

Bilevel formulation and two exact resolution approaches for the Maximum Capture Facility Location with Random OWA utilities

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JULY 2023



The 23rd Conference of the International Federation
of Operational Research Societies

July 10-14 • SANTIAGO, CHILE



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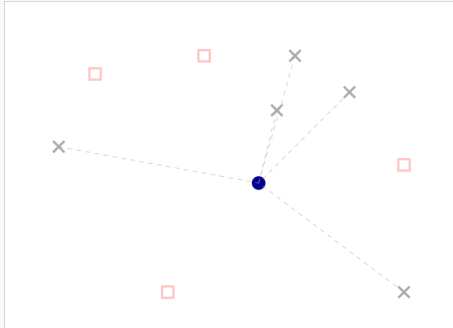


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Introduction to the Cooperative Maximum Capture Facility Location (CMCFL)

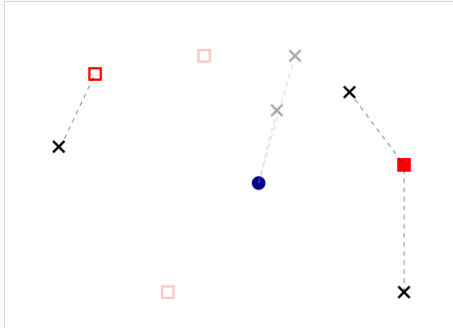
Maximum Capture Facility Location (MCFL)



Company?

Maximize customer capture by opening plants.

Maximum Capture Facility Location (MCFL)



Company?

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Customers?

Binary rule

(Utility maximization):
They patronize their favorite open plant

Cooperative MCFL. Applications?

Binary decision rule + cooperative setting

- **Cooperative capturing:** The captured utility is given in terms of an aggregation of the partial utilities.
- With **binary decision rule:** Customers are **loyal** to the company.

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Specific product:



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- **Cooperative capturing:** The captured utility is given in terms of an aggregation of the partial utilities.
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Specific product:



Membership for the service:



Modeling the captured utility

Captured utility

Ordered Median function

Assigns importance weights to the sorted partial utilities, and then aggregates the weighted utilities.

Consider a set of customers $\mathcal{I}$, a set of plants $\mathcal{J}$, and **partial utilities** u_{ij}^{ts} for each time period t and scenario s . For $i \in \mathcal{I}$, define the associated weighting vector $\lambda_i = (\lambda_{i1}, \dots, \lambda_{i|\mathcal{J}|})$. Then the **captured utility** is:

$$U_i^{ts} := \Phi_{\lambda_i}(u_{i1}^{ts}, \dots, u_{i|\mathcal{J}|}^{ts}) = \sum_{j \in \mathcal{J}} \lambda_{ij} u_{ij}^{ts},$$

where $u_{i(r)}^{ts}$ is such that $u_{i(1)}^{ts} \geq \dots \geq u_{i(|\mathcal{J}|)}^{ts}$.

$$\begin{cases} \lambda = (1, 2, 3) \\ u_i = (3, 7, 0) \end{cases} \Rightarrow U_i^{ts} = 1 \cdot 7 + 2 \cdot 3 + 3 \cdot 0 = 13.$$

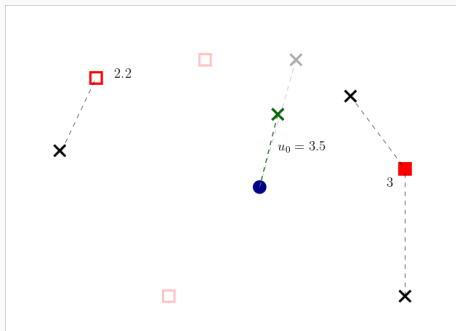
Unified cooperative framework

Different λ for different approaches:

Utility Maximization $\lambda_i = (1, 0, \dots, 0)$ (non cooperative)

Consideration set of size ℓ $\lambda_i = (\underbrace{1, \dots, 1}_{\ell}, 0, \dots, 0)$

Average $\lambda_i = (\frac{1}{|\mathcal{I}|}, \dots, \frac{1}{|\mathcal{I}|})$



Cooperative setting with
 $\lambda_i = (1, \frac{1}{2}, 0, \dots, 0)$:

Small plant: $U = 2.2$

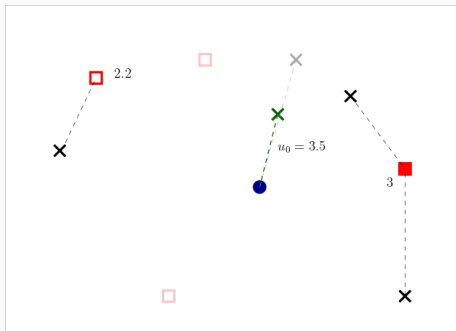
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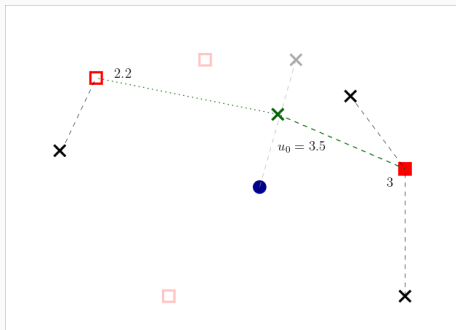
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Cooperative setting with

