

Minmax Regret Maximal Covering on Networks with Edge Demands

Marta Baldomero Naranjo ¹ Jörg Kalcsics ²
Antonio M. Rodríguez Chía ¹

¹Universidad de Cádiz

²University of Edinburgh

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 - Introduction
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Introduction

- Single-facility location problem on a network.

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- Our aim is to maximise the covered demand.

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- Single-facility location problem on a network.
- Our aim is to maximise the covered demand.
- The demand distributed along the edges.
- The demand is uncertain with only a known interval estimation.

Introduction

Objective

Given:

- $G = (V, E)$: an undirected network; $|V| = n$ and $|E| = m$.
- R : coverage radius.
- For $e \in E$, two non-negative continuous functions $lb_e : [0, 1] \rightarrow \mathbb{R}^+$ and $ub_e : [0, 1] \rightarrow \mathbb{R}^+$ that specify the minimal and maximal demand along the edge.

The aim is to locate a facility along the network that minimise the *worst-case* of coverage loss.

Criteria

- Coverage criterion: Maximal Covering.



R. Church and C. ReVelle.

The maximal covering location problem.

Papers of the Regional Science Association, 3:101–118,
1974.

Criteria

- Coverage criterion: Maximal Covering.

Let $x \in G$ be a facility:

- $z \in G$ is *covered* by x , if $d(x, z) \leq R$.
- $C(x) := \{z \in G \mid d(x, z) \leq R\}$ is the *coverage area* of x .

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The covered demand on an edge $e \in E$ by x for a specific demand realisation w :

$$g_e(x, w) = \int_{y=(e,t) \in C_e(x)} w_e(t) dt. \quad (1)$$

The total amount of covered demand on the network:

$$g(x, w) = \sum_{e \in E} g_e(x, w). \quad (2)$$

Criteria

- Coverage criterion: Maximal Covering.

$$\max_{x \in G} g(x) = \sum_{e \in E} \int_{y=(e,t) \in C_e(x)} w_e(t) dt. \quad (1)$$



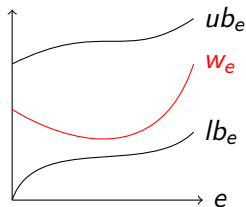
O. Berman, J. Kalcsics, and D. Krass.

On covering location problems on networks with edge demand.

Computers & Operations Research, 74:214–227, 2016.

Criteria

- Coverage criterion: Maximal Covering.
- Uncertainty Demand: MinMax Regret. Minimise the *worst-case* of coverage loss.



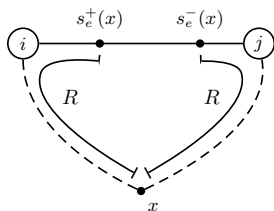
$$r^* = \min_{x \in X} \max_{lb \leq w \leq ub} \left(\max_{y \in G} g(y, w) - g(x, w) \right).$$

Edge coverage functions

If $x \notin e = [i, j]$

$$s_e^+(x) = \min \left\{ 1, \max \left\{ 0, \frac{R - d(x, i)}{\ell_e} \right\} \right\}$$

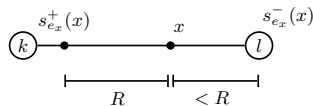
$$s_e^-(x) = \max \left\{ 0, \min \left\{ 1, 1 - \frac{R - d(x, j)}{\ell_e} \right\} \right\}$$



If $x \in e_x = [k, l]$

$$s_{e_x}^+(x) = \max \left\{ 0, \frac{d(x, k) - R}{\ell_{e_x}} \right\}$$

$$s_{e_x}^-(x) = \min \left\{ 1, 1 - \frac{d(x, l) - R}{\ell_{e_x}} \right\}$$

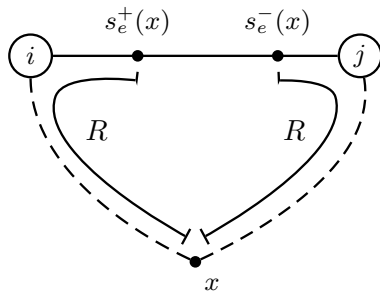


Coverage area $C_e(x)$ of partially covered edges

For $e \in E \setminus \{e_x\}$

$C_e(x) = (e, [0, s_e^+(x)] \cup [s_e^-(x), 1])$, if $0 < s_e^+(x) < s_e^-(x) < 1$

$E^P(x) = E \setminus \{E^c(x) \cup E^u(x) \cup \{e_x\}\}$.

