

On Bi-level approach for Scheduling problems

New challenges in Scheduling Theory – Aussois 2016

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Context

- ICT technology is evolving very fast → **Smart ICT**
- Decisions are generally not controlled by a **single decision maker**
- e.g. Cloud Computing, Internet of Things involve several decision makers
- Bi-level optimization involves **2** decision makers
- Equivalent to Hierarchical games (e.g. Stackelberg games) → **iterative games**

Bi-level problems

- Two **nested** optimization levels
- First level is defined as *"The upper level"* or *"The leader problem"*
- Second level is defined as *"The lower level"* or *"The follower problem"*
- Decision variables partitioned into **two** sets
- Decision maker can only act on their decision variables
- **However** they can act **indirectly** on each other
- The follower problem is **parametrized** by the leader decision
- The **feasibility** of the leader problem depends on the **optimality** of the follower problem

Bi-level formulation

$$\begin{aligned} & \min_{x, y' \in Y(x)} F(x, y') \\ & \text{s.t. } G(x, y') \leq 0 \\ & Y(x) = \operatorname{argmin}_{y \in Y} f(x, y) \\ & \text{s.t. } g(x, y) \leq 0 \\ & x, y \geq 0 \end{aligned}$$

where

- F: leader objective $\mathbf{R}^n \times \mathbf{R}^m \rightarrow \mathbf{R}$
- f: follower objective $\mathbf{R}^n \times \mathbf{R}^m \rightarrow \mathbf{R}$
- G: leader's constraints $G: \mathbf{R}^n \times \mathbf{R}^m \rightarrow \mathbf{R}^p$
- g: follower's constraints $\mathbf{R}^n \times \mathbf{R}^m \rightarrow \mathbf{R}^q$

Rational reaction set

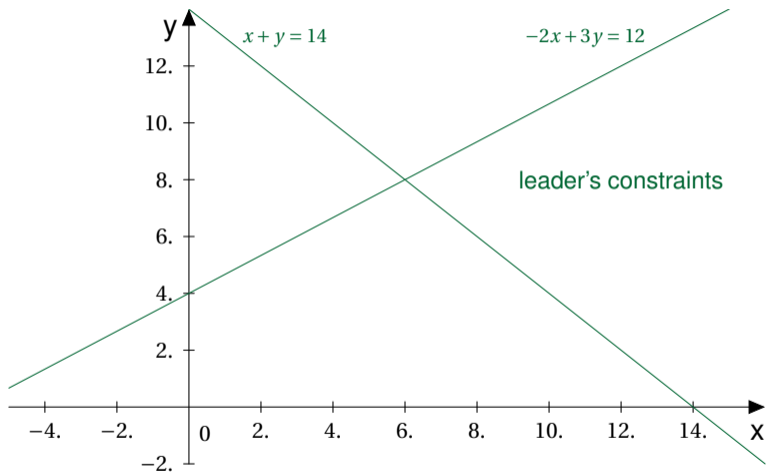
- The follower's feasible set parametrized by $x \in X$: $S(x) = \{y \in Y : g(x, y) \leq 0\}$.
- The Follower's rational decision set:

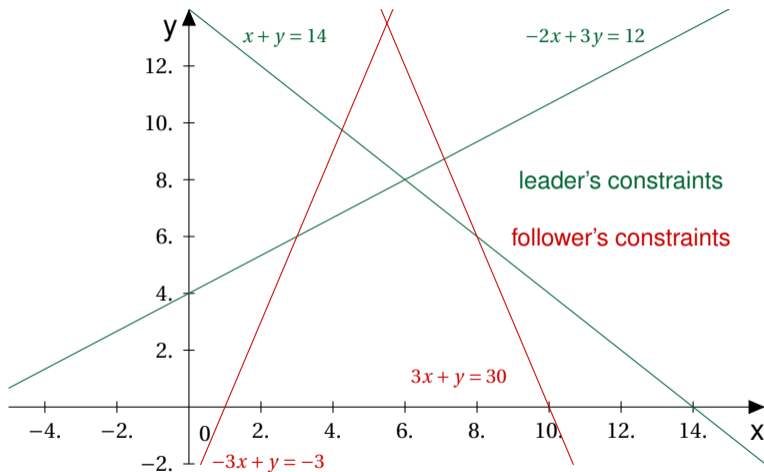
$$Y(x) = \{y' \in Y : y' \in \operatorname{argmin}[f(x, y) : y \in S(x)]\}$$
- The Inducible Region: $IR = \{(x, y') \in S, y' \in Y(x)\}$
- For a given x , the follower reacts **optimally** (if possible) to leader decisions $\rightarrow Y(x)$
- For a given x , the follower may have **several** optimal reactions (solutions) $\rightarrow |Y(x)| > 1$
- For a given x , the follower is **indifferent** to each $y' \in Y(x)$. This is **not** the case for the Leader
- If $|Y(x)| > 1$ then two cases:
 - **Optimistic** model: " $\min_{x, y' \in Y(x)}$ " \rightarrow $\min_{x, y' \in Y(x)}$
 - **Pessimistic** model: " $\min_{x, y' \in Y(x)}$ " \rightarrow $\min_x \max_{y' \in Y(x)}$

