

## Problem Set 4

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**Exercise 1** Let  $m \in \mathbb{R}$  with  $m \geq 2$ . Suppose  $A_1, A_2, \dots, A_m$  are convex sets in  $\mathbb{R}^n$ , show if the following sets are convex or not. If it is, prove; if it is not, show one counter example.

**Solution 1**  $A_1, A_2, \dots, A_m$  are convex:  $\forall x_i, y_i \in A_i, \lambda x_i + (1 - \lambda)y_i \in A_i, \forall \lambda \in [0, 1]$ .

(1)  $\cup_{i=1}^m A_i$

False. Consider the following counterexample: Let  $R = [-2, -1]$  and  $P = [1, 2]$ . Both sets are convex. Their union  $U = [-2, -1] \cup [1, 2]$ , however, is not convex. Let  $x = -1 \in U$  and  $y = 1 \in U$ . Take the convex combination  $z = 1/2x + 1/2y = 0 \notin U$ . So, the union of convex sets is not necessarily convex.

(2)  $\cap_{i=1}^m A_i$

True. Take two arbitrary points  $x$  and  $y$  in  $\cap_{i=1}^m A_i$ . Since they belong to  $\cap_{i=1}^m A_i$ , they also belong to each of the convex sets  $A_1, A_2, \dots, A_m$ . It follows that  $\lambda x + (1 - \lambda)y$  is in the intersection of all the  $A_i$  sets for any  $\lambda \in [0, 1]$ . Thus  $\cap_{i=1}^m A_i$  is convex.

(3)  $\times_{i=1}^m A_i$

True. Consider two arbitrary vectors  $(x_1, \dots, x_m), (y_1, \dots, y_m) \in \times_{i=1}^m A_i$ , then their convex combination will be  $\lambda(x_1, \dots, x_m) + (1 - \lambda)(y_1, \dots, y_m) = (\lambda x_1, \dots, \lambda x_m) + ((1 - \lambda)y_1, \dots, (1 - \lambda)y_m) = (\lambda x_1 + (1 - \lambda)y_1, \dots, \lambda x_m + (1 - \lambda)y_m)$  for an arbitrary  $\lambda \in [0, 1]$ . Since  $\lambda x_i + (1 - \lambda)y_i \in A_i, i = 1, \dots, m$  it follows directly that  $(\lambda x_1 + (1 - \lambda)y_1, \dots, \lambda x_m + (1 - \lambda)y_m) \in \times_{i=1}^m A_i$ . Thus  $\times_{i=1}^m A_i$  is convex.

(4)  $\sum_{i=1}^m A_i$

True. The convex combination of two arbitrary vectors  $(x_1 + \dots + x_m), (y_1 + \dots + y_m) \in \sum_{i=1}^m A_i$ , is  $\lambda(x_1 + \dots + x_m) + (1 - \lambda)(y_1 + \dots + y_m) = (\lambda x_1 + \dots + \lambda x_m) + ((1 - \lambda)y_1 + \dots + (1 - \lambda)y_m) = (\lambda x_1 + (1 - \lambda)y_1 + \dots + \lambda x_m + (1 - \lambda)y_m)$  for an arbitrary  $\lambda \in [0, 1]$ . Since  $\lambda x_i + (1 - \lambda)y_i \in A_i, i = 1, \dots, m$  it follows directly that  $(\lambda x_1 + (1 - \lambda)y_1, \dots, \lambda x_m + (1 - \lambda)y_m) \in \sum_{i=1}^m A_i$ . Thus  $\sum_{i=1}^m A_i$  is convex.

**Exercise 2** Apply Fourier-Motzkin elimination method to solve the following system of linear inequalities:

$$-x_1 - x_2 - x_3 \leq -1 \wedge 3x_1 - x_2 - x_3 \leq 1 \wedge -x_1 + 3x_2 - x_3 \leq -2 \wedge -x_1 - x_2 + 3x_3 \leq 3$$

