

BILEVEL OPTIMIZATION UNDER UNCERTAINTY

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BASED ON OUR RECENT ARTICLES:

A Survey on Mixed-Integer Programming Techniques in Bilevel Optimization
EURO Journal on Computational Optimization. 2021. DOI: 10.1016/j.ejco.2021.100007
Jointly with Thomas Kleinert, Martine Labbé, and Martin Schmidt

A Survey on Bilevel Optimization Under Uncertainty
Jointly with Yasmine Beck and Martin Schmidt, Optimization Online, 2022

A Brief Introduction to Robust Bilevel Optimization
Jointly with Yasmine Beck and Martin Schmidt,
Views-and-News of the SIAM Activity Group on Optimization, to appear 2022

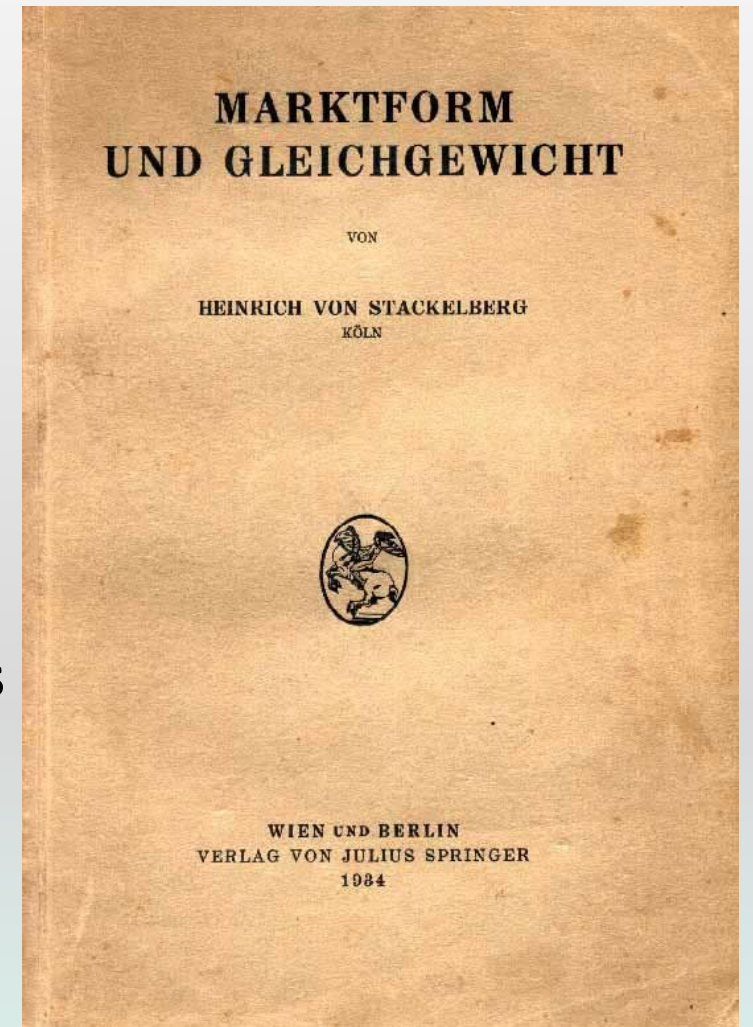


BILEVEL OPTIMIZATION

WITH DETERMINISTIC DATA

STACKELBERG GAMES

- Introduced in economy by H. v. Stackelberg in 1934
- Two-player sequential game: LEADER and FOLLOWER
- The LEADER moves before the FOLLOWER
- **Perfect information:** the leader has a perfect knowledge of the followers strategy
- The follower observes leader's action and acts rationally
- **Rationality:** agents act optimally, maximizing their payoffs
- BILEVEL OPTIMIZATION: Bracken & McGill (1973), Candler & Norton (1977)



APPLICATIONS: PRICING

Pricing: operator sets tariffs, and then customers choose the cheapest alternative

- **Tariff-setting, toll optimization (Labbé et al., 1998; Brotcorne et al., 2001; Labbé & Violin, 2016)**
- **Network Design and Pricing (Brotcorne et al., 2008)**
- **Survey (van Hoesel, 2008)**

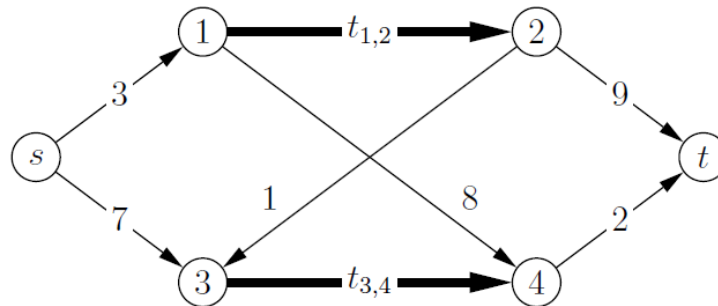


Figure 1: 1-commodity network with two tariff arcs.



