

# Mixed-Integer Linear Bilevel Optimization: A New Single-Level Reformulation

Henri Lefebvre, Martin Schmidt

## Problem Statement

- ▶ We consider bilevel optimization problems of the form

$$\begin{aligned} \min_{x \in X, y} \quad & c^\top x + d^\top y \\ \text{s.t.} \quad & Ax + By \geq a, \\ & y \in \arg \min_{y' \in Y} \{f^\top y' : Cx + Dy' \geq b\}. \end{aligned}$$

- ▶ Using the **value function**  $\varphi(x) := \min_{y \in Y} \{f^\top y : Cx + Dy \geq b\}$ :

$$\begin{aligned} \min_{x \in X, y \in Y} \quad & c^\top x + d^\top y \\ \text{s.t.} \quad & Ax + By \geq a, \quad Cx + Dy \geq b, \quad f^\top y \leq \varphi(x). \end{aligned}$$

## LP Case: Single-Level Reformulation

- ▶ Assume  $Y = \mathbb{R}^{n_y}$ , then the follower's problem is an LP.
- ▶ Replace the follower by its dual problem:

$$\varphi(x) = \max_{\lambda} \{(b - Cx)^\top \lambda : D^\top \lambda = f, \lambda \geq 0\}.$$

- ▶ We obtain the strong-duality single-level reformulation:

$$\begin{aligned} \min_{x \in X, y \in Y, \lambda} \quad & c^\top x + d^\top y \\ \text{s.t.} \quad & Ax + By \geq a, \\ & Cx + Dy \geq b, \\ & f^\top y \leq (b - Cx)^\top \lambda, \\ & D^\top \lambda = f, \lambda \geq 0. \end{aligned}$$

- ▶ Can be reformulated as a MILP and solved to global optimality.
- ▶ See Fortuny-Amat and McCarl (1981).

## LP Case: Penalty Alternating Direction Method

- ▶ The above single-level reformulation may be too hard to solve.
- ▶ We resort to a heuristic approach, the PADM.

with  $\rho \nearrow \infty$

$$\begin{aligned} \min_{x \in X, y \in Y} \quad & c^\top x + d^\top y + \rho(b^\top \bar{\lambda} - \bar{\lambda}^\top Cx - f^\top y) \\ \text{s.t.} \quad & Ax + By \geq a, \quad Cx + Dy \geq b. \end{aligned}$$

$$(\bar{x}, \bar{y}) \left( \begin{array}{c} \min_{\lambda} \quad (b - C\bar{x})^\top \lambda \\ \text{s.t.} \quad D^\top \lambda = f, \lambda \geq 0. \end{array} \right) \bar{\lambda}$$

- ▶ See Kleinert and Schmidt (2020), Lefebvre and Schmidt (2024).

## The Follower's Dantzig-Wolfe Reformulation

- ▶ Assume  $X = \{0, 1\}^{n_x}$  and  $Y \subset \mathbb{Z}^{n_y - p_y} \times \mathbb{R}^{p_y}$ .
- ▶ Consider any pair  $(\hat{x}, \hat{y})$  such that  $\hat{y} \in Y(\hat{x})$ , then

$$\varphi(x) = \min_{\alpha \in \{0, 1\}^\ell} \left\{ \sum_{k=1}^{\ell} \alpha_k f^\top \hat{y}^k : \sum_{k=1}^{\ell} \alpha_k \hat{x}^k = x, \sum_{k=1}^{\ell} \alpha_k = 1 \right\}. \quad (*)$$

- ▶ One can show that **the LP relaxation is tight** for all  $x \in X$ .
- ▶ This requires that  $x$  is binary.
- ▶ One can even show that it is the convex envelope of  $\varphi$  on  $X$ .
- ▶ With this, we go from the MILP to the LP world!

## MILP Case: A New Single-Level Reformulation

- ▶ Replace the follower by the dual of the tight LP relaxation of (\*):

$$\varphi(x) = \max_{\pi \in \mathbb{R}, \lambda \in \mathbb{R}^{n_x}} \left\{ \pi + \lambda^\top x : \pi + \lambda^\top \hat{x}^k \leq f^\top \hat{y}^k, k = 1, \dots, \ell \right\}.$$

- ▶ We obtain a **new strong-duality single-level reformulation**:

$$\begin{aligned} \min_{x \in X, y \in Y, \pi, \lambda} \quad & c^\top x + d^\top y \\ \text{s.t.} \quad & Ax + By \geq a, \\ & Cx + Dy \geq b, \\ & f^\top y \leq \pi + \lambda^\top x, \\ & \pi + \lambda^\top \hat{x}^k \leq f^\top \hat{y}^k, \quad k = 1, \dots, \ell. \end{aligned}$$

- ▶ Closed-form bounds on  $\lambda$  and  $\pi$  lead to a MILP.
- ▶ Note: exponential number of constraints.
- ▶ To be expected since MILP bilevel problems are  $\Sigma_2^P$ -hard.

## MILP Case: Penalty Alternating Direction Method

- ▶ The **PADM** can be extended to MILP bilevel problems.

