbf{p}^*), \quad \forall \mathbf{x} \geq \mathbf{0}, \forall \mathbf{p}.$$

(Thus, we have an equilibrium if no player is able to improve her performance by unilaterally modifying her choice.)

Show that a pair $(\mathbf{x}^*, \mathbf{p}^*)$ is an equilibrium if and only if \mathbf{x}^* and \mathbf{p}^* are optimal solutions to the standard form problem under consideration and its dual, respectively.

10. **EX 4.20 (Strict complementary slackness)**

- (a) Consider the following linear programming problem and its dual

$$\begin{array}{ll} \text{Min} & \mathbf{c}^\top \mathbf{x} \\ \text{s.t.} & \mathbf{Ax} = \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0}, \end{array} \quad \begin{array}{ll} \text{Max} & \mathbf{p}^\top \mathbf{b} \\ \text{s.t.} & \mathbf{p}^\top \mathbf{A} \leq \mathbf{c}^\top \\ & \mathbf{p} \text{ is free,} \end{array}$$

and assume that both problems have an optimal solution. Fix some j . Suppose that every optimal solution to the primal satisfies $x_j = 0$. Show that there exists an optimal solution \mathbf{p} to the dual such that $\mathbf{p}^\top \mathbf{A}_j < c_j$ (Here, \mathbf{A}_j is the j th column of \mathbf{A} .)
Hint: Let d be the optimal cost. Consider the problem of minimizing $-x_j$ subject to $\mathbf{Ax} = \mathbf{b}, \mathbf{x} \geq \mathbf{0}$, and $-\mathbf{c}^\top \mathbf{x} \geq -d$, and form its dual.

- (b) Show that there exist optimal solutions \mathbf{x} and \mathbf{p} to the primal and to the dual, respectively, such that for every j we have either $x_j > 0$ or $\mathbf{p}^\top \mathbf{A}_j < c_j$. *Hint:* Use part (a) for each j , and then take the average of the vectors obtained.
- (c) Consider now the following linear programming problem and its dual:

$$\begin{array}{ll} \text{Min} & \mathbf{c}^\top \mathbf{x} \\ \text{s.t.} & \mathbf{Ax} \geq \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0}, \end{array} \quad \begin{array}{ll} \text{Max} & \mathbf{p}^\top \mathbf{b} \\ \text{s.t.} & \mathbf{p}^\top \mathbf{A} \leq \mathbf{c}^\top \\ & \mathbf{p} \geq \mathbf{0} \end{array}$$

Assume that both problems have an optimal solution. Show that there exist optimal solutions to the primal and to the dual, respectively, that satisfy *complementary slackness*, that is:

- i. For every j we have either $x_j > 0$ or $\mathbf{p}^\top \mathbf{A}_j < c_j$
- ii. For every i , we have either $\mathbf{a}_i^\top \mathbf{x} > b_i$ or $p_i > 0$. (Here, \mathbf{a}_i^\top is the i th row of \mathbf{A} .)

Hint: Convert the primal to the standard form and apply part(b).

- (d) Consider the linear programming problem

$$\begin{array}{ll} \text{Min} & 5x_1 + 5x_2 \\ \text{s.t.} & x_1 + x_2 \geq 2 \\ & 2x_1 - x_2 \geq 0 \\ & x_1, x_2 \geq 0 \end{array}$$

Does the optimal primal solution $(2/3, 4/3)$, together with the corresponding dual optimal solution, satisfy strict complementary slackness? Determine all primal and dual optimal solutions and identify the set of *all* strictly complementary pairs.

11. **(EX 4.21) (Clark's theorem)** Consider the following pair of linear programming problems:

$$\begin{array}{ll} \text{(P) Min} & \mathbf{c}^\top \mathbf{x} \\ \text{s.t.} & \mathbf{Ax} \geq \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{array} \qquad \begin{array}{ll} \text{(D) Max} & \mathbf{p}^\top \mathbf{b} \\ \text{s.t.} & \mathbf{p}^\top \mathbf{A} \leq \mathbf{c}^\top \\ & \mathbf{p} \geq \mathbf{0} \end{array}$$

Suppose that at least one of these two problems has a feasible solution. Prove that the set of feasible solutions to at least one of the two problems is unbounded. *Hint:* Interpret boundedness of a set in terms of the finiteness of the optimal cost of some linear programming problem.

12. **(EX 4.27)** Let \mathbf{A} be a given matrix. Show that the following two statements are equivalent.

- (a) Every vector such that $\mathbf{Ax} \geq \mathbf{0}$ and $\mathbf{x} \geq \mathbf{0}$ must satisfy $x_1 = 0$.
- (b) there exists some \mathbf{p} such that $\mathbf{p}^\top \mathbf{A} \leq \mathbf{0}$, $\mathbf{p} \geq \mathbf{0}$, and $\mathbf{p}^\top \mathbf{A}_1 < 0$, where \mathbf{A}_1 is the first column of \mathbf{A} .

13. **(EX 4.50) (Optimality conditions)** We are interested in the problem of deciding whether a polyhedron

$$Q = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{Ax} \leq \mathbf{b}, \mathbf{Dx} \geq \mathbf{d}, \mathbf{x} \geq \mathbf{0}\}$$

is nonempty. We assume that the polyhedron $P = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{Ax} \leq \mathbf{b}, \mathbf{x} \geq \mathbf{0}\}$ is nonempty and bounded. For any vector \mathbf{p} , of the same dimension as \mathbf{d} , we define

$$g(\mathbf{p}) = -\mathbf{p}^\top \mathbf{d} + \max_{\mathbf{x} \in P} \mathbf{p}^\top \mathbf{Dx}.$$

- (a) Show that if Q is nonempty, then $g(\mathbf{p}) \geq 0$ for all $\mathbf{p} \geq \mathbf{0}$.
 - (b) Show that if Q is empty, then there exists some $\mathbf{p} \geq \mathbf{0}$, such that $g(\mathbf{p}) < 0$.
 - (c) If Q is empty, what is the minimum of $g(\mathbf{p})$ over all $\mathbf{p} \geq \mathbf{0}$.
14. Prove the equivalence between the Min Max problem and the Max Min problem in the zero-sum game using strong duality of LP.