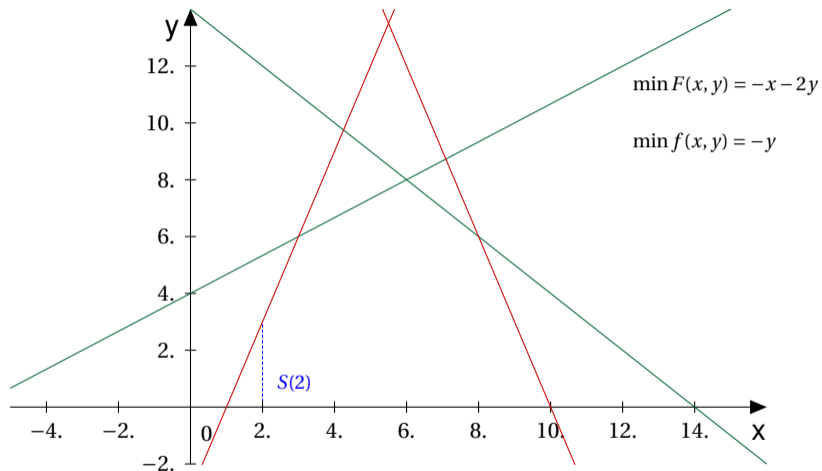
Example

Dempe et al. 2005: optimistic case

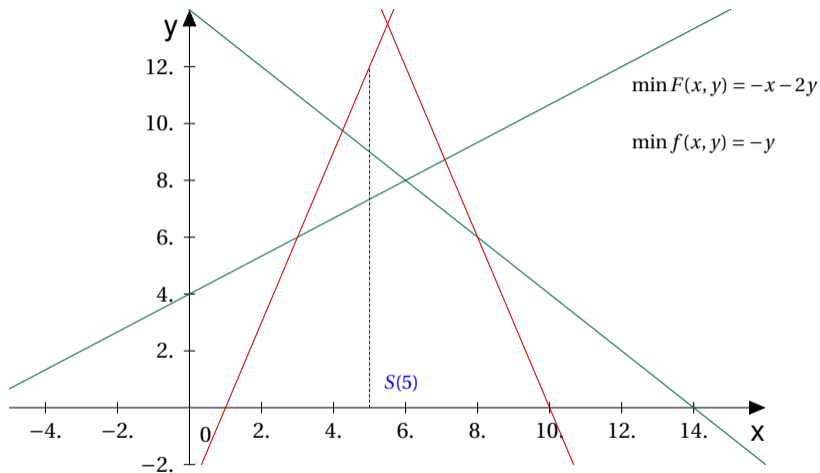
$$\begin{aligned} P = \min_{x \geq 0, y' \in Y(x)} \quad & F(x, y) = -x - 2y' \\ \text{s.t.} \quad & 2x - 3y' \geq -12 \\ & x + y' \leq 14 \\ & Y(x) = \operatorname{argmin}_{y \geq 0} \quad f(y) = -y \\ & \text{s.t.} \quad -3x + y \leq -3 \\ & \quad \quad 3x + y \leq 30 \end{aligned}$$