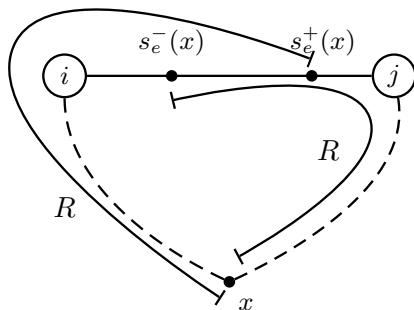


Coverage area $C_e(x)$ of covered edges

For $e \in E \setminus \{e_x\}$

$C_e(x) = e$, if $s_e^-(x) \leq s_e^+(x)$,

$E^c(x) = \{e \in E \setminus \{e_x\} \mid s_e^-(x) \leq s_e^+(x)\}$,

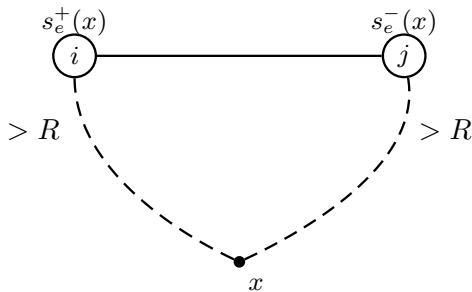


Coverage area $C_e(x)$ of uncovered edges

For $e \in E \setminus \{e_x\}$

$C_e(x) = \emptyset$, if $s_e^+(x) = 0, s_e^-(x) = 1$,

$E^u(x) = \{e \in E \setminus \{e_x\} \mid s_e^+(x) = 0 \text{ and } s_e^-(x) = 1\}$,



Coverage area

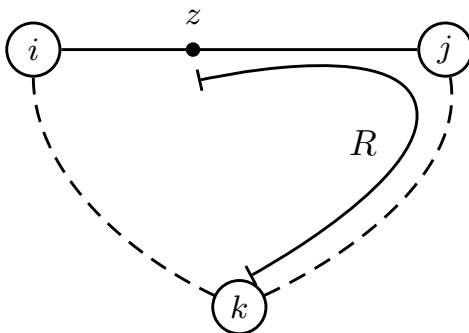
The total coverage can now be written as

$$\begin{aligned}
 g(x, w) = & \int_{s_{e_x}^+(x)}^{s_{e_x}^-(x)} w_{e_x}(u) du + \sum_{e \in E^c(x)} \int_0^1 w_e(u) du \\
 & + \sum_{e \in E^p(x)} \left(\int_0^{s_e^+(x)} w_e(u) du + \int_{s_e^-(x)}^1 w_e(u) du \right).
 \end{aligned}$$

Singularity points (PP)

- Network intersect points $NP := \bigcup_{e \in E} NP_e$

$$NP_e := \{z \in e \mid \exists k \in V : d(z, k) = R\};$$

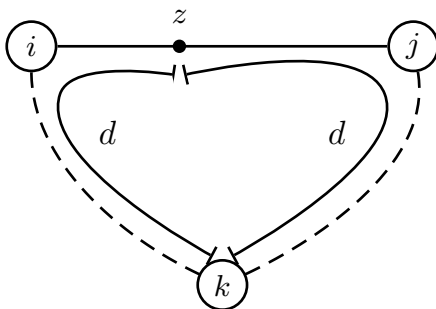


Singularity points (PP)

- Bottleneck points $BP := \bigcup_{e \in E} BP_e$

$$BP_e := \{z \in e \mid \exists k \in V : tl_e + d(k, i) = (1 - t)l_e + d(k, j)\}$$

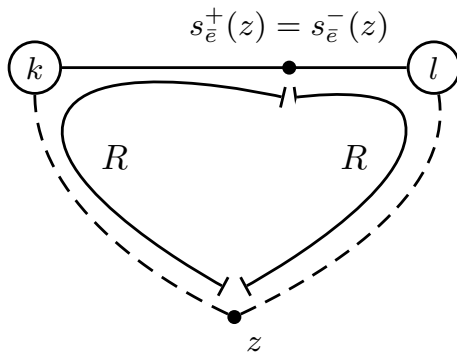
$$e = [i, j]$$



Singularity points (PP)

- Exact coverage point $EP := \bigcup_{e \in E} EP_e$

$$EP_e := \{z \in e \mid \exists \bar{e} \in E : 0 < s_{\bar{e}}^+(z) = s_{\bar{e}}^-(z) < 1\};$$



Singularity points (PP)

- $PP_e := \{i, j\} \cup NP_e \cup BP_e \cup EP_e$

There are at most $\mathcal{O}(m)$ partition points on each $e \in E$.