A DETERMINISTIC BILEVEL PROBLEM

$$\text{“min”}_{x \in X} F(x, y) \quad (1a)$$

$$\text{s.t. } G(x, y) \geq 0, \quad (1b)$$

$$y \in S(x), \quad (1c)$$

upper
level

$S(x)$: the set of optimal solutions of the x -parameterized problem

$$\min_{y \in Y} f(x, y) \quad (2a)$$

$$\text{s.t. } g(x, y) \geq 0. \quad (2b)$$

lower
level

- Both levels may involve integer decision variables. Functions can be non-linear, non-convex...
- (1) could be ill-posed (if LL solution is not unique). “min” to be replaced by

$$\min_{x \in \bar{X}} \min_{y \in S(x)} F(x, y)$$



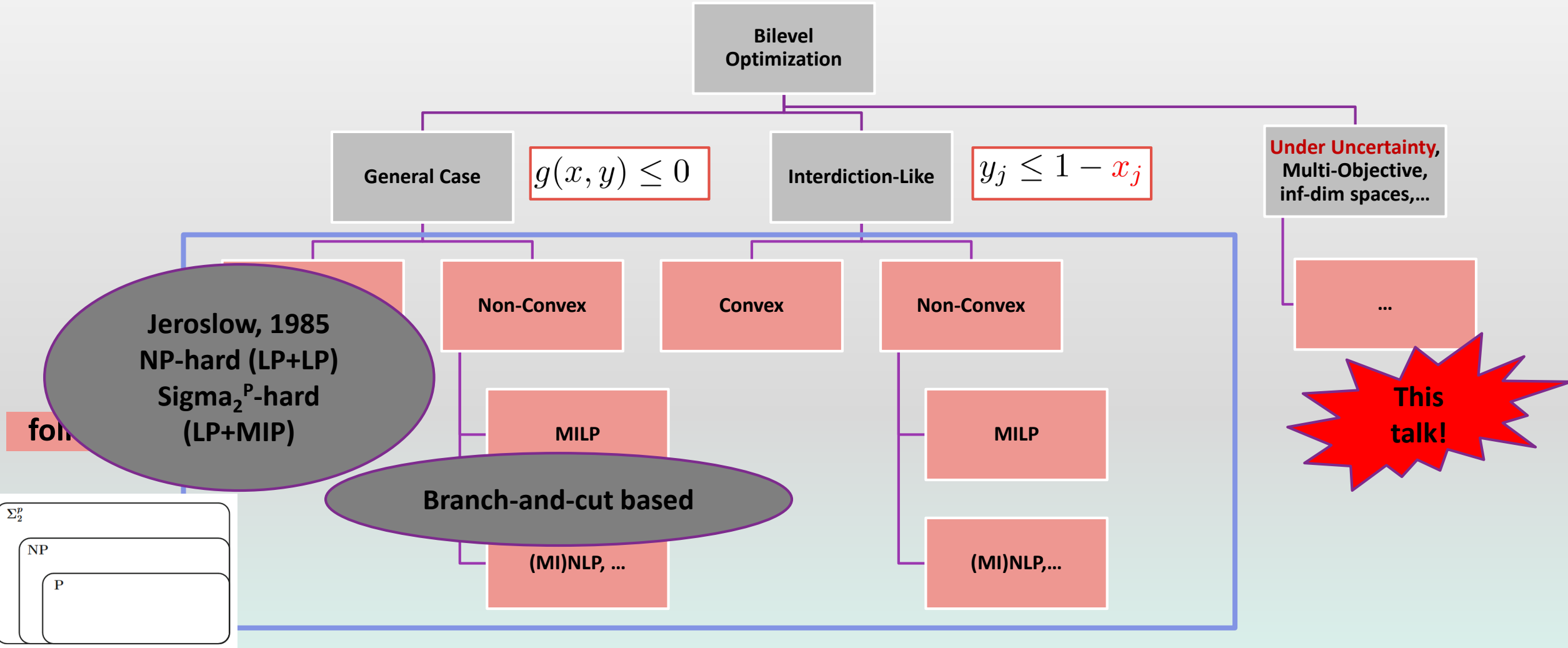
Optimistic!

$$\min_{x \in \bar{X}} \max_{y \in S(x)} F(x, y)$$



Pessimistic!

