

On the nested p-center problem



Christof Brandstetter, Markus Sinnl

Institute of Business Analytics and Technology Transformation / JKU Business School, Johannes Kepler University Linz

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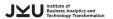
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JOHANNES KEPLER UNIVERSITY LINZ Altenberger Straße 69 4040 Linz, Austria





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 - o a set V of locations,



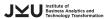
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 - the maximum distance of any location to its closest opened location is minimized.

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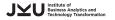
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- are called nested iff $V^h \subseteq V^{h+1}$ for h = 1, ..., H-1

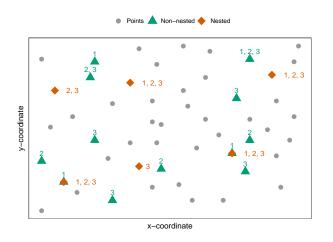
Definition

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Introduced by Roodman and Schwarz 1975 and used in e.g., Albareda-Sambola et al. 2009; Bakker and Nickel 2024; Conforti et al. 2014 and reintroduced as nesting by McGarvey and Thorsen 2022



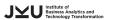
p-center problem vs nested p-center problem for $P = \{4, 5, 6\}$



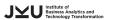


Definition part I

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Definition part II

 a feasible solution to the nested p-center problem consists of a set J^h ⊆ V



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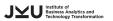
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- the goal is to find a feasible solution which minimizes $\sum_{h=1}^{H} d(\mathcal{J}^h)$,
 - $\circ \text{ where } d(\mathcal{J}^h) = \max_{i \in \mathcal{I}} \min_{j \in \mathcal{J}^h} d_{ij} \text{ for } h \in \mathcal{H}$

Mixed Integer Linear Programming (MILP) Formulations

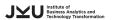




The nested p-center problem: MILP formulation

Decision variables

$$y_j^h \dots \begin{cases} 1 \dots \text{ if location } j \text{ is open in} \\ & \text{time period } h \\ 0 \dots \text{ otherwise} \end{cases}$$



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(nPC2) based on Gaar and Sinnl 2022

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 (1a)

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$$\sum_{j \in V} y_j^h = \rho^h$$

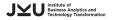
$$\forall h \in \mathcal{H}$$
 (1b)

$$z^{h} \geq d_{ij} - \sum_{j':d_{ij'} < d_{ij}} (d_{ij} - d_{ij'}) y_{j'}^{h} \qquad \forall i, j \in V, h \in \mathcal{H} \quad (1c)$$

$$y_{j}^{h} \geq y_{j}^{h-1} \qquad \forall j \in V, \forall h \in \mathcal{H} \setminus \{1\} \quad (1d)$$

$$y_j^h \ge y_j^{h-1}$$
 $\forall j \in V, \forall h \in \mathcal{H} \setminus \{1\}$ (1d)

$$(y,z) \in |V| |\mathcal{H}| \times \mathbb{R}_{\geq 0}$$
 (1e)



The nested p-center problem: Compact MILP formulation

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(nPC3) based on Ales and Elloumi 2018

$$\min \sum_{h \in \mathcal{H}} z^{h}$$
s.t.
$$\sum_{j \in V} y_{j}^{h} = p^{h} \qquad \forall h \in \mathcal{H}$$

$$D_{0} + \sum_{k=1}^{K} (D_{k} - D_{k-1}) u_{k}^{h} \leq z^{h} \ \forall h \in \mathcal{H}$$

$$u_{k}^{h} + \sum_{j: d_{ij} < D_{k}} y_{j}^{h} \geq 1 \ \forall i \in V, \forall h \in \mathcal{H}, \forall k \in S_{i} \cup \{K\}$$

$$u_{k}^{h} \geq u_{k+1}^{h} \qquad \forall h \in \mathcal{H}, \forall k \in \mathcal{K} \setminus \{K\}$$

$$y_{j}^{h} \geq y_{j}^{h-1} \qquad \forall h \in \mathcal{H} \setminus \{1\}$$

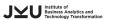
$$(u, y, z) \in |V| \ |K| \times |V| \ |\mathcal{H}| \times \mathbb{R}_{>0}$$



Comparing formulations

Table: Comparison of complexity of the formulations

	(nPC1)	(nPC2)	(nPC3)
Variables	$\mathcal{O}(\mathcal{I} \mathcal{J} \mathcal{H})$	$\mathcal{O}(\mathcal{J} \mathcal{H})$	$\mathcal{O}((\mathcal{I} + \mathcal{K}) \mathcal{H})$
Constraints	$\mathcal{O}(\mathcal{I} \mathcal{J} \mathcal{H})$	$\mathcal{O}(\mathcal{I} \mathcal{J} \mathcal{H})$	$\mathcal{O}(\text{min}(\mathcal{I} \mathcal{J} , \mathcal{I} \mathcal{K}) \mathcal{H})$



Comparing formulations

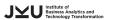
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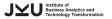
The pCP version of formulation (nPC3) has the best known linear programming (LP)-bounds, while the pCP versions of (nPC1) and (nPC2) have equally but worse LP-bounds.