- $PP := \bigcup_{e \in E} PP_e$

There are $\mathcal{O}(m^2)$ on the whole network.

Lemma

$s_e^+(x)$ and $s_e^-(x)$, $e \in E$, are continuous and piecewise linear functions over $x \in e_x$ with a constant number of pieces.

Edge coverage functions and singularity points

Theorem

Let $e_x \in E, x \in [z^1, z^2]$ such that $z^1, z^2 \in PP_{e_x}$.

- 1 The sets $E^c(x)$, $E^u(x)$, and $E^p(x)$ are identical for $x \in [z^1, z^2]$.
- 2 $s_e^+(x)$ and $s_e^-(x)$ have a unique linear representation for $x \in [z^1, z^2]$.
- 3 $g_e(x, w)$, $e \in E$, have a unique representation, for $x \in [z^1, z^2]$ and w a non-negative continuous demand function.

Linear Demand realisation

Let $lb_e(t) = a_e^{lb} + b_e^{lb} \cdot t$, $ub_e(t) = a_e^{ub} + b_e^{ub} \cdot t$, and $w_e(t) = a_e^w + b_e^w \cdot t$.

Linear Demand realisation

Let $lb_e(t) = a_e^{lb} + b_e^{lb} \cdot t$, $ub_e(t) = a_e^{ub} + b_e^{ub} \cdot t$, and $w_e(t) = a_e^w + b_e^w \cdot t$.

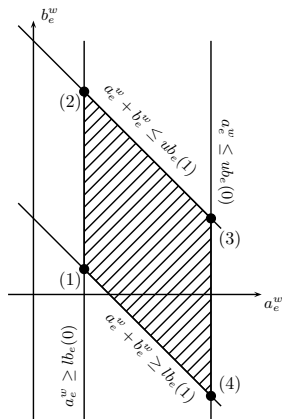
$$a_e^w \geq lb_e(0) = a_e^{lb}$$

$$a_e^w \leq ub_e(0) = a_e^{ub}$$

$$a_e^w + b_e^w \leq ub_e(1) = a_e^{ub} + b_e^{ub}$$

$$a_e^w + b_e^w \geq lb_e(1) = a_e^{lb} + b_e^{lb}$$

$$a_e^w \geq 0, b_e^w \in \mathbb{R}$$



Worst-case demand realisation

Theorem

The worst-case demand realisation for a fixed x, y , and e such that $x, y \in G$ and $e \in E$ can be obtained by solving the following linear program:

$$r_e(x, y) = \max_{lb_e \leq w_e \leq ub_e} a_e^w (c_e(y) - c_e(x)) + \frac{1}{2} b_e^w (\bar{c}_e^2(y) - \bar{c}_e^2(x)) \quad (1)$$

$$c_e(x) = \begin{cases} 1, & \text{if } e \in E^c(x), \\ 1 - (s_e^-(x) - s_e^+(x)), & \text{if } e \in E^p(x), \\ 0, & \text{if } e \in E^u(x), \\ s_{e_x}^-(x) - s_{e_x}^+(x), & \text{if } e = e_x, \end{cases} \quad \bar{c}_e^2(x) = \begin{cases} 1, & \text{if } e \in E^c(x), \\ 1 - ((s_e^-(x))^2 - (s_e^+(x))^2), & \text{if } e \in E^p(x), \\ 0, & \text{if } e \in E^u(x), \\ (s_{e_x}^-(x))^2 - (s_{e_x}^+(x))^2, & \text{if } e = e_x. \end{cases}$$

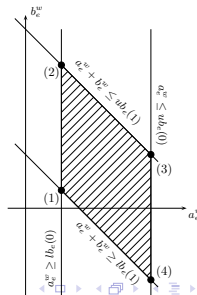
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- 1 $(a_e^{lb}, b_e^{lb}),$
- 2 $(a_e^{lb}, a_e^{ub} + b_e^{ub} - a_e^{lb}),$
- 3 $(a_e^{ub}, b_e^{ub}),$
- 4 $(a_e^{ub}, a_e^{lb} + b_e^{lb} - a_e^{ub}).$



Worst-case demand realisation

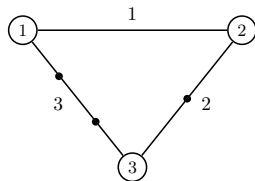
Theorem

An optimal solution of (1), (a_e^{w*}, b_e^{w*}) , is given by the first column of the following table whenever the conditions of columns 2-4 are fulfilled.