OVERVIEW OF BILEVEL OPTIMIZATION PROBLEMS



THIS TALK

- **From deterministic bilevel optimization to bilevel optimization under uncertainty**
- **Sources of uncertainty**
 - **Data uncertainty**
 - **Decision uncertainty**
- **Timing for the data uncertainty**
 - **Here-and-now follower**
 - **Wait-and-see follower**
- **Challenges & opportunities**

SOURCES OF UNCERTAINTY

UNCERTAINTY: SINGLE-LEVEL VS BILEVEL

Single-level optimization:

$$\min_x \{c^\top x : Ax \geq b\}$$

- “Only” subject to data uncertainty in A, b, c
- Stochastic optimization
- Robust optimization
- Distributionally robust, etc

Bilevel optimization:

$$\text{“min”}_{x \in X} F(x, y) \quad (1a)$$

$$\text{s.t. } G(x, y) \geq 0, \quad (1b)$$

$$y \in S(x), \quad (1c)$$

$S(x)$: optimal solutions of the x -parameterized problem

$$\min_{y \in Y} f(x, y) \quad (2a)$$

$$\text{s.t. } g(x, y) \geq 0. \quad (2b)$$

- Subject to: **data uncertainty**
- But also: **decision uncertainty**. The leader is not sure about the reaction of the follower, or the follower is not certain about the observed leader's decision.

TIMING OF UNCERTAINTY

WAIT-AND-SEE FOLLOWER

leader x \curvearrowright uncertainty u \curvearrowright follower $y = y(x, u)$.

The leader is uncertain about the optimization parameters of the follower

Example: the leader solves a robust optimization problem

$$\text{“min}_{x \in X} \max_{u \in \mathcal{U}}” \quad F(x, y) \quad \text{s.t.} \quad y \in S(x, u),$$

$$S(x, u) := \arg \min_{y \in Y} \quad f(x, u, y) \quad \text{s.t.} \quad g(x, u, y) \geq 0.$$

Example: the leader is risk-neutral wrt data uncertainty (discrete scenario set)

Optimistic or pessimistic leader

$$\text{“min}_{x \in X}” \quad \sum_{u \in \mathcal{U}} p_u F(x, y(x, u)) \quad \text{s.t.} \quad y(x, u) \in S(x, u), u \in \mathcal{U}$$

$$S(x, u) := \arg \min_{y \in Y} \quad f(x, u, y) \quad \text{s.t.} \quad g(x, u, y) \geq 0.$$

HERE-AND-NOW FOLLOWER

leader x \curvearrowright follower $y = y(x)$ \curvearrowright uncertainty u .

The follower solves the problem under data uncertainty (stochastic, robust,...).

Optimistic vs pessimistic leader

For example: optimistic leader, the robust follower hedges against uncertainty in the objective function

$$\min_{x \in X} \min_{y \in S(x)} F(x, y)$$

$$S(x) := \arg \min_{y' \in Y} \left\{ \max_{u \in \mathcal{U}} f(x, u, y') : g(x, y') \geq 0 \right\}.$$

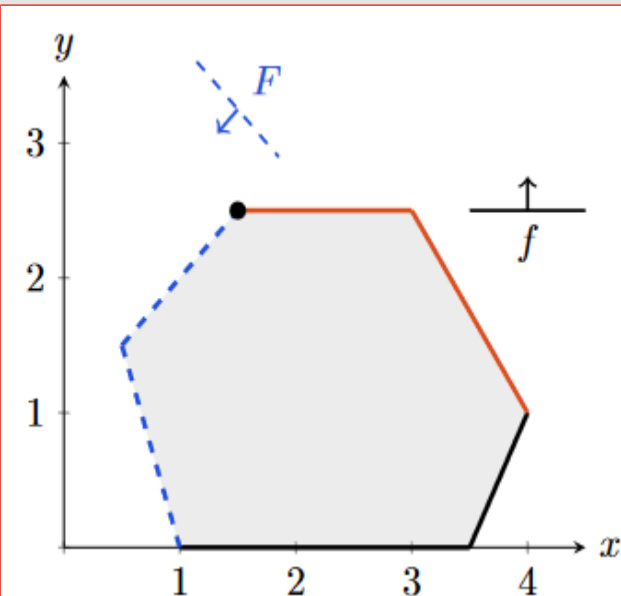
A SMALL EXAMPLE

Deterministic bilevel

$$\text{“min”}_{x \in \mathbb{R}} \quad F(x, y) = x + y$$

$$\begin{aligned} \text{s.t.} \quad & x - y \geq -1, \\ & 3x + y \geq 3, \\ & y \in S(x), \end{aligned}$$