**Solution 2** We first eliminate  $x_1$ . Normalize the coefficient of  $x_1$  to 1 in each inequality, we have:

$$\begin{aligned} x_1 &\geq 1 - x_2 - x_3 \\ x_1 &\leq 1/3 + 1/3x_2 + 1/3x_3 \\ x_1 &\geq 2 + 3x_2 - x_3 \\ x_1 &\geq -3 - x_2 + 3x_3 \end{aligned}$$

which can be summarized as

$$\max \{1 - x_2 - x_3, 2 + 3x_2 - x_3, -3 - x_2 + 3x_3\} \leq x_1 \leq 1/3 + 1/3x_2 + 1/3x_3 \quad (1)$$

If the system of inequalities has a solution, the lower bound of  $x_1$  must be no greater than its upper bound. Thus,

$$\begin{aligned} 1 - x_2 - x_3 &\leq 1/3 + 1/3x_2 + 1/3x_3 \\ 2 + 3x_2 - x_3 &\leq 1/3 + 1/3x_2 + 1/3x_3 \\ -3 - x_2 + 3x_3 &\leq 1/3 + 1/3x_2 + 1/3x_3 \end{aligned}$$

We then eliminate  $x_2$ . Normalize the coefficient of  $x_2$  to 1 in each inequality, we have:

$$x_2 \leq -5/8 + 1/2x_3 \wedge x_2 \geq 1/2 - x_3 \wedge x_2 \geq -5/2 + 2x_3$$

which can be summarized as

$$\max \{1/2 - x_3, -5/2 + 2x_3\} \leq x_2 \leq -5/8 + 1/2x_3 \quad (2)$$

If the system of inequalities has a solution, the lower bound of  $x_2$  must be no greater than its upper bound. Thus,

$$1/2 - x_3 \leq -5/8 + 1/2x_3 \wedge -5/2 + 2x_3 \geq -5/8 + 1/2x_3$$

This further implies

$$3/4 \leq x_3 \leq 5/4 \quad (3)$$

To summarize, the set of solutions to our system of linear inequalities consists of all  $x \in \mathbb{R}^3$  such that inequalities 3, 2 and 1 hold.

**Exercise 3** Let  $C$  be the convex cone spanned by the vectors  $(1, 2, 2)^T, (3, 2, 5)^T$  and  $(1, 1, 0)^T$ . For each value of  $t$  determine if the point  $(t, t, t)^T$  lies in  $C$  or find a hyperplane that separates  $(t, t, t)^T$  from  $C$ .

**Solution 3**

We use Farka's Lemma:  $Ax = b, x \geq 0$

$$\left[ \begin{array}{ccc|c} 1 & 3 & 1 & t \\ 2 & 2 & 1 & t \\ 2 & 5 & 0 & t \end{array} \right] \xrightarrow{-R_2 + 2R_1, -2R_3 + R_1} \left[ \begin{array}{ccc|c} 1 & 3 & 1 & t \\ 0 & 4 & 1 & t \\ 0 & 1 & 2 & t \end{array} \right] \xrightarrow{4R_3 - R_2} \left[ \begin{array}{ccc|c} 1 & 3 & 1 & t \\ 0 & 4 & 1 & t \\ 0 & 0 & 7 & 3t \end{array} \right]$$

$$\implies x_3 = \frac{3t}{7}, x_2 = \frac{1t}{7}, x_1 = \frac{1t}{7} \implies x = t \begin{pmatrix} 1 \\ 1 \\ 3 \end{pmatrix}$$

Thus, for  $t \geq 0, x_1, x_2, x_3$  are greater or equal to zero as well. Therefore, for  $t \geq 0$  the point lies in the cone. For  $t < 0$  we want to find a hyperplane that, according to Farka's

Lemma, satisfies the following conditions:  $y^T A \geq 0^T, y^T b < 0$ . We choose  $y = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$  and

$$b = \begin{pmatrix} t \\ t \\ t \end{pmatrix} \text{ with } t < 0: y^T A = (1, 3, 1) \geq (0, 0, 0) \text{ and } (1, 0, 0)(t, t, t)^T = t < 0$$

Thus, we know that a separating hyperplane is given by the set:

$$\{y \in \mathbb{R}^3 : y_1 = 0\}$$

We show that this in fact a separating hyperplane. Let  $z \in C$ , there exists  $x \geq 0$  such that  $Ax = z$  (Note that  $Ax = (a_1, a_2, a_3)x$ ). This point lies in the positive halfspace since

$$y^T z = y^T (a_1 x_1 + a_2 x_2 + a_3 x_3) = x_1 y^T a_1 + x_2 y^T a_2 + x_3 y^T a_3 \geq 0.$$

However, for  $t < 0$  we know  $y^T t < 0$ . So it is in the negative halfspace. This establishes that  $y^T x = 0$  is a separating hyperplane.