$\lambda_i = (1, \frac{1}{2}, 0, \dots, 0)$:

Small plant: $U = 2.2$

Big plant: $U = 3$

Both: $U = 3 \cdot 1 + \frac{2.2}{2} = 4.4$

Modeling the OMf

Variables

$$\sigma_{ijr}^{ts} := \begin{cases} 1, & u_{ij}^{ts} \text{ is the } r\text{-th largest utility for customer } i, \\ 0, & \text{otherwise.} \end{cases}$$

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The **assignment problem** is:

$$\begin{aligned} U_i^{ts} = \max_{\sigma_i^{ts}} \quad & \sum_{j \in \mathcal{J}} \sum_{r \in \mathcal{J}} \lambda_{ir} u_{ij}^{ts} \sigma_{ijr}^{ts} \\ \text{s.t.} \quad & \sum_{j \in \mathcal{J}} \sigma_{ijr}^{ts} = 1, \quad \forall r \in \mathcal{J}, \\ & \sum_{r \in \mathcal{J}} \sigma_{ijr}^{ts} = 1, \quad \forall j \in \mathcal{J}, \\ & \sum_{j \in \mathcal{J}} u_{ij}^{ts} \sigma_{ijr-1}^{ts} \geq \sum_{j \in \mathcal{J}} u_{ij}^{ts} \sigma_{ijr}^{ts}, \quad \forall r \in \mathcal{J} \setminus \{1\}, \\ & \sigma_{ijr}^{ts} \in \{0, 1\}, \quad \forall j, r \in \mathcal{J}. \end{aligned}$$

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* Assuming a non-increasing vector λ_i : $\lambda_{i1} \geq \dots \geq \lambda_{i|\mathcal{J}|}$.

Bilevel Model for the CMCFL

Cooperative Maximum Capture Facility Location

Bilevel structure

First level Company deciding the locations of the plants:

$$x_{jk}^t := \begin{cases} 1, & \text{a facility of type } k \text{ is installed in } j \text{ at time period } t, \\ 0, & \text{otherwise.} \end{cases}$$

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Second level Customers maximizing their utility:

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Partial utilities u_{ij}^{ts} (depend on the location variables x).

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Partial utilities u_{ij}^{ts} (depend on the location variables x).

Captured utility U_i^{ts} Embedded assignment problem in terms of σ , u .

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 \max_{\mathbf{x}, \mathbf{u}} \quad & \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{I}} n_i^t \frac{1}{|\mathcal{S}|} \sum_{s \in \mathcal{S}} z_i^{ts} \\
 \text{s.t.} \quad & \sum_{k \in \mathcal{K}_j} x_{jk}^t \leq 1, \quad \forall j, t, \\
 & \sum_{k' \in \mathcal{K}_j: k' \geq k} x_{jk'}^{t-1} \leq \sum_{k' \in \mathcal{K}_j: k' \geq k} x_{jk'}^t, \quad \forall j, t \in \mathcal{T} \setminus \{1\}, k, \\
 & \sum_{t' \in \mathcal{T}: t' \leq t} \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} c_{jk}^{t'} (x_{jk}^{t'} - x_{jk}^{t'-1}) \leq \sum_{t' \in \mathcal{T}: t' \leq t} b^{t'}, \quad \forall t, \\
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 & z_i^{ts} \in \arg \max_{z_i^{ts} \in \{0, 1\}} u_{i0}^{ts} (1 - z_i^{ts}) + U_i^{ts} z_i^{ts} \quad \forall i, t, s, \\
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First approach: mixed-integer linear model

Single-level MIP formulation

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 \max \quad & \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{I}} n_i^t \frac{1}{|\mathcal{S}|} \sum_{s \in \mathcal{S}} z_i^{ts} \\
 \text{s.t.} \quad & \sum_{k \in \mathcal{K}_j} x_{jk}^t \leq 1, \quad \forall j, t, \\
 & \sum_{k' \in \mathcal{K}_j: k' \geq k} x_{jk'}^{t-1} \leq \sum_{k' \in \mathcal{K}_j: k' \geq k} x_{jk'}^t, \quad \forall j, t \in \mathcal{T} \setminus \{1\}, k, \\
 & \sum_{t' \in \mathcal{T}: t' \leq t} \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} c_{jk}^{t'} (x_{jk}^{t'} - x_{jk}^{t'-1}) \leq \sum_{t' \in \mathcal{T}: t' \leq t} b^{t'}, \quad \forall t, \\
 & u_{ij}^{ts} = \sum_{k \in \mathcal{K}_j} a_{ijk}^{ts} x_{jk}^t, \quad \forall i, j, t, s, \\
 & x_{jk}^t \in \{0, 1\}, \quad \forall j, t, k, \\
 & z_i^{ts} \in \arg \max_{z_i^{ts} \in \{0, 1\}} u_{i0}^{ts} (1 - z_i^{ts}) + U_i^{ts} z_i^{ts} \quad \forall i, t, s, \\
 & U_i^{ts} = \max_{\sigma_i^{ts}} \sum_{j \in \mathcal{J}} \sum_{r \in \mathcal{J}} \lambda_{ir} u_{ij}^{ts} \sigma_{ijr}^{ts} \quad \forall i, t, s, \\
 & \text{s.t.} \quad \sum_{j \in \mathcal{J}} \sigma_{ijr}^{ts} \leq 1, \quad \forall i, t, s, r, \\
 & \quad \sum_{r \in \mathcal{J}} \sigma_{ijr}^{ts} \leq 1, \quad \forall i, t, s, j, \\
 & \quad \sigma_{ijr}^{ts} \in [0, 1], \quad \forall i, t, s, j, r.
 \end{aligned}$$

