

Exploiting Dual Functions in Mixed Integer Bilevel Linear Programs

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Mixed Integer Linear Bilevel Optimization

- **Bilevel Problems** (BP) involve optimizing the strategy of a **leader** and the subsequent reaction of a **follower**:

$$\begin{aligned} \min \quad & cx + d^1y \\ \text{s.t.} \quad & A^1x + G^1y \geq b^1 \\ & x \in X \\ & y \in \operatorname{argmin} \{d^2y : G^2y \geq b^2 - A^2x, y \in Y\} \end{aligned} \tag{BP}$$

- Stackelberg game: two-players sequential game
- **Leader's** decision must take into account **follower's** optimal reaction
- $x \in X \subseteq \mathbb{Z}_+^{r_1} \times \mathbb{R}_+^{n_1-r_1}$ are controlled by the **leader**
- $y \in Y \subseteq \mathbb{Z}_+^{r_2} \times \mathbb{R}_+^{n_2-r_2}$ are controlled by the **follower**

Input Data

1. Upper Level Problem (ULP): $c \in \mathbb{Q}^{n_1}, d^1 \in \mathbb{Q}^{n_2}, A^1, G^1 \in \mathbb{Q}^{m_1 \times n_1}, b^1 \in \mathbb{Q}^{m_1}$
2. Second Level Problem (SLP): $d^2 \in \mathbb{Q}^{n_2}, A^2, G^2 \in \mathbb{Q}^{m_2 \times n_1}, b^2 \in \mathbb{Q}^{m_2}$

Value Function Reformulation

Definition

Let $\phi : \mathbb{R}^{m_2} \rightarrow \mathbb{R} \cup \{\pm\infty\}$ be a function such that

$$\phi(\beta) = \min \{d^2 y : G^2 y \geq \beta, y \in Y\}.$$

We assume

$$\phi(\beta) = +\infty, \text{ if (SLP) is } \mathbf{infeasible},$$

$$\phi(\beta) = -\infty, \text{ if (SLP) is } \mathbf{unbounded},$$

for some $\beta \in \mathbb{R}^{m_2}$.

- Then (BP) becomes:

$$\min cx + d^1 y$$

$$\text{s.t. } A^1 x + G^1 y \geq b^1$$

$$A^2 x + G^2 y \geq b^2$$

$$d^2 y \leq \phi(b^2 - A^2 x)$$

$$x \in X, y \in Y$$

(BP-VF)

On Solving MILBP

Relaxation

In a Branch-and-Bound-or-Cut framework, an LP relaxation of (BP-VF) is considered at each node by dropping:

- 1 x and y integrality
- 2 Second level optimality $d^2y \leq \phi(b^2 - A^2x)$

Bilevel feasibility

Checking feasibility of an integer point (\hat{x}, \hat{y}) involves the evaluation of

$$d^2\hat{y} \leq \phi(b^2 - A^2\hat{x})$$

which is an **NP-Hard** task.

Complexity

(BP) is Σ_2^P -hard \Rightarrow Given an oracle for $\phi(\beta)$, (BP) requires **nondeterministic polynomial time** to be solved.

About this Work

We consider two open-source solvers:

MibS [Tahernejad, Ralphs, and DeNegre 2020]

- Branch-and-Cut for (BP)
- Checking feasibility using MILP solvers

SYMPHONY [T. K. Ralphs and Güzelsoy 2005]

- Implementing Branch-and-Cut solver for MILPs (SLP)
- Warm-starting capabilities for RHS changes

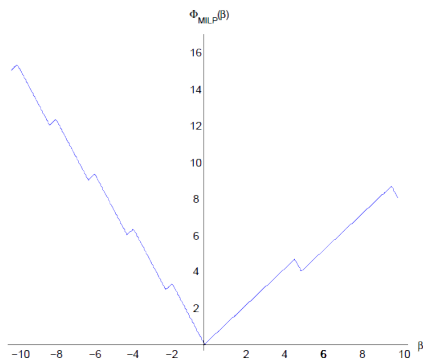
Our contribution

- ① How to construct an iteratively refined approximation of ϕ using SYMPHONY
- ② How to use it to efficiently evaluate $d^2y \leq \phi(b^2 - A^2x)$ to improve the feasibility check

On the MILP Value Function

Let us consider the following (SLP):

$$\begin{aligned}\phi(\beta) = \min \quad & 6y_1 + 4y_2 + 3y_3 + 4y_4 + 5y_5 + 7y_6 \\ \text{s.t.} \quad & 2y_1 + 5y_2 - 2y_3 - 2y_4 + 5y_5 + 5y_6 = \beta \\ & y_1, y_2, y_3 \in \mathbb{Z}_+, y_4, y_5, y_6 \in \mathbb{R}_+\end{aligned}$$



Some properties

ϕ of a MILP is:

- Lower semi-continuous
- Subadditive
- Piecewise polyhedral

On the MILP Value Function

A description of ϕ

Let I and C index the integer and continuous (resp.) variables in (SLP). As noted in [T. K. Ralphs and Hassanzadeh 2014], we have that

$$\phi(\beta) = \min_{y_I \in \mathbb{Z}_+^{r_2}} \{ \phi_I(G_{2I}y_I) + \phi_C(\beta - G_{2I}y_I) \}.$$

Integer Restriction:

$$\phi_I(\gamma) = \min_{y_I \in \mathbb{Z}_+^{r_2}} \{ d_{2I}^\top y_I \mid G_{2I}y_I = \gamma \}$$

Continuous Restriction:

$$\phi_C(\gamma) = \min_{y_C \in \mathbb{R}_+^{r_2 - r_1}} \{ d_{2C}^\top y_C \mid G_{2C}y_C = \gamma \}$$

Theorem [T. K. Ralphs and Hassanzadeh 2014]

Under the assumption that $\{ \beta \in \mathbb{R}^{m_2} \mid \phi_I(\beta) < \infty \}$ is finite, there exists a finite set $\mathcal{S} \subset \mathbb{Z}_+^{r_2}$ such that

$$\phi(\beta) = \min_{y_I \in \mathcal{S}} \{ \phi_I(G_{2I}y_I) + \phi_C(\beta - G_{2I}y_I) \}.$$

Lower Approximations of ϕ

Dual functions [Hassanzadeh and T. K. Ralphs 2014]

A function $f : \mathbb{R}^{m_2} \rightarrow \mathbb{R} \cup \{\pm\infty\}$ is said to be dual to ϕ if

$$f(\beta) \leq \phi(\beta), \quad \beta \in \mathbb{R}^{m_2}.$$

Moreover, f is strong at $\hat{\beta} \in \mathbb{R}^{m_2}$ is $f(\hat{\beta}) = \phi(\hat{\beta})$.

Strong dual functions usages examples:

- warm-start MILP resolution for different RHS $\beta' \in \mathbb{R}^{m_2}$
- sensitivity analysis
- **optimality proofs**

Strong Dual Functions from a Branch-and-Bound Tree

- Let $\hat{\beta} \in \mathbb{R}^{m_2}$ be such that $\phi(\hat{\beta}) < \infty$
- Let T indexing the set of leaf nodes of the optimal B&B tree for (SLP)
- For $t \in T$, let
 1. $l^t, u^t \in \mathbb{Z}^{r_2}$ be the branching bounds
 2. $(\pi^t, \underline{\pi}^t, \bar{\pi}^t) \in \mathbb{R}^{m_2+2n_2}$ be the optimal dual solution
- Then the LP primal-dual pair is

$$\min_{y \in \mathbb{R}_+^{n_2}} \{d^2 y \mid G^2 y = \hat{\beta}, l^t \leq y \leq u^t\} \quad (P^t)$$

$$\max_{(\pi, \underline{\pi}, \bar{\pi}) \in \mathcal{D}} \{\hat{\beta} \pi + l^t \underline{\pi} - u^t \bar{\pi}\} \quad (D^t)$$

$$\mathcal{D} = \{(\pi, \underline{\pi}, \bar{\pi}) \in \mathbb{R}^{m_2+2n_2} \mid G^2 \pi + \underline{\pi} - \bar{\pi} \leq d^2, \underline{\pi} \geq 0, \bar{\pi} \geq 0\}$$

- \mathcal{D} is independent from $\hat{\beta}, l^t, u^t$, then by LP duality

$$\max_{(\pi, \underline{\pi}, \bar{\pi}) \in \mathcal{D}} \{\beta \pi + l^t \underline{\pi} - u^t \bar{\pi}\} \leq \min_{y \in \mathbb{R}_+^{n_2}} \{d^2 y \mid G^2 y = \beta, l^t \leq y \leq u^t\}, \forall \beta \in \mathbb{R}^{m_2}$$

Strong Dual Functions from a Branch-and-Bound Tree

Theorem [Wolsey 1981]

The function $\underline{\phi}$ defined by

$$\underline{\phi}(\beta) = \min_{t \in T} (\beta \pi^t + l^t \underline{\pi}^t - u^t \overline{\pi}^t) \quad \forall \beta \in \mathbb{R}^{m_2},$$

is dual to ϕ and strong at RHS $\hat{\beta}$.

Warm-starting with SYMPHONY at a glance

Given a B&B tree evaluating $\phi(\hat{\beta})$, we can warm-start the solution of $\phi(\tilde{\beta})$ by:

- 1 Re-optimize the LP duals each at leaf nodes $t \in T$
- 2 Keep branching on nodes with fractional solutions (if any)

Remark

The warm-started B&B encodes a dual solution strong at $\tilde{\beta}$, but it **might not** be strong at $\hat{\beta}$ anymore. (Dual solutions at leaf nodes may change).