**Exercise 4** For each of the following functions determine whether it is convex or concave. Are they quasi-concave? Motivate your answer.

**Solution 4**

(1)  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$  with  $f(x) = (x_1^2 + x_2^2)^2$

We can decompose  $f$  into  $f(x) = h_1(h_2(x))$  with  $h_1(t) = t^2, t \geq 0$ , and  $h_2(x) = x_1^2 + x_2^2$ .  $h_1$  is convex by Example 17.1 and  $h_2$  is convex by Example 17.1 and Theorem 17.6 (b). Moreover,  $h_1$  is nondecreasing since  $h_1'(x) = 2x \geq 0$  for all  $x \geq 0$  and defined over the convex interval  $(0, \infty)$ . By Theorem 17.6 (d),  $f(x)$  is convex. By Theorem 17.7 (a)

$f$  is quasiconcave if and only if for each  $r \in R$ , the set  $\{x \in C : f(x) \geq r\}$  is a convex set. The upper contour set of  $f$  consists of points with distance from the origin in the  $x_1, x_2$  plane greater than or equal to  $r^{1/4}$ . It is not convex, since, for example, the linepiece between such two points that have coordinates of the opposite sign includes points that are closer to the origin than  $c^{1/4}$ .  $f$  is not quasiconcave.

(2)  $f : \mathbb{R}_+^2 \rightarrow \mathbb{R}$  with  $f(x) = \sqrt{x_1 + x_2}$

Claim: linear functions are convex and concave. Using definition 17.1  $f$  is convex if  $\forall \lambda \in [0, 1] : f(\lambda x_1 + (1-\lambda)x_2) \leq \lambda f(x_1) + (1-\lambda)f(x_2)$ . Functions  $f = x$  as well as  $-f = -x$  are therefore convex since  $\forall \lambda \in [0, 1] : f(\lambda x_1 + (1-\lambda)x_2) = \lambda f(x_1) + (1-\lambda)f(x_2)$ . And since  $f$  is concave if and only if  $-f$  is convex, the claim follows.

We can decompose  $f$  into  $f(x) = h_1(h_2(x))$  with  $h_1(t) = \sqrt{t}$ ,  $t \geq 0$ , and  $h_2(x) = x_1 + x_2$ .  $h_1$  is concave by Theorem 17.4 as  $-h_1''(x) = 1/4x^{-3/2} \geq 0$  for all  $x \geq 0$ .  $h_2$  is concave by claim above and Theorem 17.6 (b). Moreover,  $h_1$  is nondecreasing since  $h_1'(x) = 1/2x \geq 0$  for all  $x \geq 0$  and defined over the convex interval  $(0, \infty)$ . By Theorem 17.6 (d),  $f(x)$  is concave. By Theorem 17.7 (a)  $f(x)$  is quasiconcave.

(3)  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$  with  $f(x) = (x_1 + x_2)^3$

We can decompose  $f$  into  $f(x) = h_1(h_2(x))$  with  $h_1(t) = t^3$ ,  $t \in \mathbb{R}$ , and  $h_2(x) = x_1 + x_2$ .  $h_1$  is neither concave or convex by Theorem 17.4 as  $h_1''(x) = 6x < 0$  for  $x < 0$  and  $-h_1''(x) = -6x < 0$  for  $x > 0$ . It follows that  $f$  is neither concave or convex

The upper contour set of  $f$  is a halfspace, namely the set of solutions to a single linear inequality  $x_1 + x_2 \geq r^{1/3}$ ,  $r \in R$ . It is convex by Example 14.1. By Theorem 17.7 (b)  $f(x)$  is quasiconcave.

