

IE 539 Convex Optimization Assignment 2

Fall 2024

Out: 14th October 2024

Due: 25th October 2024 at 11:59pm

Instructions

- Submit a PDF document with your solutions through the assignment portal on KLMS by the due date. Please ensure that your name and student ID are on the front page.
- Late assignments will **not** be accepted except in extenuating circumstances. Special consideration should be applied for in this case.
- It is **required** that you typeset your solutions in LaTeX. Handwritten solutions will not be accepted.
- Spend some time ensuring your arguments are **coherent** and your solutions **clearly** communicate your ideas.

Question:	1	2	3	4	5	6	Total
Points:	15	15	15	15	20	20	100

1. (15 points) For a fixed z , derive closed form optimal solutions for
 - (a) $\min_{x \in \mathbb{R}^d} \left\{ \frac{1}{2} \|x - z\|_2^2 + \lambda \|x\|_2^2 \right\}$. (Here, $\lambda > 0$.)
 - (b) $\min_{x \in \mathbb{R}^d} \left\{ \frac{1}{2} \|x - z\|_2^2 + \lambda \|x\|_1 \right\}$. (Here, $\lambda > 0$.)
2. (15 points) Let $X = \{x \in \mathbb{R}^d : \|x - x_0\|_2 \leq r\}$ be a ball of radius r centred at the point x_0 . Find an expression for the projection onto X . In other words, for $z \in \mathbb{R}^d$, find $\arg \min_{x \in X} \left\{ \frac{1}{2} \|x - z\|_2^2 : x \in X \right\}$. To get full marks, you must also *prove* that your expression is correct.
3. (15 points) Recall the ℓ_1 - and ℓ_∞ -norms are defined as

$$\|x\|_1 := \sum_{i \in [d]} |x_i|, \quad \|x\|_\infty := \max_{i \in [d]} |x_i|.$$

The dual norm is defined as

$$\|x\|_* := \max_{z: \|z\| \leq 1} x^\top z.$$

Show that the ℓ_1 -norm is the dual norm to the ℓ_∞ -norm, and vice versa.

4. (15 points) Prove that (SOCP) is a conic program.

$$\begin{aligned} & \text{minimize} && f^\top x \\ & \text{subject to} && \|A_i x + b_i\|_2 \leq c_i^\top x + d_i \quad \text{for } i = 1, \dots, m, \\ & && E x = g. \end{aligned} \tag{SOCP} \quad \boxed{\text{eq: socp}}$$

5. (20 points) In this question we will work through the convergence proof of projected gradient descent and derive rates for Lipschitz continuous functions under various step sizes. The algorithm proceeds as follows:

- Choose $x_1 \in \mathbb{R}^d$.
- For $t = 1, 2, 3, \dots, T + 1$:
 - Obtain $\nabla f(x_t)$ and step size $\eta_t > 0$.
 - Compute $x_{t+1} = x_t - \eta_t \nabla f(x_t)$.

Assume that f is L -Lipschitz continuous in the ℓ_2 -norm.

- (a) Set $\eta_t = \frac{1}{\sqrt{t}}$ for $t \geq 1$. Prove that

$$f \left(\left(\sum_{t=1}^T \eta_t \right)^{-1} \sum_{t=1}^T \eta_t x_t \right) - f(x^*) = O \left(\frac{\log T}{\sqrt{T}} \right).$$

- (b) Set $\eta_t = \frac{1}{t}$ for $t \geq 1$. Prove that

$$f \left(\left(\sum_{t=1}^T \eta_t \right)^{-1} \sum_{t=1}^T \eta_t x_t \right) - f(x^*) = O \left(\frac{1}{\log T} \right).$$

6. (20 points) In this question we will work through the convergence proof of projected gradient descent and derive rates for Lipschitz continuous functions. The algorithm proceeds as follows:

- Choose $x_1 \in C$.
- For $t = 1, 2, 3, \dots, T + 1$:
 - Obtain $\nabla f(x_t)$ and step size $\eta_t > 0$.
 - Compute $x_{t+1} = \arg \min_{x \in C} \|x_t - \eta_t \nabla f(x_t) - x\|_2^2$.

Assume that f is L -Lipschitz continuous in the ℓ_2 -norm and that there exists some constant $R > 0$ such that

$$\|x - y\|_2^2 \leq R^2$$

for all $x, y \in C$.

- (a) Show that for each t , we have

$$\|x_{t+1} - x^*\|_2^2 \leq \|x_t - \eta_t \nabla f(x_t) - x^*\|_2^2$$

where x^* is an optimal solution to $\min_{x \in C} f(x)$.

(b) Set $\eta_t = \frac{\|x_1 - x^*\|_2}{L\sqrt{T}}$. Then use part (a) to show that

$$f\left(\frac{1}{T} \sum_{t=1}^T x_t\right) - f(x^*) \leq \frac{L\|x_1 - x^*\|_2}{\sqrt{T}}.$$

(c) Set $\eta_t = \frac{1}{\sqrt{t}}$. Prove that

$$f\left(\frac{1}{T} \sum_{t=1}^T x_t\right) - f(x^*) = O\left(\frac{1}{\sqrt{T}}\right)$$