Single-level MIP formulation

$$\begin{aligned}
 \max \quad & \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{I}} n_i^t \frac{1}{|\mathcal{S}|} \sum_{s \in \mathcal{S}} z_i^{ts} \\
 \text{s.t.} \quad & \sum_{k \in \mathcal{K}_j} x_{jk}^t \leq 1, \quad \forall j, t, \\
 & \sum_{k' \in \mathcal{K}_j: k' \geq k} x_{jk'}^{t-1} \leq \sum_{k' \in \mathcal{K}_j: k' \geq k} x_{jk'}^t, \quad \forall j, t \in \mathcal{T} \setminus \{1\}, k, \\
 & \sum_{t' \in \mathcal{T}: t' \leq t} \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} c_{jk}^{t'} (x_{jk}^{t'} - x_{jk}^{t'-1}) \leq \sum_{t' \in \mathcal{T}: t' \leq t} b^{t'}, \quad \forall t, \\
 & u_{ij}^{ts} = \sum_{k \in \mathcal{K}_j} a_{ijk}^{ts} x_{jk}^t, \quad \forall i, j, t, s, \\
 & x_{jk}^t \in \{0, 1\}, \quad \forall j, t, k, \\
 & z_i^{ts} \in \arg \max_{z_i^{ts} \in [0, 1]} u_{i0}^{ts} (1 - z_i^{ts}) + U_i^{ts} z_i^{ts} \quad \forall i, t, s, \\
 & U_i^{ts} = \sum_{j \in \mathcal{J}} \sum_{r \in \mathcal{J}} \lambda_{ir} u_{ij}^{ts} \sigma_{ijr}^{ts} \quad \forall i, t, s, \\
 & \sum_{j \in \mathcal{J}} \sigma_{ijr}^{ts} \leq 1, \quad \forall i, t, s, r, \\
 & \sum_{r \in \mathcal{J}} \sigma_{ijr}^{ts} \leq 1, \quad \forall i, t, s, j, \\
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 \end{aligned}$$

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 \max \quad & \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{I}} n_i^t \frac{1}{|\mathcal{S}|} \sum_{s \in \mathcal{S}} z_i^{ts} \\
 \text{s.t.} \quad & \sum_{k \in \mathcal{K}_j} x_{jk}^t \leq 1, \quad \forall j, t, \\
 & \sum_{k' \in \mathcal{K}_j: k' \geq k} x_{jk'}^{t-1} \leq \sum_{k' \in \mathcal{K}_j: k' \geq k} x_{jk'}^t, \quad \forall j, t \in \mathcal{T} \setminus \{1\}, k, \\
 & \sum_{t' \in \mathcal{T}: t' \leq t} \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} c_{jk}^{t'} (x_{jk}^{t'} - x_{jk}^{t'-1}) \leq \sum_{t' \in \mathcal{T}: t' \leq t} b^{t'}, \quad \forall t, \\
 & u_{ij}^{ts} = \sum_{k \in \mathcal{K}_j} a_{ijk}^{ts} x_{jk}^t, \quad \forall i, j, t, s, \\
 & x_{jk}^t \in \{0, 1\}, \quad \forall j, t, k, \\
 & u_{i0}^{ts} z_i^{ts} \leq U_i^{ts} z_i^{ts}, \quad \forall i, t, s, \\
 & U_i^{ts} = \sum_{j \in \mathcal{J}} \sum_{r \in \mathcal{J}} \lambda_{ir} u_{ij}^{ts} \sigma_{ijr}^{ts} \quad \forall i, t, s, \\
 & \sum_{j \in \mathcal{J}} \sigma_{ijr}^{ts} \leq 1, \quad \forall i, t, s, r, \\
 & \sum_{r \in \mathcal{J}} \sigma_{ijr}^{ts} \leq 1, \quad \forall i, t, s, j, \\
 & \sigma_{ijr}^{ts}, z_i^{ts} \in [0, 1], \quad \forall i, t, s, j, r.
 \end{aligned}$$

