The Chinese University of Hong Kong, Shenzhen SDS · School of Data Science



Xiao Li \cdot Junfeng Wu \cdot Ming Yan \cdot Fall Semester 2023

MAT 3007 - Optimization

Assignment 7

Due: 11:59pm, Nov. 24 (Friday), 2023

Instructions:

- Homework problems must be carefully and clearly answered to receive full credit. Complete sentences that establish a clear logical progression are highly recommended.
- Please submit your assignment on Blackboard.
- The homework must be written in English.
- Late submission will not be graded.
- Each student must not copy homework solutions from another student or from any other source.

Problem 1 (Convex Sets):

(approx. 25 points)

In this exercise, we study convexity of various sets.

a) Verify whether the following sets are convex or not and explain your answer!

$$\Omega_1 = \{ x \in \mathbb{R}^n : \alpha \le (a^\top x)^2 \le \beta \}, \quad \alpha, \beta \in \mathbb{R}, \ 0 < \alpha \le \beta, \ a \in \mathbb{R}^n,$$

$$\Omega_2 = \{ (x, t) \in \mathbb{R}^n \times \mathbb{R} : x^\top x \le t^2 \}.$$

- b) Decide whether the following statements are true or false. Explain your answer and either present a proof / verification or a counter-example.
 - The intersection of two convex sets $\Omega_1, \Omega_2 \subset \mathbb{R}^n$ is always a convex set.
 - Let $\Omega \subset \mathbb{R}^n$ be a convex set and suppose that the set $S := \{(x,t) \in \Omega \times \mathbb{R} : f(x) \le t\} \subset \mathbb{R}^n \times \mathbb{R}$ is convex. Then, $f : \Omega \to \mathbb{R}$ is a convex function.

Problem 2 (Convex Compositions):

(approx. 20 points)

Either prove or find a counterexample for each of the following statements (you can assume that all functions are twice continuously differentiable if needed):

- a) If $f: \mathbb{R} \to \mathbb{R}$ and $g: \mathbb{R}^n \to \mathbb{R}$ are concave, then the composition $f \circ g: \mathbb{R}^n \to \mathbb{R}$, $(f \circ g)(x) = f(g(x))$ is concave.
- b) Let $\Omega \subset \mathbb{R}^n$ be a convex set and suppose that $g: \Omega \to \mathbb{R}$ is concave and $f: I \to \mathbb{R}$ is concave and nondecreasing where $I \supseteq g(\Omega)$ is an interval containing $g(\Omega)$. Then, $f \circ g$ is convex.
- c) If $f: \mathbb{R} \to \mathbb{R}$ is convex, then $x \mapsto |f(x)|$ is a convex function on \mathbb{R} .

Problem 3 (Convex Functions):

(approx. 30 points)

In this exercise, convexity properties of different functions are investigated.

- a) Let $r: \mathbb{R}^n \to \mathbb{R}$ be defined as $r(x) = \max_i |x_i|$. Show that r is a convex function.
- b) Verify that the following functions are convex over the specified domain:
 - $-f: \mathbb{R} \times \mathbb{R}_{++} \to \mathbb{R}, f(x) := x_1^2/x_2, \text{ where } \mathbb{R}_{++} := \{x \in \mathbb{R} : x > 0\}.$
 - $-f: \mathbb{R}^n \to \mathbb{R}, f(x) := \frac{1}{2} ||Ax b||^2 + \mu ||Lx||_{\infty}, \text{ where } A \in \mathbb{R}^{m \times n}, L \in \mathbb{R}^{p \times n}, b \in \mathbb{R}^m, \text{ and } \mu > 0 \text{ are given and } ||y||_{\infty} := \max_{i=1,\dots,p} |y_i|, y \in \mathbb{R}^p.$
 - $-f: \mathbb{R}^{n+1} \to \mathbb{R}, f(x,y) := \frac{\lambda}{2} ||x||^2 + \sum_{i=1}^m \max\{0, 1 b_i(a_i^\top x + y))\}, \text{ where } a_i \in \mathbb{R}^n \text{ and } b_i \in \{-1, 1\} \text{ are given data points for all } i = 1, ..., m \text{ and } \lambda > 0 \text{ is a parameter.}$
- c) Let us set $f(x) = ||x||_1 := \sum_{i=1}^n |x_i|$ and define $g : \mathbb{R}^n \to \mathbb{R}$, $g(x) := \max_{y \in \mathbb{R}^n} y^\top x f(y)$. Calculate g(x) explicitly and verify that the function g is convex.

Problem 4: (approx. 25 points)

- a) Let $A \in \mathbb{R}^{4 \times 4}$ be a symmetric matrix with nonnegative components and A1 = 1, i.e., each row of the matrix A has sum 1. Prove that I A is positive semidefinite.
- b) Let $a \in \mathbb{R}^n$. We define $f : \mathbb{R}^n \to \mathbb{R}$, $f(x) := \log(1 + \exp(a^{\top}x))$. Show that f is a convex function.
- c) Let $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$, $c \in \mathbb{R}^n$, and $d \in \mathbb{R}$. Prove that the nonconvex optimization problem

$$\min_{x \in \mathbb{R}^n} \quad \frac{\|Ax - b\|}{c^\top x + d} \\
\text{subject to} \quad \|x\| \le 1, c^T x + d > 0$$
(1)

is equivalent to the convex optimization problem

$$\min_{\substack{y \in \mathbb{R}^n, t \\ \text{subject to}}} ||Ay - bt||
\sup_{\substack{y \in \mathbb{R}^n, t \\ c^{\top}y + dt = 1}} (2)$$

d) Use CVX (in MATLAB or Python) to solve problem (2) with

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}, \quad b = \begin{bmatrix} 2 \\ 4 \end{bmatrix}, \quad c = \begin{bmatrix} 4 \\ 3 \end{bmatrix}, \quad d = 1$$