

# On the nested $p$ -center problem



Master Thesis Seminar

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# Introduction

- **$p$ -center problem**: Open  $p$  facilities such that the maximum distance between any customer and its nearest open facility is minimized
- Related to the set cover problem and the assignment problem
- Nesting: Open additional facilities subsequently
- **Nested  $p$ -center problem**: Open  $p_h$  facilities in time period  $h$  such that the sum of the maximum distances between any customer and its nearest open facility in time period  $h$  is minimized.
- Use case: Ambulance/logistics stations, network design, screening/vaccination stations
- Optimizes expansion or retraction path

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# Definition of the nested $p$ -center problem

## Definition:

- given a set of customer demand points  $\mathcal{I}$ , potential facility locations  $\mathcal{J}$ , time periods  $\mathcal{H} = \{1, \dots, H\}$  and a set of integers  $\mathcal{P} = \{p^1, \dots, p^H\}$  where  $p^h \leq p^{h+1}$  for  $h = 1, \dots, H - 1$

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- open  $p^h$ -facilities at different locations  $j \in \mathcal{J}$
- the sum of maximum distance of any customer demand point to its closest opened facility in time period  $h$  is minimized
- and facilities open in period  $h$  has to be open in period  $h + 1$  as well for  $h = 1, \dots, H - 1$

This problem is **NP-hard** (kariv1979)

# Related work



## $p$ -center problem / TODO

- Hakimi (1964): first introduction of the  $p$ -center problem
- Daskin (2013): the standard textbook MILP formulation
- Contardo, Iori, and Kramer (2018): state-of-the-art model based on the set cover problem
- Ales and Elloumi (2018): compact MILP formulations for the  $p$ -center problem
- Gaar and Sinnl (2022): projection-based branch-and-cut algorithm for the  $p$ -center problem

# Nested facility location problems

- Roodman and Schwarz (1975): first introduction of the nesting constraint
- Roodman and Schwarz (1977): extension of the nesting to a phase-in and phase out
- McGarvey and Thorsen (2022): reintroduction of the nesting on the  $p$ -median problem
- Calogiuri et al. (2021): multi-period  $p$ -center problem with time-dependent travel times but no additional opening of facilities
  - uses multiple periods, but do not nest the facilities

**Methodology / Mehtod is how I do things, Mehtodology is more about which questions I can answer**



# Mixed Integer Linear Programming (MILP) formulation of the nested $p$ -center

$\mathcal{I}$  ... set of customer demand points with  $i \in \mathcal{I}$

$\mathcal{J}$  ... set of potential facility locations with  $j \in \mathcal{J}$

$\mathcal{H}$  ... set of time periods,  $\mathcal{H} = \{1, 2, \dots, H\}$

$p_h$  = number of facilities to be opened in time period  $h \in \mathcal{H}$  with  $p_h > p_{h-1}$

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$d_{ij}$  = distance between customer demand point  $i$  and potential facility location  $j$

$x_{ijh} = \begin{cases} 1 & \dots \text{ if customer } i \text{ is assigned to facility } j \text{ in time period } h \\ 0 & \dots \text{ otherwise} \end{cases}$

$y_{jh} = \begin{cases} 1 & \dots \text{ if a facility is opened at location } j \text{ in time period } h \\ 0 & \dots \text{ otherwise} \end{cases}$

$z_h$  = maximum distance between any customer  $i$  and its nearest open facility in time period  $h$ .



# First MILP formulation

$$\begin{array}{ll}
 (nPC1) \quad \min & \sum_{h \in \mathcal{H}} z_h \quad \text{(objective)} \\
 \text{s.t.} & \sum_{j \in \mathcal{J}} y_{jh} = p_h \quad \forall h \in \mathcal{H} \quad \text{(facility)} \\
 & \sum_{j \in \mathcal{J}} x_{ijh} = 1 \quad \forall i \in \mathcal{I}, h \in \mathcal{H} \quad \text{(assignment)} \\
 & x_{ijh} \leq y_{jh} \quad \forall i \in \mathcal{I}, j \in \mathcal{J}, h \in \mathcal{H} \quad \text{(open)} \\
 & \sum_{j \in \mathcal{J}} d_{ij} x_{ijh} \leq z_h \quad \forall i \in \mathcal{I}, h \in \mathcal{H} \quad \text{(push)} \\
 & y_{jh} \geq y_{jh-1} \quad \forall h \in \mathcal{H} \setminus \{1\} \quad \text{(nesting)} \\
 & x_{ijh}, y_{jh} \in \{0, 1\} \quad \forall i \in \mathcal{I}, j \in \mathcal{J}, h \in \mathcal{H} \quad \text{(binary)} \\
 & z_h \in \mathbb{R}_{\geq 0} \quad \forall h \in \mathcal{H} \quad \text{(non-negativity)}
 \end{array}$$

