

The University of Texas at Austin
Optimization

HOMEWORK 1

Constantine Caramanis, Sujay Sanghavi

Submitting solutions: *Please submit your solutions as a single pdf file. If you have code or figures, please include these in the pdf.*

1. Convex Sets, Convex Functions, Preservation of Convexity

- (a) Show that the intersection of convex sets is convex. Thus, show that for any convex sets C_1 and C_2 , the intersection $C = C_1 \cap C_2$ is also convex. Do this using the definition of convexity, i.e., show that if $\mathbf{x}, \mathbf{y} \in C$, then for any $\lambda \in [0, 1]$, the point $\lambda\mathbf{x} + (1 - \lambda)\mathbf{y} \in C$.
- (b) Give an example where the union of two convex sets is not convex.
- (c) Show that the maximum of convex functions is convex. Note that given any two function, $f_1 : \mathbb{R}^n \rightarrow \mathbb{R}$ and $f_2 : \mathbb{R}^n \rightarrow \mathbb{R}$, their max is defined pointwise:

$$f_{\max}(\mathbf{x}) = \max\{f_1(\mathbf{x}), f_2(\mathbf{x})\}.$$

2. More Convex Sets, Convex Functions, Preservation of Convexity

- (a) Give an example where the minimum of two convex functions (defined analogously to the max of two functions, as explained above) is not convex.
- (b) (Boyd and Vandenberghe, Exercise 2.23). Give an example of two closed convex sets that are disjoint but cannot be strictly separated (read the chapter for the relevant definitions).
- (c) The sub-level sets of a function $f(\mathbf{x})$ are the sets of the form:

$$L_c = \{\mathbf{x} : f(\mathbf{x}) \leq c\},$$

for different constants $c \in \mathbb{R}$ (as defined in Lecture 2). Show that any sub-level set of a convex function is also convex. Also, give an example to show the converse is NOT true, i.e., give an example of a function that is not convex, but all of its sub-level sets are convex. Hint: you don't have to get too complicated: think of a simple 1-variable (univariate) function, hence, one whose graph you can draw, so that you can visually assess convexity of the sub-level sets, and lack of convexity of the function.

3. Consider two points, $\mathbf{v}_1, \mathbf{v}_2 \in \mathbb{R}^n$. In this exercise you will show that there exists a vector $\mathbf{c} \in \mathbb{R}^n$ and a scalar $d \in \mathbb{R}$ such that

$$\{\mathbf{x} : \|\mathbf{x} - \mathbf{v}_1\|_2 \leq \|\mathbf{x} - \mathbf{v}_2\|_2\} = \{\mathbf{x} : \mathbf{c}^\top \mathbf{x} \leq d\}.$$

- Two dimensions: consider the above problem in two dimensions, where $v_1 = (-1, 0)^\top$, and $v_2 = (1, 0)^\top$. Draw the set of points that are closer to v_1 than v_2 , by shading the plane. This should be a half-space!
- For the above example, find the two-dimensional vector $c = (c_1, c_2)^\top$ and the scalar d that describes the shaded region from the above. So in other words, the shaded region you drew must correspond to:

$$\{(x_1, x_2)^\top : c_1 x_1 + c_2 x_2 \leq d\}.$$

- Finally, generalize the above. For general points $v_1, v_2 \in \mathbb{R}^n$, find a vector $c \in \mathbb{R}^n$ and a scalar $d \in \mathbb{R}$ such that

$$\{x : \|x - v_1\|_2 \leq \|x - v_2\|_2\} = \{x : c^\top x \leq d\}.$$

Thus, you are showing that the set of points in \mathbb{R}^n that are closer to point v_1 than to point v_2 , form a half-space.

4. Prove that the set $\{x : \|Ax + b\|_2 \leq c^\top x + d\}$ is a convex set – **Optional**

5. (Boyd and Vandenberghe, Exercise 2.6) – **Optional**

- (a) When does one halfspace contain another? Give conditions under which

$$\{x | a^\top x \leq b\} \subseteq \{x | \tilde{a}^\top x \leq \tilde{b}\}$$

(where $a \neq 0, \tilde{a} \neq 0$). Also find the conditions under which the two halfspaces are equal.

- (b) What is the distance between the two parallel hyperplanes $\{x \in \mathbb{R}^n | a^\top x = b_1\}$ and $\{x \in \mathbb{R}^n | a^\top x = b_2\}$?