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\max \quad & \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{I}} n_i^t \frac{1}{|\mathcal{S}|} \sum_{s \in \mathcal{S}} z_i^{ts} \\
\text{s.t.} \quad & \sum_{k \in \mathcal{K}_j} x_{jk}^t \leq 1, \quad \forall j, t, \\
& \sum_{k' \in \mathcal{K}_j: k' \geq k} x_{jk'}^{t-1} \leq \sum_{k' \in \mathcal{K}_j: k' \geq k} x_{jk'}^t, \quad \forall j, t \in \mathcal{T} \setminus \{1\}, k, \\
& \sum_{t' \in \mathcal{T}: t' \leq t} \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} c_{jk}^{t'} (x_{jk}^{t'} - x_{jk}^{t'-1}) \leq \sum_{t' \in \mathcal{T}: t' \leq t} b^{t'}, \quad \forall t, \\
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& x_{jk}^t \in \{0, 1\}, \quad \forall j, t, k, \\
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& \sum_{j \in \mathcal{J}} \sigma_{ijr}^{ts} \leq 1, \quad \forall i, t, s, r, \\
& \sum_{r \in \mathcal{J}} \sigma_{ijr}^{ts} \leq 1, \quad \forall i, t, s, j, \\
& \sigma_{ijr}^{ts}, z_i^{ts} \in [0, 1], \quad \forall i, t, s, j, r.
\end{aligned}$$

$$u_{i0}^{ts} z_i^{ts} \leq \sum_{j \in \mathcal{J}} \sum_{r \in \mathcal{J}} \lambda_{ir} u_{ij}^{ts} \sigma_{ijr}^{ts} z_i^{ts}, \quad \forall i, t, s,$$

$$\sum_{j \in \mathcal{J}} \sigma_{ijr}^{ts} \leq 1, \quad \forall i, t, s, r,$$

$$\sum_{r \in \mathcal{J}} \sigma_{ijr}^{ts} \leq 1, \quad \forall i, t, s, j,$$

$$\sigma_{ijr}^{ts}, z_i^{ts} \in [0, 1], \quad \forall i, t, s, j, r.$$

Linearization

1. Apply a perspective transformation. z remains continuous

$$u_{i0}^{ts} z_i^{ts} \leq \sum_{j \in \mathcal{J}} \sum_{r \in \mathcal{J}} \lambda_{ir} u_{ij}^{ts} \sigma_{ijr}^{ts}, \quad \forall i, t, s,$$

$$\sum_{j \in \mathcal{J}} \sigma_{ijr}^{ts} \leq z_i^{ts}, \quad \forall i, t, s, r,$$

$$\sum_{r \in \mathcal{J}} \sigma_{ijr}^{ts} \leq z_i^{ts}, \quad \forall i, t, s, j,$$

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$$\sigma_{ijr}^{ts}, z_i^{ts} \in [0, 1], \quad \forall i, t, s, j, r.$$

Linearization

1. Apply a perspective transformation. z remains continuous

$$u_{i0}^{ts} z_i^{ts} \leq \sum_{j \in \mathcal{J}} \sum_{r \in \mathcal{J}} \lambda_{ir} w_{ijr}^{ts}, \quad \forall i, t, s,$$

$$\sum_{j \in \mathcal{J}} \sigma_{ijr}^{ts} \leq z_i^{ts}, \quad \forall i, t, s, r,$$

$$\sum_{r \in \mathcal{J}} \sigma_{ijr}^{ts} \leq z_i^{ts}, \quad \forall i, t, s, j,$$

$$z_i^{ts} \in [0, 1], \quad \forall i, t, s, j, r,$$

$$\sigma_{ijr}^{ts} \in \{0, 1\}, \quad \forall i, t, s, j, r.$$

2. Standard linearization of bilinear terms with binary variables. σ needs to be integer.

Define new **auxiliary variables** $w_{ijr}^{ts} := u_{ij}^{ts} \sigma_{ijr}^{ts}$, $\forall j, r \in \mathcal{J}$, and use constraints:

$$w_{ijr}^{ts} \leq M_{ij}^{ts} \sigma_{ijr}^{ts}, \quad \forall i \in \mathcal{I}, t \in \mathcal{T}, s \in \mathcal{S}, j, r \in \mathcal{J},$$

$$w_{ijr}^{ts} \leq u_{ij}^{ts}, \quad \forall i \in \mathcal{I}, t \in \mathcal{T}, s \in \mathcal{S}, j, r \in \mathcal{J},$$

Valid inequalities and preprocessing

$$w_{ijr}^{ts} := u_{ij}^{ts} \sigma_{ijr}^{ts}$$

$$w_{ijr}^{ts} := \begin{cases} u_{ij}^{ts} = \sum_{k \in \mathcal{K}_j} a_{ijk}^{ts} x_{jk}^t, & u_{ij}^{ts} \text{ is the } r\text{-th largest utility for customer } i, \\ 0, & 0 = a_{ij0}^{ts} \leq a_{ij1}^{ts} \leq \dots \leq a_{ij|\mathcal{K}_j|}^{ts} \\ & \text{otherwise.} \end{cases}$$