## Second MILP formulation

$$\begin{aligned}
 (nPC2) \quad & \min \sum_{h \in H} z_h && \text{(objective-y)} \\
 \text{s.t.} \quad & \sum_{j \in J} y_{jh} = p_h && \forall h \in H \quad \text{(facility-y)} \\
 & z_h \geq d_{ij} - \sum_{j': d_{ij'} < d_{ij}} (d_{ij} - d_{ij'}) y_{j'h} && \forall i \in I, j \in J, h \in H \quad \text{(push-y)} \\
 & y_h \geq y_{h-1} && \forall j \in J, h \in H \setminus \{1\} \quad \text{(nesting-y)} \\
 & y_{jh} \in \{0, 1\} && \forall j \in J, h \in H \quad \text{(binary-y)} \\
 & z_h \in \mathbb{R}_{\geq 0} && \forall h \in H \quad \text{(non-negativity-y)}
 \end{aligned}$$

# Third formulation

# Constraint lifting

Theorem [Brandstetter(2023)]

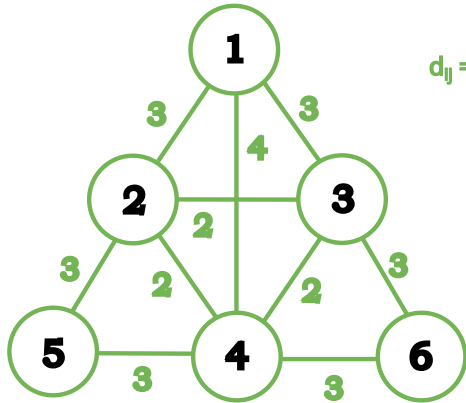
Let  $LB_h$  being a lower bound on the decision variable  $z_h$  of (nPC1) for every  $i \in I, j \in J, h \in H$  then

$$\sum_{j \in J} \max\{LB_h, d_{ij}\} x_{ijh} \leq z_h \quad \forall i \in I, h \in H \quad (\text{lift})$$

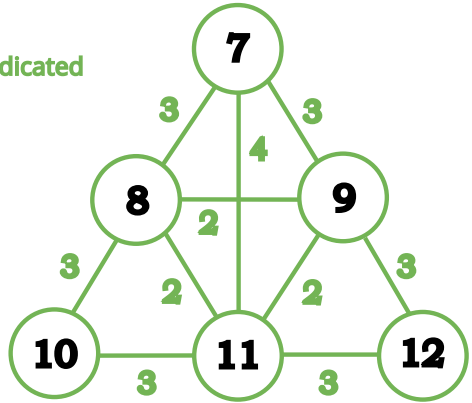
is a valid equality. Theorem is based on Gaar and Sinnl (2022).

- also valid for the y-formulation
- using various techniques to obtain  $LB_h$

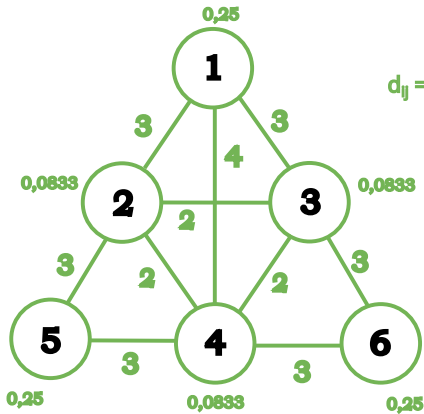
# Constraint lifting / Delete



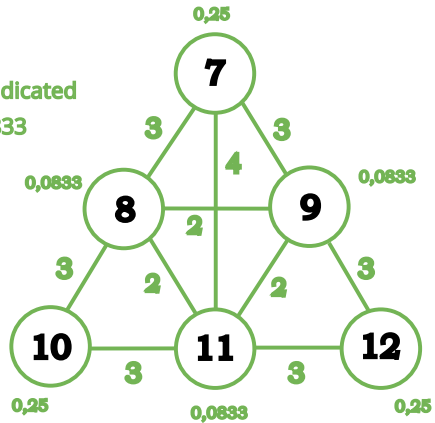
$p = 2$   
 $d_{ij} = 0$   
 $d_{ij} = 4$  if not indicated