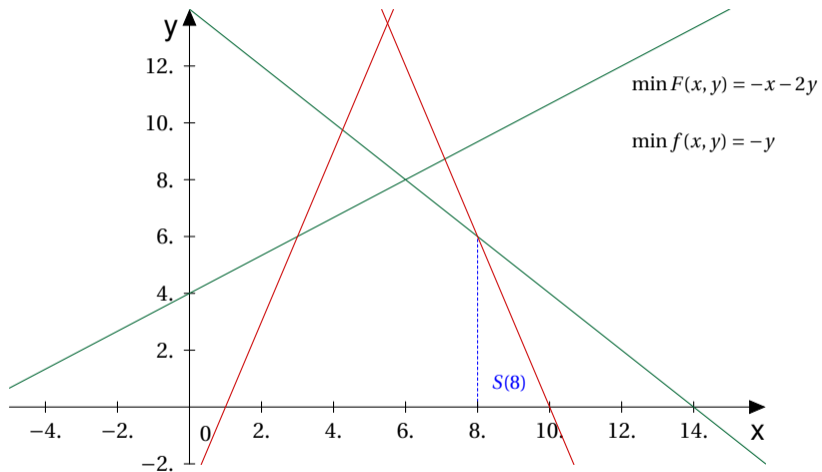


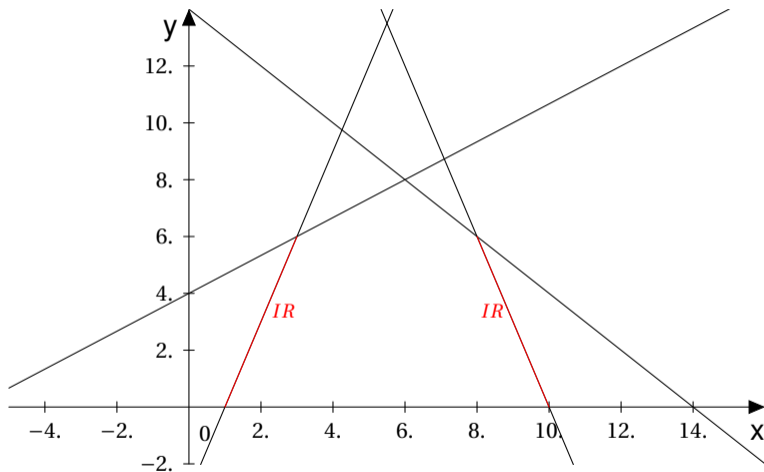
The blue dashed line is the parametrized follower decision set for $x = 2$



⚠ The follower is indifferent to leader's constraints

⚠ $Y(5)$ is not feasible for the leader





State of the Art I

Dempe et al. 2005

Even for convex bi-level problems:

- The decision set may be not convex
- The decision set may be discontinued

Jeroslow et al. 1985

- The linear bi-level problem has been shown NP-hard

Resolution in continuous case

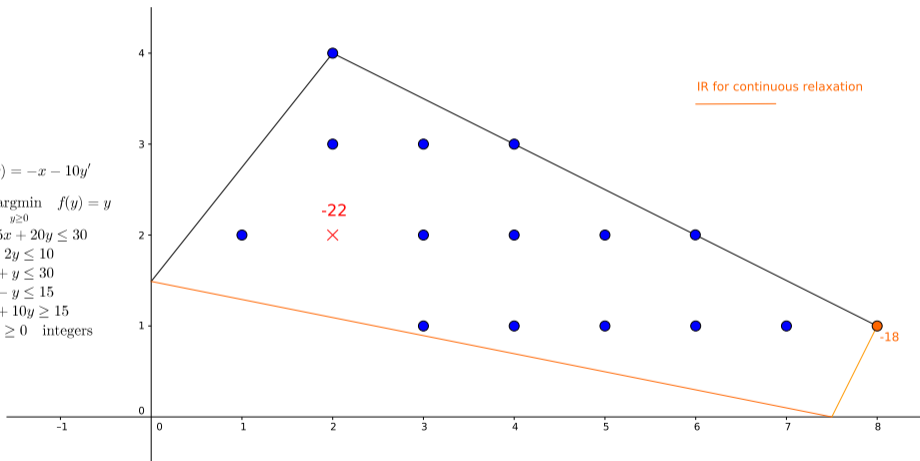
- The follower problem is mostly replaced by its Karush-Kuhn-Tucker conditions
- The resulting single-level problem is solved using non-linear optimization algorithms

State of the Art II

Resolution in discrete case

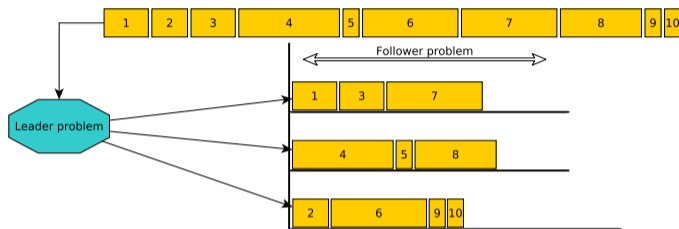
- No transformations
- Solution of relaxed bi-level problems does not provide valid bounds
- An integer solution (found using a Branch & Bound) is not necessary bi-level feasible
- Integer solutions does not generate sterile nodes