Valid inequalities and preprocessing

$$w_{ijr}^{ts} := u_{ij}^{ts} \sigma_{ijr}^{ts}$$

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$$w_{ijr}^{ts} \leq a_{ijk}^{ts} \sigma_{ijr}^{ts} + \sum_{\substack{k' \in \mathcal{K}_j: \\ k' > k}} (a_{ijk'}^{ts} - a_{ijk}^{ts}) x_{jk'}^t, \quad \forall j, r, k,$$

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$$\sum_{r \in \mathcal{J}} w_{ijr}^{ts} \leq \sum_{r \in \mathcal{J}} a_{ijk}^{ts} \sigma_{ijr}^{ts} + \sum_{\substack{k' \in \mathcal{K}_j: \\ k' > k}} (a_{ijk'}^{ts} - a_{ijk}^{ts}) x_{jk'}^t, \quad \forall j, k,$$

Valid inequalities and preprocessing

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$$\sum_{r \in \mathcal{J}} w_{ijr}^{ts} \leq \sum_{r \in \mathcal{J}} a_{ijk}^{ts} \sigma_{ijr}^{ts} + \sum_{\substack{k' \in \mathcal{K}_j: \\ k' > k}} (a_{ijk'}^{ts} - a_{ijk}^{ts}) x_{jk'}^t, \quad \forall j, k,$$

$$\sum_{r \in \mathcal{J}} \sigma_{ijr}^{ts} \leq \sum_{k \in \mathcal{K}_j} x_{jk}^t, \quad \forall j,$$

$$\sigma_{ijr}^{ts} = 0 \quad \forall i, t, s, j, r : \lambda_{ir} = 0.$$

Second approach: Benders' decomposition

Master Problem

$$\begin{aligned}
 \max_{x,z} \quad & \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{I}} n_i^t \frac{1}{|\mathcal{S}|} \sum_{s \in \mathcal{S}} z_i^{ts} \\
 \text{s.t.} \quad & \sum_{k \in \mathcal{K}_j} x_{jk}^t \leq 1, \quad \forall j, t, \\
 & \sum_{\substack{k' \in \mathcal{K}_j: \\ k' \geq k}} x_{jk'}^{t-1} \leq \sum_{\substack{k' \in \mathcal{K}_j: \\ k' \geq k}} x_{jk'}^t, \quad \forall j, t \in \mathcal{T} \setminus \{1\}, k, \\
 & \sum_{\substack{t' \in \mathcal{T}: \\ t' \leq t}} \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} c_{jk}^{t'} (x_{jk}^{t'} - x_{jk}^{t'-1}) \leq \sum_{\substack{t' \in \mathcal{T}: \\ t' \leq t}} b^{t'}, \quad \forall t, \\
 & x_{jk}^t \in \{0, 1\}, \quad \forall j, t, k, \\
 & u_{ij}^{ts} = \sum_{k \in \mathcal{K}_j} a_{ijk}^{ts} x_{jk}^t, \quad \forall i, j, t, s, \\
 & z_i^{ts} \in \arg \max_{z_i^{ts}} u_{i0}^{ts} (1 - z_i^{ts}) + U_i^{ts} z_i^{ts} \quad \forall i, t, s, \\
 & \text{s.t.} \quad z_i^{ts} \in \{0, 1\}, \\
 & U_i^{ts} = \Phi_{\lambda_i}(u_{i1}^{ts}, \dots, u_{i|\mathcal{J}|}^{ts}), \quad \forall i, t, s.
 \end{aligned}$$

Master Problem

$$\begin{aligned}
 \max_{x, z} \quad & \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{I}} n_i^t \frac{1}{|\mathcal{S}|} \sum_{s \in \mathcal{S}} z_i^{ts} \\
 \text{s.t.} \quad & \sum_{k \in \mathcal{K}_j} x_{jk}^t \leq 1, \quad \forall j, t, \\
 & \sum_{\substack{k' \in \mathcal{K}_j: \\ k' \geq k}} x_{jk'}^{t-1} \leq \sum_{\substack{k' \in \mathcal{K}_j: \\ k' \geq k}} x_{jk'}^t, \quad \forall j, t \in \mathcal{T} \setminus \{1\}, k, \\
 & \sum_{\substack{t' \in \mathcal{T}: \\ t' \leq t}} \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} c_{jk}^{t'} (x_{jk}^{t'} - x_{jk}^{t'-1}) \leq \sum_{\substack{t' \in \mathcal{T}: \\ t' \leq t}} b^{t'}, \quad \forall t, \\
 & x_{jk}^t \in \{0, 1\}, \quad \forall j, t, k, \\
 & u_{ij}^{ts} = \sum_{k \in \mathcal{K}_j} a_{ijk}^{ts} x_{jk}^t, \quad \forall i, j, t, s, \\
 & z_i^{ts} \in \arg \max_{z_i^{ts}} u_{i0}^{ts} (1 - z_i^{ts}) + U_i^{ts} z_i^{ts} \quad \forall i, t, s, \\
 & \text{s.t.} \quad z_i^{ts} \in \{0, 1\}, \\
 & U_i^{ts} = \Phi_{\lambda_i}(u_{i1}^{ts}, \dots, u_{i|\mathcal{J}|}^{ts}), \quad \forall i, t, s.
 \end{aligned}$$