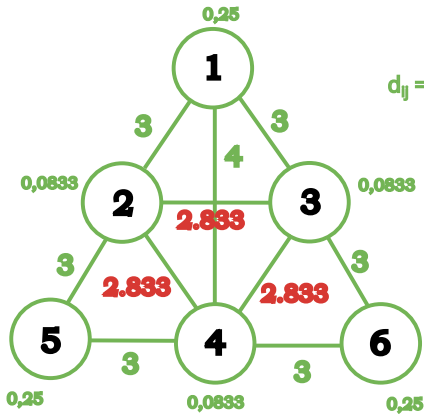
# Constraint lifting / Delete



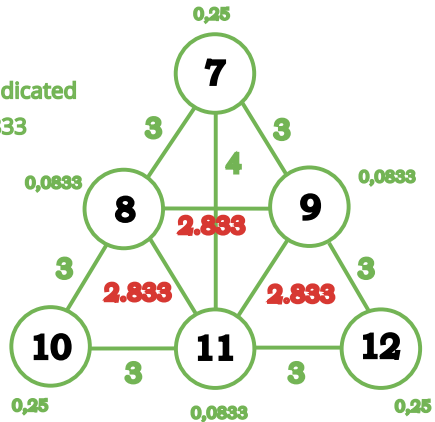
$p = 2$   
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LB = 2.833



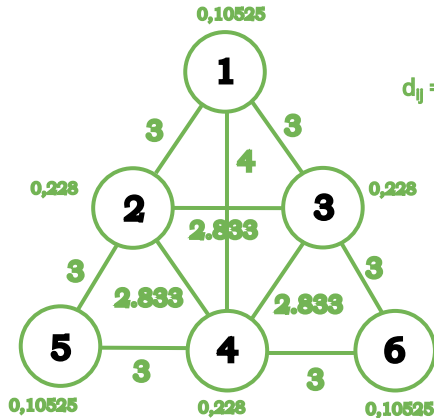
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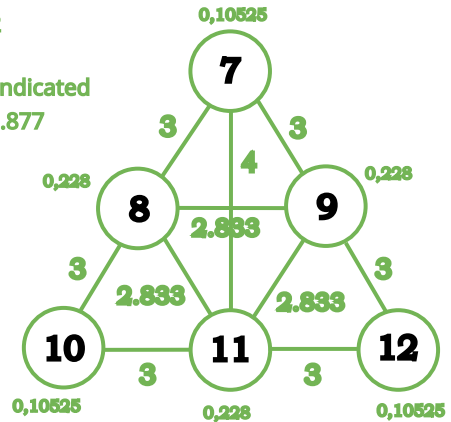
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 $d_{ij} = 0$   
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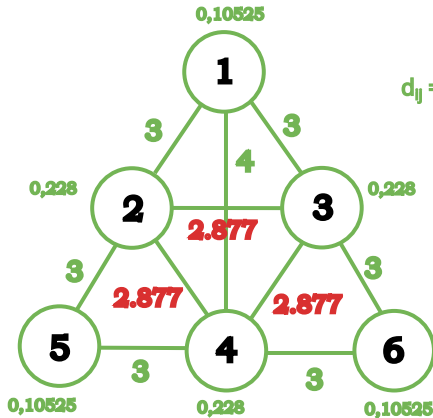


$p = 2$   
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 $LB = 2.877$

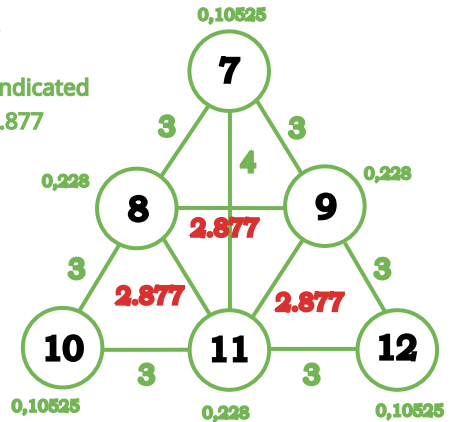




# Constraint lifting / Delete



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- start with lower number of (push) constraints
- add (lift) inequalities
- allows for shorter solving times

# Outline of the results / This should be the main topic for the final presentation



# Data

- data set **PMED** from "A note on solving large  $p$ -median problems" by Beasley (1985)
  - set of 40 instances
  - between 100 and 900 nodes



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  - set of 40 instances
  - between 100 and 900 nodes
- data set **TSPLIB** 2D-Euclidean distances from "TSPLIB – A Traveling Salesman Problem Library" by Reinelt (1991)
  - set of 80 instance
  - between 51 and 18512 nodes
  - rounded to the nearest integer value

# Comparison

- comparing the results between (nPC1) and (nPC2)

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- comparing the results between (nPC1) and (nPC2)
- comparing the different lifting methods
- analysing the solutions of the nested and comparing it to the standard  $p$ -center problem
- analysing the results regarding managerial insights

# On the nested $p$ -center problem



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## Further research topics

- $u$ -space model introduced by Elloumi, Labbé, and Pochet (2004)
- maximal regret as objective function
- trying to improve the performance of the nested  $p$ -median problem?

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