Improving the formulations





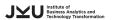
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Lemma 1

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can be replaced by

$$z^h \geq \max\{\underline{z^h}, d_{ij}\} - \sum_{j': d_{ij'} < d_{ij}} \left(\max\{\underline{z^h}, d_{ij}\} - \max\{\underline{z^h}, d_{ij'}\} \right) y_{j'}^h$$
 (nL-OPT)

The lemma is based on Lemma 5 in Gaar and Sinnl (2022).



Observation 1

Let $\underline{z^h}$ be a valid lower bound and $\overline{z^h}$ be a valid upper bound on the decision variable z^h for $h \in \mathcal{H}$, then the distinct distance D_k can only be the optimal distance for z^h if $\underline{z^h} \leq D^k \leq \overline{z^h}$ holds.

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Following Observation 1, we observe that decision variables $u_h^k = 0$ for $k: D_k > \overline{z^h}$ and $u_h^k = 1$ for $k: D_k < \underline{z^h}$ in any optimal solution.

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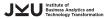
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Proposition 1

The decision variables u_k^h which are zero or one following Observation 2, are redundant.



Obtaining bounds



Obtaining bounds

Observation 3

For $\mathcal{H} = \{1\}$ the n-pCP reduces to the pCP where $p = p^1$, so the optimal objective value (z'^{h*}) of the pCP, where $p = p^h$ is a lower bound \underline{z}^h on the decision variable z^h of the n-pCP.



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Proposition 2

Given an valid upper bound UB on the objective value of the n-pCP and valid lower bounds \underline{z}^h on the variable z^h can be obtained by the following equation:

$$\overline{z^h} = \frac{UB - \sum_{x=h+1}^{H} \underline{z^h}}{h} \tag{3}$$





• C++, CPLEX 20.1

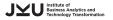
- C++, CPLEX 20.1
- preprocessing
 - Solving p-center problem with $p = p^H$
 - Use optimal solution value as lower bound for p-center problem with $p = p^{H-1}$
 - Repeat for remaining p^h and calculate the upper bounds $\overline{z^h}$

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 - branch-and-cut algorithm
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 - o branch-and-bound algorithm



noPP: without preprocessing and without lifting

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• PP: with preprocessing



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- PH: PP with starting heuristic



Results



Instance from literature

- instance set PMED
 - 40 instances
 - $\circ \mathcal{P} = \{p, p+1, p+2\}, p \text{ from 5 to 200, } |V| \text{ from 100 and 900 nodes}$

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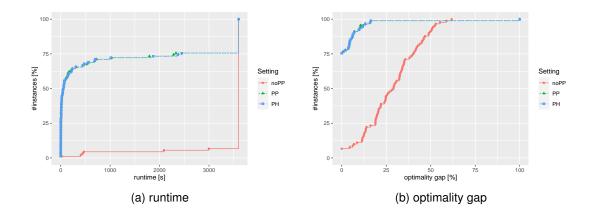
$$P = \{p, p + 1, p + 2\}, p \text{ from 5 to 200, } |V| \text{ from 100 and 900 nodes}$$

- instance set TSPLIB
 - 50 instances
 - $\circ \mathcal{P} = \{4, 5, 6\}, |V| \text{ from 51 and 1002 nodes}$

Instance from literature

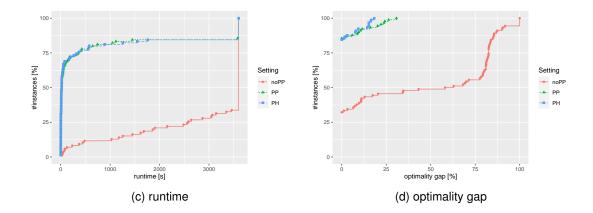
- instance set PMED
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 - $P = \{p, p + 1, p + 2\}, p \text{ from 5 to 200, } |V| \text{ from 100 and 900 nodes}$
- instance set TSPLIB
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 - $\circ \mathcal{P} = \{4, 5, 6\}, |V| \text{ from 51 and 1002 nodes}$
- computational setup
 - single core of Intel Xeon X5570 with 2.93 GHz and 8 GB RAM
 - timelimit of 3600 seconds.

