

Mock Exam 2: Question 4

April 26, 2022

Consider the problem

$$\min_{\mathbf{x} \in \mathbb{R}^n} \sum_{i=1}^m \|\mathbf{A}_i \mathbf{x} + \mathbf{b}_i\|$$

where $\mathbf{A}_i \in \mathbb{R}^{k_i \times n}$, $\mathbf{b}_i \in \mathbb{R}^{k_i}$, $i = 1, 2, \dots, m$.

- i Using auxiliary variables $\mathbf{y}_i = \mathbf{A}_i \mathbf{x} + \mathbf{b}_i$, write an equivalent constrained optimization problem, derive its Lagrangian, and the dual objective function.
- ii Derive an explicit formulation of the dual problem.

Solutions

- i) Using the suggested auxiliary variables, the equivalent formulation reads

$$\begin{aligned} \min_{\mathbf{x}, \mathbf{y}_i} \quad & \sum_{i=1}^m \|\mathbf{y}_i\| \\ \text{s.t.} \quad & \mathbf{y}_i = \mathbf{A}_i \mathbf{x} + \mathbf{b}_i, \quad i = 1, 2, \dots, m \end{aligned}$$

Associating a Lagrange multiplier vector $\lambda_i \in \mathbb{R}^{k_i}$ with the i th set of constraints $\mathbf{y}_i = \mathbf{A}_i \mathbf{x} + \mathbf{b}_i$, we obtain the following Lagrangian:

$$\begin{aligned} L(\mathbf{x}, \lambda_1, \lambda_2, \dots, \lambda_m) &= \sum_{i=1}^m \|\mathbf{y}_i\| + \sum_{i=1}^m \lambda_i^\top (\mathbf{y}_i - \mathbf{A}_i \mathbf{x} - \mathbf{b}_i) \\ &= \sum_{i=1}^m [\|\mathbf{y}_i\| + \lambda_i^\top \mathbf{y}_i] - \left(\sum_{i=1}^m \mathbf{A}_i^\top \lambda_i \right)^\top \mathbf{x} - \sum_{i=1}^m \mathbf{b}_i^\top \lambda_i \end{aligned}$$

By the separability of the Lagrangian with respect to $\mathbf{x}, \mathbf{y}_1, \mathbf{y}_2, \dots, \mathbf{y}_m$, it follows that the dual objective function is given by

$$q(\lambda_1, \lambda_2, \dots, \lambda_m) = \sum_{i=1}^m \min_{\mathbf{y}_i \in \mathbb{R}^{k_i}} [\|\mathbf{y}_i\| + \lambda_i^\top \mathbf{y}_i] + \min_{\mathbf{x} \in \mathbb{R}^n} \left[- \left(\sum_{i=1}^m \mathbf{A}_i^\top \lambda_i \right)^\top \mathbf{x} \right] - \sum_{i=1}^m \mathbf{b}_i^\top \lambda_i.$$

- ii) Observing the terms in the dual objective, it is clear that

$$\min_{\mathbf{x} \in \mathbb{R}^n} \left[- \left(\sum_{i=1}^m \mathbf{A}_i^\top \lambda_i \right)^\top \mathbf{x} \right] = \begin{cases} 0, & \sum_{i=1}^m \mathbf{A}_i^\top \lambda_i = \mathbf{0}, \\ -\infty & \text{else.} \end{cases}$$

In addition, for any $i = 1, \dots, m$ we have

$$\min_{y_i \in \mathbb{R}^{k_i}} [\|y_i\| + \lambda_i^\top y_i] = \begin{cases} 0, & \|\lambda_i\| \leq 1, \\ -\infty & \text{else} \end{cases}.$$

Combining these expressions, the dual objective function is

$$q(\lambda_1, \lambda_2, \dots, \lambda_m) = \begin{cases} -\sum_{i=1}^m \lambda_i^\top \mathbf{b}_i, & \|\lambda_i\| \leq 1, i = 1, 2, \dots, m, \sum_{i=1}^m \mathbf{A}_i^\top \lambda_i = 0, \\ -\infty & \text{else.} \end{cases}$$

The dual problem is therefore

$$\begin{aligned} \max \quad & -\sum_{i=1}^m \mathbf{b}_i^\top \lambda_i \\ \text{s.t.} \quad & \sum_{i=1}^m \mathbf{A}_i^\top \lambda_i = 0, \\ & \|\lambda_i\| \leq 1, \quad i = 1, 2, \dots, m \end{aligned}$$