Master Problem

$$\begin{aligned}
 \max_{x,z} \quad & \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{I}} n_i^t \frac{1}{|\mathcal{S}|} \sum_{s \in \mathcal{S}} z_i^{ts} \\
 \text{s.t.} \quad & \sum_{k \in \mathcal{K}_j} x_{jk}^t \leq 1, \quad \forall j, t, \\
 & \sum_{\substack{k' \in \mathcal{K}_j: \\ k' \geq k}} x_{jk'}^{t-1} \leq \sum_{\substack{k' \in \mathcal{K}_j: \\ k' \geq k}} x_{jk'}^t, \quad \forall j, t \in \mathcal{T} \setminus \{1\}, k, \\
 & \sum_{\substack{t' \in \mathcal{T}: \\ t' \leq t}} \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} c_{jk}^{t'} (x_{jk}^{t'} - x_{jk}^{t'-1}) \leq \sum_{\substack{t' \in \mathcal{T}: \\ t' \leq t}} b^{t'}, \quad \forall t, \\
 & x_{jk}^t \in \{0, 1\}, \quad \forall j, t, k, \\
 & u_{0i}^{ts} z_i^{ts} \leq U_i^{ts}(x) z_i^{ts}, \quad \forall i \in \mathcal{I}, t \in \mathcal{T}, s \in \mathcal{S}.
 \end{aligned}$$

Master Problem

$$\begin{aligned}
 \max_{\mathbf{x}, \mathbf{z}} \quad & \sum_{t \in \mathcal{T}} \sum_{i \in \mathcal{I}} n_i^t \frac{1}{|\mathcal{S}|} \sum_{s \in \mathcal{S}} z_i^{ts} \\
 \text{s.t.} \quad & \sum_{k \in \mathcal{K}_j} x_{jk}^t \leq 1, \quad \forall j, t, \\
 & \sum_{\substack{k' \in \mathcal{K}_j: \\ k' \geq k}} x_{jk'}^{t-1} \leq \sum_{\substack{k' \in \mathcal{K}_j: \\ k' \geq k}} x_{jk'}^t, \quad \forall j, t \in \mathcal{T} \setminus \{1\}, k, \\
 & \sum_{\substack{t' \in \mathcal{T}: \\ t' \leq t}} \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} c_{jk}^{t'} (x_{jk}^{t'} - x_{jk}^{t'-1}) \leq \sum_{\substack{t' \in \mathcal{T}: \\ t' \leq t}} b^{t'}, \quad \forall t, \\
 & x_{jk}^t \in \{0, 1\}, \quad \forall j, t, k, \\
 & u_{0i}^{ts} z_i^{ts} \leq U_i^{ts}(\mathbf{x}) z_i^{ts} \leq U_i^{ts}(\bar{\mathbf{x}}) z_i^{ts} + \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \left[\bar{s}_{jk} (x_{jk}^t - \bar{x}_{jk}^t) \right]^+, \quad \forall i \in \mathcal{I}, t \in \mathcal{T}, s \in \mathcal{S}.
 \end{aligned}$$

Subproblems

$$u_{0i}^{ts} z_i^{ts} \leq U_i^{ts}(\bar{x}) z_i^{ts} + \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \left[\bar{s}_{jk} (x_{jk}^t - \bar{x}_{jk}^t) \right]^+$$

$$\begin{aligned} \text{(PR)} \quad & \max_{\sigma} \quad \sum_{j \in \mathcal{J}} \sum_{r \in \mathcal{R}} \sum_{k \in \mathcal{K}_j} \lambda_r a_{jk} \sigma_{jkr} \\ \text{s.t.} \quad & \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \sigma_{jkr} \leq 1, \quad \forall r, \\ & \sum_{r \in \mathcal{R}} \sum_{k \in \mathcal{K}_j} \sigma_{jkr} \leq 1, \quad \forall j, \\ & \sum_{r \in \mathcal{R}} \sigma_{jkr} \leq \bar{x}_{jk}, \quad \forall j, k, \\ & \sigma_{jkr} \geq 0, \quad \forall j, r, k. \end{aligned}$$