Setting comparison on formulation (nPC2)



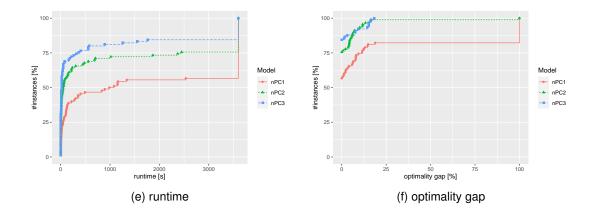


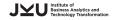
Setting comparison on formulation (nPC3)



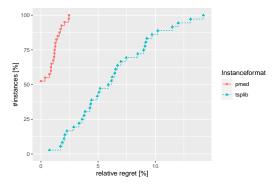


Formulation comparison on setting PH



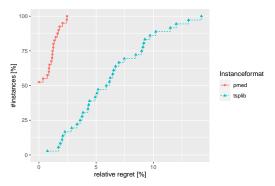


Managerial insights

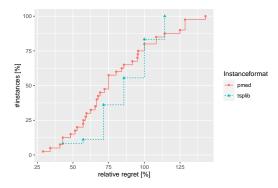


(g) Relative regret of the optimal solution value

Managerial insights



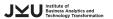
(i) Relative regret of the optimal solution value



(j) Relative regret of # of opened facilities

Conclusion

- (nPC3) best performance in runtimes and optimality gap
- The (nPC2) outperforms the (nPC1)
- Preprocessing brings a large speed up on all formulations
- Starting heuristic little effect, shows good upper bound obtained in preprocessing
- Maximal relative regret of the optimal objective function value of 15%
- Maximal relative regret of number of opened facilities above 150%





On the nested p-center problem



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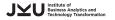
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References I

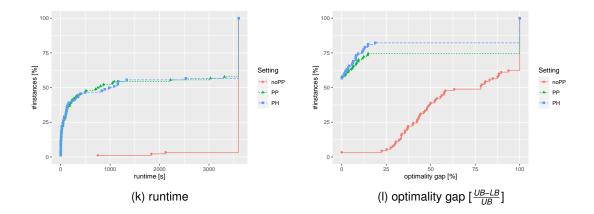
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Setting comparison on formulation (nPC1)





Decision variables

 $x_{ij}^h \dots \begin{cases} 1 \dots \text{ if location } i \text{ is assigned to} \\ \text{location } j \text{ in time period } h \\ 0 \dots \text{ otherwise} \end{cases}$



Decision variables

$$x_{ij}^h \dots \begin{cases} 1 \dots \text{ if location } i \text{ is assigned to} \\ & \text{location } j \text{ in time period } h \\ 0 \dots \text{ otherwise} \end{cases}$$

$$y_j^h \dots \begin{cases} 1 \dots \text{ if location } j \text{ is open in} \\ & \text{time period } h \\ 0 \dots \text{ otherwise} \end{cases}$$

Decision variables

$$X_{ij}^h \dots \begin{cases} 1 \dots \text{ if location } i \text{ is assigned to} \\ & \text{location } j \text{ in time period } h \\ 0 \dots \text{ otherwise} \end{cases}$$

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 z^h ... maximum distance between any customer i and its nearest open facility in period h

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z^h... maximum distance between any customer i and its nearest open facility in period h

(nPC1) based on Daskin 2013

$$\min \sum_{h \in \mathcal{H}} z^h \tag{4a}$$

s.t.
$$\sum_{j \in V} y_j^h = p^h \quad \forall h \in \mathcal{H}$$
 (4b)

$$\sum_{j\in V} x_{ij}^h = 1 \qquad \forall i\in V, h\in \mathcal{H}$$
 (4c)

$$x_{ij}^h \leq y_j^h \qquad \forall i, j \in V, h \in \mathcal{H}$$
 (4d)

$$\sum_{j \in V} d_{ij} X_{ij}^h \le z^h \qquad \forall i \in V, h \in \mathcal{H} \qquad (4e)$$

$$y_j^h \ge y_j^{h-1} \quad \forall h \in \mathcal{H} \setminus \{1\}$$
 (4f)

$$(x, y, z) \in |V|^2 |\mathcal{H}| \times |V| |\mathcal{H}| \times \mathbb{R}_{>0}$$
 (4g)