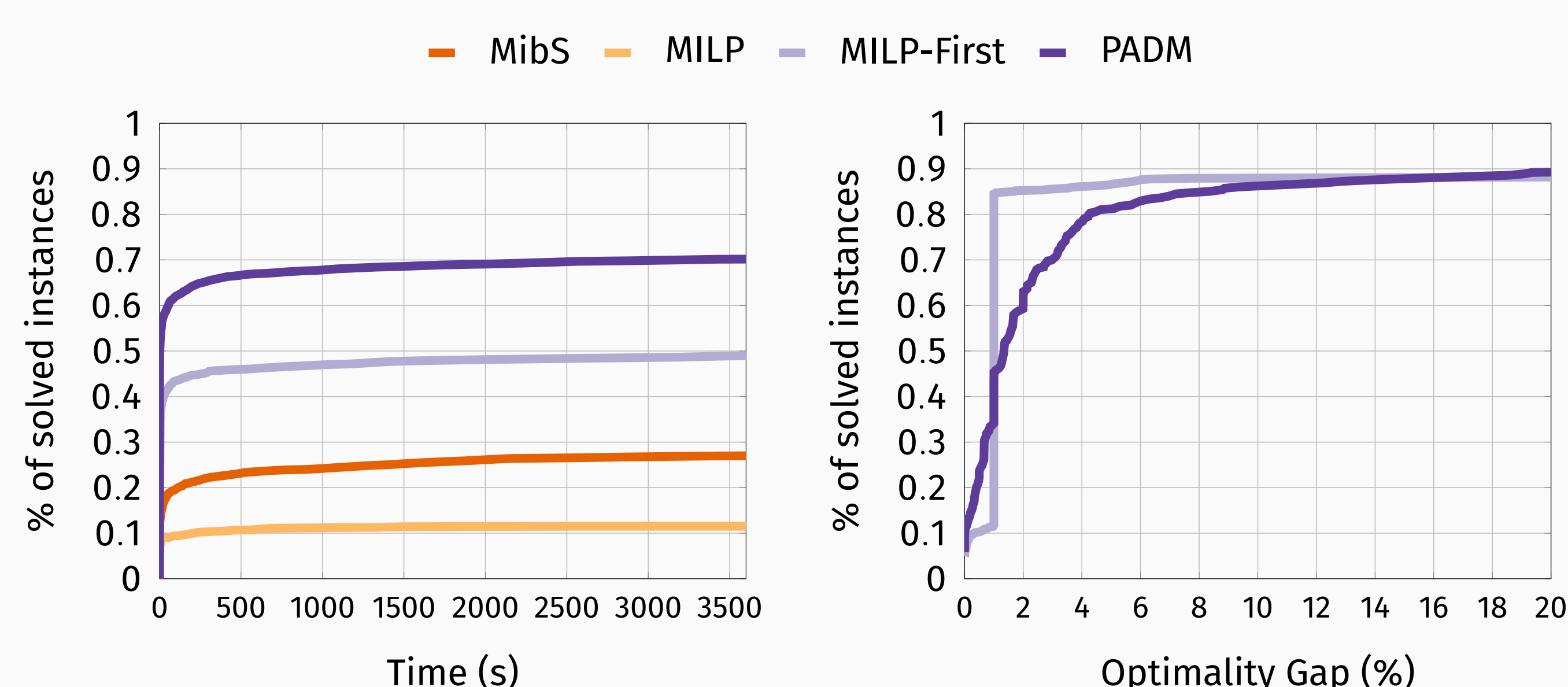
with  $\rho \nearrow \infty$

$$\begin{aligned} \min_{x \in X, y \in Y} \quad & c^\top x + d^\top y + \rho(f^\top y - \bar{\pi} - x^\top \bar{\lambda}) \\ \text{s.t.} \quad & Ax + By \geq a, \quad Cx + Dy \geq b, \end{aligned}$$

$$(\bar{x}, \bar{y}) \left( \begin{array}{c} \max_{\lambda, \pi} \left\{ \pi + \lambda^\top \bar{x} : \pi + \lambda^\top \hat{x}^j \leq f^\top \hat{y}^j, j = 1, \dots, \ell \right\} \end{array} \right) (\bar{\lambda}, \bar{\pi})$$

- ▶ The second sub-problem is solved by cut generation.
- ▶ It reduces to compute  $\varphi(x)$  and  $\bar{\lambda} \in \partial \text{vex}_X(\varphi)(\bar{x})$ .

## Numerical Results on Instances from the BOBILib



## Conclusion

- ▶ A new single-level reformulation of MILP bilevel problems
  - ▶ Can be solved as a MILP via cut generation.
  - ▶ Encouraging results.
- ▶ Two heuristic approaches
  - ▶ Both returning high-quality feasible points.
- ▶ Future Work
  - ▶ Improve bounds on the dual variables  $\lambda$  and  $\pi$ .
  - ▶ Combine this approach with existing techniques like intersection cuts (Fischetti et al., 2018).