Subproblems

$$u_{0i}^{ts} z_i^{ts} \leq U_i^{ts}(\bar{x}) z_i^{ts} + \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \left[\bar{s}_{jk} (x_{jk}^t - \bar{x}_{jk}^t) \right]^+$$

$$(\text{PR}) \quad \max_{\sigma} \quad \sum_{j \in \mathcal{J}} \sum_{r \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \lambda_r a_{jk} \sigma_{jkr}$$

$$\text{s.t.} \quad \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \sigma_{jkr} \leq 1, \quad \forall r, \quad (\gamma_r)$$

$$\sum_{r \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \sigma_{jkr} \leq 1, \quad \forall j, \quad (\delta_j)$$

$$\sum_{r \in \mathcal{J}} \sigma_{jkr} \leq \bar{x}_{jk}, \quad \forall j, k \quad (\eta_{jk})$$

$$\sigma_{jkr} \geq 0, \quad \forall j, r, k.$$

$$(\text{DU}) \quad \min \quad \sum_{r \in \mathcal{J}} \gamma_r + \sum_{j \in \mathcal{J}} \delta_j + \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \bar{x}_{jk} \eta_{jk}$$

$$\text{s.t.} \quad \gamma_r + \delta_j + \eta_{jk} \geq \lambda_r a_{jk}, \quad \forall j, r \in \mathcal{J}, k \in \mathcal{K}_j.$$

Subproblems

$$u_{0i}^{ts} z_i^{ts} \leq U_i^{ts}(\bar{x}) z_i^{ts} + \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \left[\bar{s}_{jk} (x_{jk}^t - \bar{x}_{jk}^t) \right]^+$$

$$\begin{aligned}
 (\text{PR}) \quad & \max_{\sigma} \sum_{j \in \mathcal{J}} \sum_{r \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \lambda_r a_{jk} \sigma_{jkr} \\
 \text{s.t.} \quad & \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \sigma_{jkr} \leq 1, \quad \forall r, \\
 & \sum_{r \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \sigma_{jkr} \leq 1, \quad \forall j, \\
 & \sum_{r \in \mathcal{J}} \sigma_{jkr} \leq \bar{x}_{jk}, \quad \forall j, k \\
 & \sigma_{jkr} \geq 0, \quad \forall j, r, k.
 \end{aligned}$$

$$\begin{aligned}
 (\text{DU}) \quad & \min \sum_{r \in \mathcal{J}} \gamma_r + \sum_{j \in \mathcal{J}} \delta_j + \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \bar{x}_{jk} \eta_{jk} \\
 \text{s.t.} \quad & \gamma_r + \delta_j + \eta_{jk} \geq \lambda_r a_{jk}, \quad \forall j, r \in \mathcal{J}, k \in \mathcal{K}_j.
 \end{aligned}$$

Subproblems

$$u_{0i}^{ts} z_i^{ts} \leq U_i^{ts}(\bar{x}) z_i^{ts} + \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \left[\bar{s}_{jk} (x_{jk}^t - \bar{x}_{jk}^t) \right]^+$$

$$\begin{aligned}
 (\text{PR}) \quad & \max_{\sigma} \quad \sum_{j \in \mathcal{J}} \sum_{r \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \lambda_r a_{jk} \sigma_{jkr} \\
 \text{s.t.} \quad & \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \sigma_{jkr} \leq 1, \quad \forall r, \\
 & \sum_{r \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \sigma_{jkr} \leq 1, \quad \forall j, \\
 & \sum_{r \in \mathcal{J}} \sigma_{jkr} \leq \bar{x}_{jk}, \quad \forall j, k \\
 & \sigma_{jkr} \geq 0, \quad \forall j, r, k.
 \end{aligned}$$

$$\begin{aligned}
 (\text{DU}) \quad & \min \quad \sum_{r \in \mathcal{J}} \gamma_r + \sum_{j \in \mathcal{J}} \delta_j + \sum_{j \in \mathcal{J}} \sum_{k \in \mathcal{K}_j} \bar{x}_{jk} \eta_{jk} \\
 \text{s.t.} \quad & \gamma_r + \delta_j + \eta_{jk} \geq \lambda_r a_{jk}, \quad \forall j, r \in \mathcal{J}, k \in \mathcal{K}_j.
 \end{aligned}$$

Tailored Primal-Dual algorithm to solve (PR)-(DU) at each iteration.

Computational Experiments

1. Computational Performance using synthetic data

Instance sizes and other parameters

Facilities and types: $|\mathcal{L}| \in \{10, 30\}$, $|\mathcal{K}_j| = 4$.

Customers: $|\mathcal{I}| \in \{20, 30, 40, 50\}$.

Scenarios: $|\mathcal{S}| \in \{5, 10\}$.

Budget and time period: $|\mathcal{T}| = 3$, $b^t \in \{5, 10\}$.

$u_0 \in \{10, 12, 15\}$.

Five instances of each size, a total of 1680.

Time limit equal to 3600 seconds.

The total number of instances per row is 30.

Type C: $\lambda = (1, 0, \dots, 0)$. Standard RUM.

Type G: $\lambda = (1, \frac{1}{9}, \frac{1}{27}, 0, \dots, 0)$. Geometric rule.

Type K: $\lambda = (1, 1, 0, \dots, 0)$. Aggregation with *consideration set* of size 2.

Type L: $\lambda = (1, \frac{1}{2}, 0, \dots, 0)$. Linear rule.

Comparison of formulations (10 scenarios)

SL: MILP

SL+VI: MILP with valid ineq.

