

Multiobjective bilevel optimization: A set-valued optimization view point

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Recent Advances in Multi-Objective Optimization

University of Lancaster, June 24, 2016

Bilevel optimization problem

Conceptual definition:

“min” $F(x, y)$ s.t. $x \in X$, $y \in S(x)$

where the map S is defined by

$$S(x) := \arg \min_y \{f(x, y) : y \in K(x)\}$$

Case where problem is well-posed

(i.e., $S(x) = \{y(x)\}$ for all $x \in X$):

$$(P_i) \quad \min_x F(x, y(x)) \quad \text{s.t.} \quad x \in X$$

Outrata et al. (1998), Dempe (2002), Falk & Liu (1995), Kolstad & Ladson (1990), Savard & Gauvin (1994), Vicente et al. (1994), Mersha (2011), etc.

Forcing uniqueness in the lower-level problem

$$\min_y f(x, y) + \alpha \pi(x, y) \quad \text{s.t.} \quad g(x, y) \leq 0 \quad (\alpha > 0)$$

Example: $\pi(x, y) := \|y\|^2$ (**Tikhonov regularization**)

See Dempe & Schmidt (1996), Dempe & Bard (2001), Morgan & Patrone (2006), Bergounioux & Haddou (2008), Molodtsov (1976), etc.

Purpose of the talk

Set-valued optimization model:

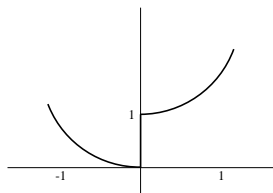
$$(P_s) \quad \min_x \mathcal{F}(x) := F(x, S(x)) \quad \text{s.t. } x \in X$$

Example by Lucchetti et al. (1987):

$$F(x, y) := x^2 + y^2, \quad X := [-1, 1]$$

$$S(x) := \arg \min_y \{-xy : y \in [0, 1]\}$$

$$S(x) = \begin{cases} 0 & \text{if } x < 0 \\ 1 & \text{if } x > 0 \\ [0, 1] & \text{if } x = 0 \end{cases}$$



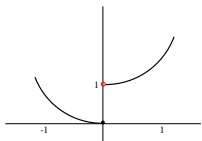
(c) graph of \mathcal{F}

$(\bar{x}, \bar{z}) \in \text{gph } \mathcal{F} := \{(x, z) \in \mathbb{R}^n \times \mathbb{R} : z \in \mathcal{F}(x)\}$ is a **local Pareto optimal solution** of (P_s) if there exists a neighborhood U of \bar{x} such that

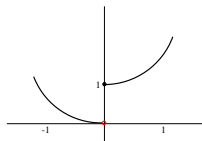
$$(\mathcal{F}(X \cap U) - \bar{z}) \cap (-\infty, 0] = \emptyset.$$

How can the set-valued model be useful?

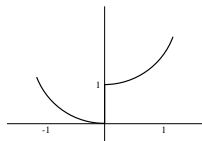
- ▶ (P) $\min_{x,y} F(x,y)$ s.t. $x \in X, y \in S(x)$
- ▶ (P_o) $\min_{x \in X} \varphi_o(x) := \min_y \{F(x,y) \mid y \in S(x)\}$
- ▶ (P_p) $\min_{x \in X} \varphi_p(x) := \max_y \{F(x,y) \mid y \in S(x)\}$



(a) graph of φ_o



(b) graph of φ_p



(c) graph of \mathcal{F}

$$\begin{array}{ccccccc}
 (P_p) & \overset{(**)}{\iff} & (P_i) & \overset{(**)}{\iff} & (P_o) & \iff & (P_s) \\
 & & & & \updownarrow (*) & & \\
 & & & & (P) & &
 \end{array}$$

Another perspective

Bilevel optimization as multiobjective optimization

- ▶ Marcotte and Savard (1991)
- ▶ Fülöp (1993)
- ▶ Fliege and Vicente (2006)
- ▶ Ruuska, Miettinen and Wiecek (2012), etc.

Related solution approaches

- ▶ Glackin, Ecker and Kupferschmid (2009)
- ▶ Alves, Dempe and Júdice (2010)
- ▶ Pieume, Fotso and Siarry (2009), etc.

Other set-valued related bilevel models

- ▶ Bonnel (2006), Bonnel and Morgan (2012)
- ▶ Eichfelder (2008), Ye (2011)
- ▶ Dempe and Gadhi (2010), Dempe and Pilecka (2015), etc.

Summary of results for scalar objectives

Theorem: The following properties are satisfied:

- (a) $(\text{gph } \varphi_o \cup \text{gph } \varphi_p) \cap (X \times \mathbb{R}) \subseteq \text{gph } \mathcal{F}$.
- (b) Let $F(x, y) := a(x)^\top y + b(x)$ with $a : \mathbb{R}^n \rightarrow \mathbb{R}^m$ and $b : \mathbb{R}^n \rightarrow \mathbb{R}$, and assume that S is convex-valued on X^m . Then, for all $x \in X$,

$$\mathcal{F}(x, S(x)) \begin{cases} = \varphi_o(x) = \varphi_p(x) & \text{if } x \in X^u \\ \supseteq [\varphi_o(x), \varphi_p(x)] & \text{if } x \in X^m \end{cases}$$

Theorem: \bar{x} is an optimal solution of problem (P_o) if and only if there exists a vector $\bar{z} \in \mathcal{F}(\bar{x})$ such that (\bar{x}, \bar{z}) is a Pareto optimal solution of problem (P_s) .