**Exercise 5** Solve the problem:

$$\max f(x_1, x_2) = \sqrt{x_1} + \sqrt{x_2} \text{ with } 4x_1 + x_2 \leq 100, x_1 + x_2 \leq 60, x_1, x_2 \geq 0$$

**Solution 5** The goal function is given by the addition of two continuous functions. Hence, by Theorem 8.4 (b), it is continuous. The feasible set is the intersection of 3 closed sets and, by Theorem 7.3 (b), it is closed. We can show that it is bounded as follows: The bounds for  $x_1$  can be calculated by setting  $x_2 = 0$  in  $4x_1 + x_2 \leq 100$  and solving:  $0 \leq x_1 \leq 25$ . For  $x_2$ , we know that  $x_2 \geq 0$  and  $x_2 \leq 60 - x_1$ ,  $x_2 \leq 100 - 4x_1$  which implies that  $x_2 \leq 60$ . Therefore, the feasible set is bounded. By Heine-Borel we conclude that it is compact. It is also nonempty, as it contains point  $(x_1, x_2) = (0, 0)$ . By the Extreme Value Theorem, we conclude that this constrained maximization problem has a solution.

Rewrite the problem in the standard form: maximize  $f(x) = \sqrt{x_1} + \sqrt{x_2}$  subject to  $h_1(x) = -x_1 \leq 0, h_2(x) = -x_2 \leq 0, h_3(x) = 4x_1 + x_2 - 100 \leq 0, h_4(x) = x_1 + x_2 - 60 \leq 0$ .

The constraints  $h_1, h_2, h_3, h_4$  are affine functions, so we are in Case 3 of Theorem 19.4: a maximum must satisfy the KKT conditions. The Lagrangian is

$$\mathcal{L}(x, \mu) = \sqrt{x_1} + \sqrt{x_2} - \mu_1(-x_1) - \mu_2(-x_2) - \mu_3(4x_1 + x_2 - 100) - \mu_4(x_1 + x_2 - 60) - .$$

In a local maximum  $x$ , the following KKT-conditions must hold:

$$1/2x_1^{-1/2} + \mu_1 - 4\mu_3 - \mu_4 = 0 \quad (4)$$

$$1/2x_2^{-1/2} + \mu_2 - \mu_3 - \mu_4 = 0 \quad (5)$$

$$x_1 \geq 0 \quad (6)$$

$$x_2 \geq 0 \quad (7)$$

$$4x_1 + x_2 \leq 100 \quad (8)$$

$$x_1 + x_2 \leq 60 \quad (9)$$

$$\mu_1, \mu_2, \mu_3, \mu_4 \geq 0 \quad (10)$$

$$\mu_1 x_1 = 0 \quad (11)$$

$$\mu_2 x_2 = 0 \quad (12)$$

$$\mu_3(4x_1 + x_2 - 100) = 0 \quad (13)$$

$$\mu_4(x_1 + x_2 - 60) = 0 \quad (14)$$

Apparently  $(x_1, x_2) = (0, 0)$  cannot be optimal in our utility maximization. Further, since the square root and its first derivative are, respectively, increasing and decreasing functions, we know that the value of  $\sqrt{a} + \sqrt{b} > \sqrt{a+b}$ . It follows that in optimum,  $x_1 > 0, x_2 > 0, \mu_1 = \mu_2 = 0$ . Now distinguish four cases, depending on whether the nonnegativity constraints  $h_3, h_4$  are binding:

1.  $\mu_1 = \mu_2 = \mu_3 = \mu_4 = 0$  : By (4) and (5):  $x_1 = x_2 = 0$ .
2.  $\mu_1 = \mu_2 = \mu_3 = 0, \mu_4 > 0$  : By (4), (5) and (14):  $x_1 = x_2 = 30$ , contradicting (13).
3.  $\mu_1 = \mu_2 = \mu_4 = 0, \mu_3 > 0$  : By (4), (5) and (13):  $x_1 = 5, x_2 = 80$ , contradicting (14).
4.  $\mu_1 = \mu_2 = 0, \mu_3 > 0, \mu_4 > 0$  : By (13) and (14):  $x_1 = 40/3, x_2 = 140/3$ .

We showed that there is a maximum. We argued that  $(x_1, x_2) = (0, 0)$  cannot be optimal and found only one additional candidate. Conclude: the function is maximized if

$$x^* = \left( \frac{40}{3}, \frac{140}{3} \right)$$

and the maximal value is

$$f(x^*) = 2\sqrt{\frac{5}{3}(9 + 2\sqrt{14})}.$$