$$\begin{aligned} S(x) := \arg \min_{y \in \mathbb{R}} \quad & f(x, y) = -0.1y \\ \text{s.t.} \quad & -2x + y \geq -7, \\ & -3x - 2y \geq -14, \\ & 0 \leq y \leq 2.5. \end{aligned}$$

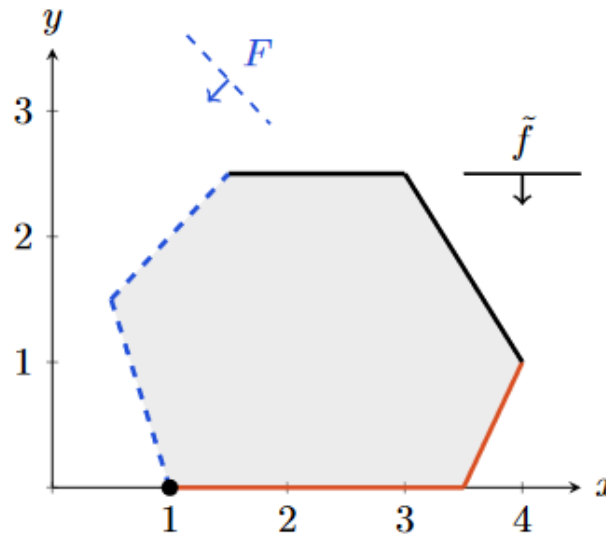


(a)

Here-and-now follower

$$\mathcal{U} := \{u \in \mathbb{R} : |u| \leq 0.5\}$$

$$\begin{aligned} S(x) := \arg \min_{y \in \mathbb{R}} \quad & \max_{u \in \mathcal{U}} \tilde{f}(x, u, y) = (-0.1 + u)y \\ \text{s.t.} \quad & -2x + y \geq -7, \\ & -3x - 2y \geq -14, \\ & 0 \leq y \leq 2.5. \end{aligned}$$



(b)

Wait-and-see follower

$$\text{“min”}_{x \in \mathbb{R}} \quad \max_{u \in \mathcal{U}} \quad F(x, y) = x + y$$

$$\begin{aligned} \text{s.t.} \quad & x - y \geq -1, \\ & 3x + y \geq 3, \\ & y \in S(x, u), \end{aligned}$$

$$\begin{aligned} S(x, u) := \arg \min_{y \in \mathbb{R}} \quad & (-0.1 + u)y \\ \text{s.t.} \quad & -2x + y \geq -7, \\ & -3x - 2y \geq -14, \\ & 0 \leq y \leq 2.5. \end{aligned}$$

$$\min_x \quad \hat{F}(x) \quad \text{s.t.} \quad 1.5 \leq x \leq 4$$

$$\hat{F}(x) = \begin{cases} x + 2.5, & 1.5 \leq x \leq 3, \\ -0.5x + 7, & 3 \leq x \leq 4. \end{cases}$$

$$u \in [-0.5, 0.1) \Rightarrow (a)$$

$$u \in (0.1, 0.5] \Rightarrow (b)$$

$$u = 0.1 \Rightarrow \text{all LL sol feasible}$$

CHALLENGES

PROBLEM COMPLEXITY

Robust **single-level** LPs:

- Interval, ball, ellipsoidal, polyhedral or Gamma-uncertainty preserve “tractability” of their deterministic counterpart (Ben-Tal & Nemirovski, Bertsimas & Sim)

$$\begin{aligned} \min \quad & c^T x \\ \text{s.t.} \quad & (a + u)^T x \leq b \text{ for all } u \in \mathcal{U} \end{aligned}$$

Robust **bilevel** optimization:

- Here-and-now follower**: tractability of the lower-level remains preserved for these uncertainty types
- Continuous convex lower level: KKT-based, strong duality-based reformulations still possible
- Discrete lower level: branch-and-cut still possible
- Major challenge: much larger in size, parallelization

Robust bilevel optimization:

- Wait-and-see follower**: the problems may climb up in the complexity hierarchy!

