

# Simple Bilevel Programming and Extensions: Theory and Algorithms

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## Abstract

Let  $S = \operatorname{argmin} \{h(x) : x \in C\}$  denote the solution set of a convex optimization problem, where  $C \subset \mathbb{R}^n$  is a closed convex set and  $f, h : \mathbb{R}^n \rightarrow \mathbb{R}$  are real-valued convex functions. The simple bilevel optimization problem is

$$\min\{f(x) : x \in S\}. \quad (1)$$

To investigate this convex optimization problem, the lower level problem needs to be transformed. If  $\alpha = \min\{h(x) : x \in C\}$  denote the optimal value of this problem, it is equivalent to  $\min\{f(x) : h(x) \leq \alpha, x \in C\}$ . Slater's regularity condition is violated for this convex optimization problem. Using a variational inequality to express the set  $S$ , a simple MPEC

$$\min\{f(x) : x \in C, \psi_y(x) \leq 0 \forall y \in C\} \quad (2)$$

arises, where  $\psi_y(x) = \langle \nabla h(y), x - y \rangle$ . (2) is a convex optimization problem. Results from semi-infinite optimization or the use of a gap function for the variational inequality can be applied to derive necessary and sufficient optimality conditions for (2) and, hence, for the original problem. One such optimality condition reads as:

A feasible point  $\bar{x}$  is optimal for (2) if and only if there exist  $k \in \mathbb{N}$ ,  $\lambda_1, \lambda_2, \dots, \lambda_k > 0$ ,  $y_1, y_2, \dots, y_k \in C$  such that

$$0 \in \partial f(\bar{x}) + \sum_{i=1}^k \lambda_i \nabla h(y_i) + N_C(\bar{x}) \text{ and } \langle \nabla h(y_i), \bar{x} - y_i \rangle = 0, \text{ for all } i = 1, 2, \dots, k$$

provided some closedness qualification condition is satisfied and  $\nabla h(x)$  is continuous and monotone.

In the second part of the talk, an idea for solving the problem will be given. Basis for this algorithm is a penalization  $\xi_\varepsilon(x) = h(x) + \varepsilon f(x)$ , where both  $f, h$  are assumed to be convex but not necessarily differentiable. The algorithm computes a sequence of  $\eta_k$ -optimal solutions of minimizing a Moreau-Yosida regularization of the function  $\xi_\varepsilon(x)$  over  $C$  which converges to to an optimal solution provided this problem has a solution.

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