Maintaining a Strong Dual Function

- A stronger dual function can be obtained considering dual solutions from **all nodes** in the tree
- Given any B&B tree, let:
 - ① $D \subseteq \mathcal{D}$ be its set of dual solutions generated over all nodes in the B&B (including re-optimizations)
 - ② T be the set of the leaf nodes

A stronger dual function

$$\underline{\phi}^+(\beta) = \min_{t \in T} \max_{(\pi, \underline{\pi}, \bar{\pi}) \in D} (\beta^\top \pi + l^t \underline{\pi} - u^t \bar{\pi})$$

- This new dual function is strong at all RHSs previously solved and

$$\underline{\phi}(\beta) \leq \underline{\phi}^+(\beta) \leq \phi(\beta)$$

Evaluating $\underline{\phi}^+(\beta)$

Given any $\beta \in \mathbb{R}^{m_2}$, let us consider the following matricial representation of $\underline{\phi}^+$:

$$M_D = \begin{bmatrix} \pi^{d_1} & \underline{\pi}^{d_1} & \overline{\pi}^{d_1} \\ \pi^{d_2} & \underline{\pi}^{d_2} & \overline{\pi}^{d_2} \\ \vdots & \vdots & \vdots \\ \pi^{d_{|D|}} & \underline{\pi}^{d_{|D|}} & \overline{\pi}^{d_{|D|}} \end{bmatrix} \quad N_T = \begin{bmatrix} \beta & \beta & \dots & \beta \\ |^{t_1} & |^{t_2} & \dots & |^{t_{|T|}} \\ u^{t_1} & u^{t_2} & \dots & |^{t_{|T|}} \end{bmatrix}$$

Then $\underline{\phi}^+(\beta)$ can be evaluated with the following operation:

$$M_D \cdot N_T = \begin{bmatrix} \beta\pi^{d_1} + |^{t_1}\underline{\pi}^{d_1} - u^{t_1}\overline{\pi}^{d_1} & \dots & \beta\pi^{d_1} + |^{t_{|T|}}\underline{\pi}^{d_1} - u^{t_{|T|}}\overline{\pi}^{d_1} \\ \beta\pi^{d_2} + |^{t_1}\underline{\pi}^{d_2} - u^{t_1}\overline{\pi}^{d_2} & \dots & \beta\pi^{d_2} + |^{t_{|T|}}\underline{\pi}^{d_2} - u^{t_{|T|}}\overline{\pi}^{d_2} \\ \vdots & & \vdots \\ \beta\pi^{d_{|D|}} + |^{t_1}\underline{\pi}^{d_{|D|}} - u^{t_1}\overline{\pi}^{d_{|D|}} & \dots & \beta\pi^{d_{|D|}} + |^{t_{|T|}}\underline{\pi}^{d_{|D|}} - u^{t_{|T|}}\overline{\pi}^{d_{|D|}} \end{bmatrix}$$

$$\underbrace{\begin{matrix} \downarrow & & \downarrow \\ \max(\cdot) & \dots & \max(\cdot) \end{matrix}}_{\min(\cdot) = \underline{\phi}^+(\beta)}$$

Checking Bilevel Feasibility Using $\underline{\phi}^+$

Algorithm 1

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1: Given  $(\hat{x}, \hat{y}) \in X \times Y$ , a B&B Tree and matrices  $M_D, N_T$  then
2: Evaluate  $\underline{\phi}^+(b^2 - A^2\hat{x})$  using  $M_D, N_T$ 
3: if  $d^2\hat{y} \leq \underline{\phi}^+(b^2 - A^2\hat{x})$  then
4:   return  $(\hat{x}, \hat{y})$  is Bilevel Feasible
5: else
6:   Warm-start the evaluation of  $\phi(b^2 - A^2\hat{x})$  using SYMPHONY
7:   B&B Tree is updated with new leaf nodes  $T'$  and (possibly new) duals  $D'$ 
8:   Update  $M_D, N_T$  with  $M_{D'}, N_{T'}$ 
9:   if  $d^2\hat{y} \leq \phi(b^2 - A^2\hat{x})$  then
10:    return  $(\hat{x}, \hat{y})$  is Bilevel Feasible
11:   else
12:    return  $(\hat{x}, \hat{y})$  is not Bilevel Feasible
13:   end if
14: end if
```

Conclusions

Remarks

- The dimension of M_D, N_T should be kept as small as possible
 - Duplicate Dual solutions should be avoided \rightarrow Hash Table
 - Some Dual solutions may become dominated and may be safely discarded
- Algorithm 1 is iteratively building parts of ϕ needed to solve (BP)

Takeaways

- Dual functions can provide optimality proofs for (SLP)
- SYMPHONY's warm-starting provides stronger and stronger dual functions
- Effectiveness of this approach is an empirical question

Roadmap

- ① Refine the implementation
- ② Parameter tuning via numerical experience