$$\begin{aligned}
 P = \min_{x \geq 0, y' \geq 0} & \quad F(x, y) = -x - 10y' \\
 \text{s. t. } & \quad Y(x) = \operatorname{argmin}_{y \geq 0} f(y) = y \\
 \text{s. t. } & \quad -25x + 20y \leq 30 \\
 & \quad x + 2y \leq 10 \\
 & \quad 3x + y \leq 30 \\
 & \quad 2x - y \leq 15 \\
 & \quad 2x + 10y \geq 15 \\
 & \quad x, y \geq 0 \text{ integers}
 \end{aligned}$$



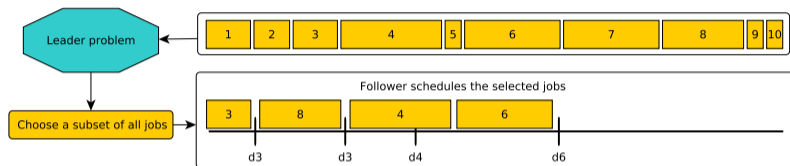
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Bi-level total weighted completion time [Kis and Kovács, 2010]



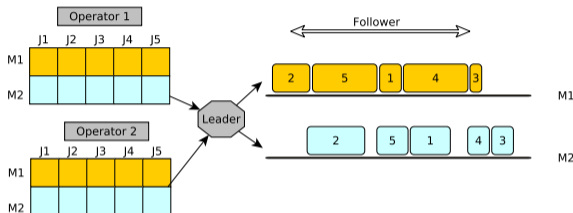
- M identical machines, N jobs
- Each jobs has a processing time p_j and two non-negative weights, w_j^1 and w_j^2
- Leader assigns jobs to machines and minimize $\sum_j w_j^1 C_j$
- Follower schedules the assigned jobs on each machine and minimize $\sum_j w_j^2 C_j$
- Decision problem is NP-complete in the strong sense ([Kis and Kovács, 2010])

Bi-level order acceptance [Kis and Kovács, 2010]



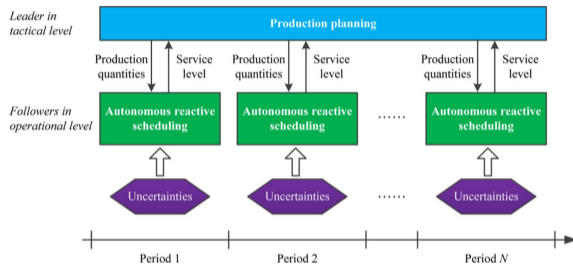
- 1 machine, N jobs
- Each jobs has a processing time p_j , a deadline d_j and two non-negative weights, w_j^1 and w_j^2
- Leader assigns jobs to machines and minimize $\sum_j w_j^1 R_j$ with $R_j = 1$ if $C_j > d_j$
- Follower schedules the assigned jobs on each machine and minimize $\sum_j w_j^2 C_j$
- Decision problem is NP-complete in the strong sense

Bi-level flow-shop problem [Karlof and Wang, 1996]



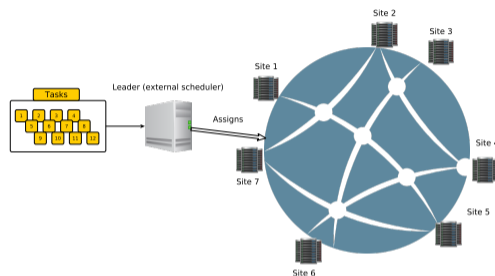
- M machine, N jobs, M operators
- Each operator has its own time table for each jobs on each machine
- Leader assigns operators to machines to minimize the total flow-time
- Follower schedules then all jobs according to the processing time provided by the assigned operators to minimize the makespan

Scheduling under production uncertainties [Chu et al., 2015]



- Planning and scheduling are two core decision intricately linked.
- Leader -- > planning problem based on customers orders
- Followers -- > scheduling problems to provide a production schedule
- Multiple products can be manufactured in a single period, the production need to be scheduled
- The scheduling problem is parametrized by the production quantity obtained by the leader

Bi-level Grid Scheduling [Bianco et al., 2015]



- Each task has a release date and a due-date
- Leader -- > External scheduler assigns tasks to grid computing sites
- Followers -- > Multiple followers with own objectives
- Leader wants to minimize a cost proportional to the tardiness
- Followers wants to maximise the computational resource usage efficiency

Thank you

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