

MSBD5007 Optimization and Matrix Computation

Homework 5

Due date: 4 May, Sunday

1. Find explicit formulas for the projection of $\mathbf{y} \in \mathbb{R}^n$ onto the following non-empty, closed, and convex sets $S \subset \mathbb{R}^n$, respectively.

- (a) The unit ∞ -norm ball

$$S = \{\mathbf{x} \in \mathbb{R}^n \mid \|\mathbf{x}\|_\infty \leq 1\}.$$

- (b) The closed halfspace

$$S = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{a}^T \mathbf{x} \leq b\},$$

where $\mathbf{a} \in \mathbb{R}^n$, $\mathbf{a} \neq \mathbf{0}$, and $b \in \mathbb{R}$ are given.

2. Consider the optimization problem in $\mathbf{x} = (x_1, x_2, x_3) \in \mathbb{R}^3$:

$$\begin{aligned} \min_{\mathbf{x} \in \mathbb{R}^3} \quad & \frac{1}{2}(x_1^2 + 4x_2^2 + 9x_3^2) - (4x_1 + 2x_2), \\ \text{s.t.} \quad & x_1 + x_2 + x_3 = 3, \\ & x_i \geq 0, \quad i = 1, 2, 3. \end{aligned}$$

- (a) Write down the KKT conditions (stationarity, feasibility, complementary slackness).
(b) Solve the KKT system to find the optimal solution \mathbf{x}^* , the Lagrange multiplier λ^* for the equality constraint, and the multipliers $\boldsymbol{\mu}^* = (\mu_1, \mu_2, \mu_3)$ for the inequality constraints.
3. We wish to compute the projection of $\mathbf{y} \in \mathbb{R}^n$ onto the unit ℓ_1 ball, i.e. solve

$$\min_{\mathbf{x} \in \mathbb{R}^n} \frac{1}{2} \|\mathbf{x} - \mathbf{y}\|_2^2 \quad \text{s.t.} \quad \|\mathbf{x}\|_1 \leq 1.$$

- (a) Derive the Lagrange dual problem and show it can be written as

$$\max_{\lambda \geq 0} d(\lambda), \quad d(\lambda) = \sum_{i=1}^n h_\lambda(y_i) - \lambda,$$

where $h_\lambda : \mathbb{R} \rightarrow \mathbb{R}$ is the so-called Huber's function (which is a smooth function consisting of a quadratic and two linear pieces) defined by

$$h_\lambda(t) = \begin{cases} \frac{1}{2} t^2, & |t| \leq \lambda, \\ \lambda |t| - \frac{1}{2} \lambda^2, & |t| \geq \lambda. \end{cases}$$

- (b) Prove that strong duality holds.
(c) Find the optimal dual multiplier λ^* .
(d) Give an expression of the projection in terms of λ^* .

4. Let $S \subseteq \mathbb{R}^n$ be a nonempty, closed, convex set, and let $\|\cdot\|_2$ denote 2-norm. The projection of any point $\mathbf{y} \in \mathbb{R}^n$ onto S is defined by

$$\mathcal{P}_S(\mathbf{y}) = \arg \min_{\mathbf{x} \in S} \|\mathbf{x} - \mathbf{y}\|_2.$$

Prove that the projection $\mathcal{P}_S : \mathbb{R}^n \rightarrow S$ is *nonexpansive*: for all $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$,

$$\|\mathcal{P}_S(\mathbf{x}) - \mathcal{P}_S(\mathbf{y})\|_2 \leq \|\mathbf{x} - \mathbf{y}\|_2.$$