ROBUST BILEVEL OPTIMIZATION

Deterministic bilevel

$$\begin{aligned} & \text{“max”}_{x \in X} \quad d^T y \\ & \text{s.t. } y \in S(x) \\ & S(x) := \arg \max \{ \textcolor{red}{u}^T y : Ay \leq Bx + b \} \\ & X \subseteq \{0, 1\}^{n_x} \end{aligned}$$

NP-hard

Robust bilevel: Wait-and-see follower

$$\begin{aligned} & \text{“max}_{x \in X} \quad \min_{\textcolor{red}{u} \in U} \quad d^T y \\ & \text{s.t. } y \in S(x, \textcolor{red}{u}) \\ & S(x, \textcolor{red}{u}) := \arg \max \{ \textcolor{red}{u}^T y : Ay \leq Bx + b \} \\ & X \subseteq \{0, 1\}^{n_x} \end{aligned}$$

$$\mathcal{U} := [u_1^-, u_1^+] \times \cdots \times [u_{n_y}^-, u_{n_y}^+]$$

Under interval uncertainty, the robust counterpart is **Sigma₂^P-hard**
The “adversarial problem” (inner min) is NP-hard

Buchheim, Henke, Hommelsheim:

On the complexity of robust bilevel optimization with uncertain follower's objective. OR Letters 49(5): 703-707 (2021)

OPPORTUNITIES

BILEVEL STOCHASTIC MIP

Discrete scenario set

$$\begin{aligned} \min_{x \in X, y} \quad & c^T x + \sum_{u \in \mathcal{U}} p_u \, d_L^T y(x, u) \\ \text{s.t.} \quad & y(x, u) \in \arg \min_{y \in Y} \{d_F^T y : Ay \leq B_u x + b_u\} \\ & X \subseteq \{0, 1\}^{n_x}, Y \subseteq \{0, 1\}^{n_y} \end{aligned}$$

Value function:

$$\Phi(x, u) = \min_{y \in Y} \{d_F^T y : Ay \leq B_u x + b_u\}$$

Value-function reformulation (optimistic)

$$\begin{aligned} \min_{x \in X, y} \quad & c^T x + \sum_{u \in \mathcal{U}} p_u \, d_L^T y_u \\ \text{s.t.} \quad & d_F^T y_u \leq \Phi(x, u), \quad u \in \mathcal{U} \\ & Ay_u \leq B_u x + b_u, \quad u \in \mathcal{U} \\ & y_u \in Y, \quad u \in \mathcal{U} \\ & X \subseteq \{0, 1\}^{n_x}, Y \subseteq \{0, 1\}^{n_y} \end{aligned}$$

Single-leader, multiple independent followers

**Leverage on the existing branch-and-cut methods
(Fischetti et al, 2017; Tahernejad et al, 2020)**

CRITICISM... AND OUTLOOK

PERFECT INFORMATION AND RATIONALITY OF DECISION MAKERS

DECISION UNCERTAINTY: EXAMPLES

- Leader hedges against sub-optimal follower reactions → **near-optimal robust bilevel models** (Besancon et al, 2019).
- If the level of cooperation/confrontation of the follower is unknown → **intermediate cases, between the optimistic and the pessimistic one** (Aboussoror & Loridan, 1995; Mallozzi & Morgan, 1996).
- The follower cannot perfectly observe the decision of the leader → hedges against all possible leader decision given the noisy observation (Bagwell, 1995; vanDamme & Hurkens, 1997; Beck & Schmidt: 2021).
- Limited intellectual or computational resources render it impossible for the follower to take a globally optimal decision → the **follower resorts to heuristic approaches** and the leader may be uncertain w.r.t. which heuristic is used (Zare et al, 2020).

CONCLUSIONS

- **Connections between bilevel and robust/stochastic optimization still to be better understood**
- **When can we retain the tractability of the deterministic bilevel counterpart?**
- **When can we solve uncertain bilevel problems through a series of deterministic ones?**
- **When do the bilevel problems under uncertainty become significantly harder?**
- **How can we better exploit the existing computational frameworks for deterministic bilevel optimization? (decomposition, SAA, scenario aggregation...)**
- **Data uncertainty vs Decision uncertainty, which paradigm to follow?**



LITERATURE

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