B: Benders Decomposition

$\mathcal{I}$	λ	$\mathcal{J}$	Time[s] (#Solved)			MIPGap[%]		
			SL	SL+VI	B	SL	SL+VI	B
30	C	10	0.9 (30)	11.7 (30)	6.3 (30)	0.0	0.0	0.0
		30	6.7 (30)	112.0 (25)	49.2 (25)	0.0	27.2	2.0
	G	10	1890.4 (1)	1210.5 (13)	671.5 (20)	56.3	27.3	14.8
		30	TL (0)	TL (0)	TL (0)	126.7	219.0	47.0
	K	10	2539.9 (9)	355.5 (30)	174.6 (28)	14.9	0.0	14.5
		30	TL (0)	3365.0 (2)	475.2 (14)	38.9	14.4	10.3
	L	10	1548.3 (11)	791.1 (18)	665.3 (28)	23.4	21.5	6.7
		30	TL (0)	TL (0)	1151.3 (5)	86.1	50.7	31.8
50	C	10	1.6 (30)	30.1 (30)	14.7 (30)	0.0	0.0	0.0
		30	13.6 (30)	215.2 (20)	73.0 (18)	0.0	44.3	8.6
	G	10	TL (0)	1597.1 (5)	338.6 (10)	56.6	33.2	20.3
		30	TL (0)	TL (0)	TL (0)	166.8	404.1	67.1
	K	10	3128.0 (1)	656.3 (22)	299.3 (21)	18.5	12.7	6.8
		30	TL (0)	TL (0)	1184.4 (12)	59.7	23.8	17.9
	L	10	2723.4 (3)	1421.0 (13)	1191.6 (17)	31.3	22.6	7.7
		30	TL (0)	TL (0)	2659.9 (4)	101.5	61.8	33.6

Comparison of formulations (10 scenarios)

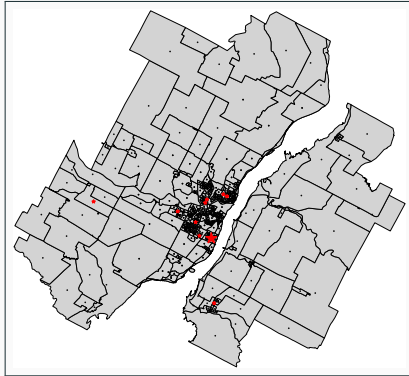
SL: MILP

SL+VI: MILP with valid ineq.

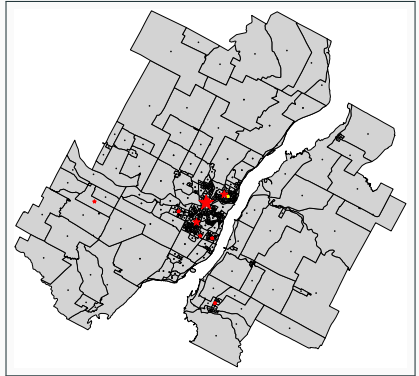
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Case Study: Electric Vehicle Charging Stations in Trois-Rivières, Canada

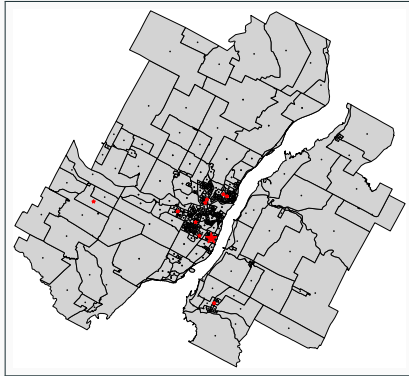


(a) $\lambda = C = (1, 0, \dots, 0)$

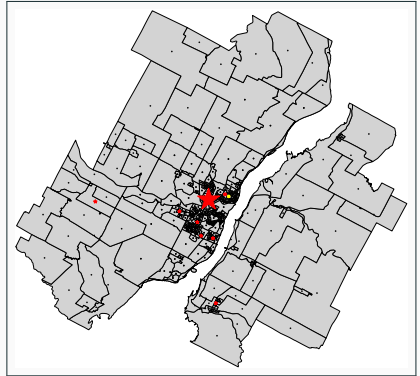


(b) $\lambda = K = (1, 1, 0, \dots, 0)$

Case Study: Electric Vehicle Charging Stations in Trois-Rivières, Canada



(a) $\lambda = C = (1, 0, \dots, 0)$



(c) $\lambda = L = (1, \frac{1}{2}, 0, \dots, 0)$

Conclusions

The CMCFL introduces a **cooperative framework** for the MCFL with binary choice rule.

Conclusions and future work

Conclusions

The CMCFL introduces a **cooperative framework** for the MCFL with binary choice rule.

Future work

Add **capacities** to the facilities dependent on the type of plant installed $\Rightarrow$ Bilevel location-allocation problem.

Robustify the OMf associated to the utility. For instance considering **non-monotone or negative λ -weights**, or **variable λ -weights** that meet certain conditions associated to the knowledge of the customer, and optimize U in the worst-case.

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Thank you for your attention!

Questions?



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arXiv preprint arXiv:2305.15169, 2023.



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