Theorem: Let $(\bar{x}, \bar{z}) \in \text{gph } \mathcal{F}$ be a local Pareto optimal solution of (P_s) and assume that S is closed, locally bounded around \bar{x} and

$$D^*S(\bar{x}|y)(0) \cap (-N_X(\bar{x})) = \{0\} \quad \forall y \in S(\bar{x}) \text{ s.t. } \bar{z} = F(\bar{x}, y).$$

Then, there exists $\bar{y} \in S(\bar{x})$ with $\bar{z} = F(\bar{x}, \bar{y})$ such that we have

$$-\nabla_x F(\bar{x}, \bar{y}) \in D^*S(\bar{x}|\bar{y})(\nabla_y F(\bar{x}, \bar{y})) + N_X(\bar{x}).$$

Results for vector-valued objectives

$$F : \mathbb{R}^n \times \mathbb{R}^m \longrightarrow \mathbb{R}^{\ell_1} \text{ and } f : \mathbb{R}^n \times \mathbb{R}^m \longrightarrow \mathbb{R}^{\ell_2}$$

$$S_{wef}(x) = \bigcup_{y \in Y} S_s(x, y) := \arg \min_z \{ \langle y, f(x, z) \rangle : z \in K(x) \}$$

Proposition: Assume that S_s is closed and let $\bar{z} \in S_{wef}(\bar{x})$. Further suppose

$$\left[(0, -u) \in D^* S_s((\bar{x}, \bar{y}) | \bar{z})(0), u \in U(\bar{y}) \right] \implies u = 0 \quad (1)$$

hold for all $\bar{y} \in Y$ such that $\bar{z} \in S_s(\bar{x}, \bar{y})$. Then for all $z^* \in \mathbb{R}^m$, we have

$$D^* S_{wef}(\bar{x} | \bar{z})(z^*) \subseteq \left\{ x^* : (x^*, -\bar{u}) \in D^* S_s((\bar{x}, \bar{y}) | \bar{z})(z^*), \right. \\ \left. \bar{y} \in Y, \bar{u} \in U(\bar{y}), \bar{z} \in S_s(\bar{x}, \bar{y}) \right\}.$$

Theorem: Let $(\bar{x}, z^0) \in \text{gph} \mathcal{F}$ be a local Pareto solution of (P_s) and let S_s be closed and locally bounded near \bar{x} . Further let **QC (1)**

$$\left[(x^*, -u) \in D^* S_s((\bar{x}, y) | z)(0), x^* \in -N_X(\bar{x}) \right] \implies x^* = 0.$$

Then, there exist $z^* \in \mathbb{R}^{\ell_2}$ with $\|z^*\| = 1$ and a vector $(\bar{y}, \bar{u}, \bar{z})$ such that

$$-(\nabla_x F(\bar{x}, \bar{z}))^\top z^*, \bar{u} \in D^* S_s((\bar{x}, \bar{y}) | \bar{z})(\nabla_z F(\bar{x}, \bar{z}))^\top z^* + N_X(\bar{x}) \times \{0_{\ell_2}\}.$$

Illustrative examples

Scalar objectives example

$$F(x, y) := -x + 10y_1 - y_2$$

$$X := [0, \infty[$$

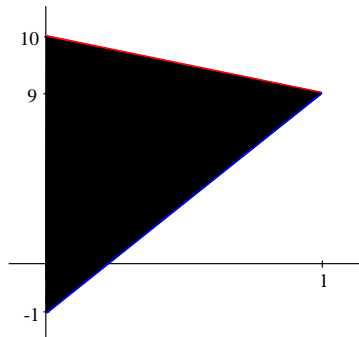
$$S(x) := \arg \min_y \{-y_1 - y_2 :$$

$$y_1, y_2 \geq 0$$

$$x - y_1 \leq 1$$

$$x + y_2 \leq 1$$

$$y_1 + y_2 \leq 1\}$$



Semivectorial example

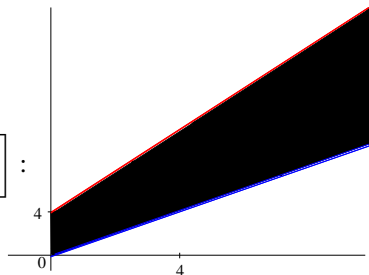
$$F(x, z) := x + z_2$$

$$X := [0, \infty[$$

$$S(x) := \arg \min_z \left\{ \begin{bmatrix} 2z_1 + 2z_2 \\ -z_1 + z_2 \end{bmatrix} : \right.$$

$$z_1, z_2 \geq 0 \left. \right\}$$

$$-x + z_1 + z_2 \geq 4 \left. \right\}$$



Ongoing & future research topics

- ▶ Solution algorithms based on set-valued optimization
 - ▶ Generalization of the implicit function model techniques
 - ▶ Further analysis of the map $\mathcal{F}(x) := F(x, S(x))$
- ▶ Implementation of the methods on the pessimistic model



Alain B. Zemkoho (2016). Solving ill-posed bilevel programs. *Set-Valued and Variational Analysis*, *in press*.