(a_e^{w*}, b_e^{w*})	Conditions		
	$c_e(y) - c_e(x)$	$\bar{c}_e^2(y) - \bar{c}_e^2(x)$	$(\bar{c}_e^2(y) - \bar{c}_e^2(x)) - 2(c_e(y) - c_e(x))$
(a_e^{lb}, b_e^{lb})	≤ 0	≤ 0	≥ 0
$(a_e^{lb}, a_e^{ub} + b_e^{ub} - a_e^{lb})$	≥ 0	≥ 0	≥ 0
	≤ 0	≥ 0	-
(a_e^{ub}, b_e^{ub})	≥ 0	≥ 0	≤ 0
$(a_e^{ub}, a_e^{lb} + b_e^{lb} - a_e^{ub})$	≤ 0	≤ 0	≤ 0
	≥ 0	≤ 0	-

Algorithm

- Determine the set PP of partition points.



Let $R = 1$,

$$lb_{[1,2]}(t) = 3 - 3t, \quad ub_{[1,2]}(t) = 15 + 7t,$$

$$lb_{[2,3]}(t) = 6 - t, \quad ub_{[2,3]}(t) = 12 + 9t,$$

$$lb_{[1,3]}(t) = 2 + 3t, \quad ub_{[1,3]}(t) = 8 + 10t.$$

Algorithm

For $e = [i, j] \in E$:

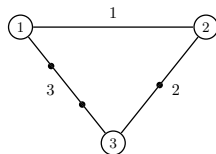
- Sort the partition $PP_e = \{z^1, \dots, z^{|PP_e|-1}\}$ in non-decreasing distance from node i .
- Derive the representation of the edge coverage functions over each sub-edge $[z^q, z^{q+1}]$, for $q \in \mathcal{I}_e := \{1, \dots, |PP_e|-1\}$.

For $x_1 \in ([1, 2], t_1)$ the edge coverage functions are given by

$$s_{[1,2]}^+(x_1) = 0, \quad s_{[1,2]}^-(x_1) = 1,$$

$$s_{[2,3]}^+(x_1) = \frac{t_1}{2}, \quad s_{[2,3]}^-(x_1) = 1,$$

$$s_{[1,3]}^+(x_1) = \frac{1-t_1}{3}, \quad s_{[1,3]}^-(x_1) = 1.$$



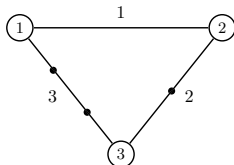
Algorithm

For $x_2 \in ([2, 3], t_2)$ the edge coverage functions are given by

$$s_{[1,2]}^+(x_2) = 0, \quad s_{[1,2]}^-(x_2) = \begin{cases} 2t_2, & \text{if } t_2 \leq \frac{1}{2}, \\ 1, & \text{o/w} \end{cases},$$

$$s_{[2,3]}^+(x_2) = \begin{cases} \frac{2t_2-1}{2}, & \text{if } t_2 \geq \frac{1}{2}, \\ 0, & \text{o/w} \end{cases}, \quad s_{[2,3]}^-(x_2) = \begin{cases} \frac{1+2t_2}{2}, & \text{if } t_2 \leq \frac{1}{2}, \\ 1, & \text{o/w} \end{cases},$$

$$s_{[1,3]}^+(x_2) = 0, \quad s_{[1,3]}^-(x_2) = \begin{cases} \frac{4-2t_2}{3}, & \text{if } t_2 \geq \frac{1}{2}, \\ 1, & \text{o/w} \end{cases}.$$



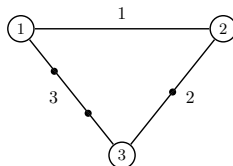
Algorithm

For $x_3 \in ([1, 3], t_3)$ the edge coverage functions are given by

$$s_{[1,2]}^+(x_3) = \begin{cases} 1 - 3t_3 & \text{if } t_3 \leq \frac{1}{3} \\ 0 & \text{o/w} \end{cases}, \quad s_{[1,2]}^-(x_3) = 1,$$

$$s_{[2,3]}^+(x_3) = 0, \quad s_{[2,3]}^-(x_3) = \begin{cases} \frac{4-3t_3}{2} & \text{if } t_3 \geq \frac{2}{3} \\ 1 & \text{o/w} \end{cases},$$

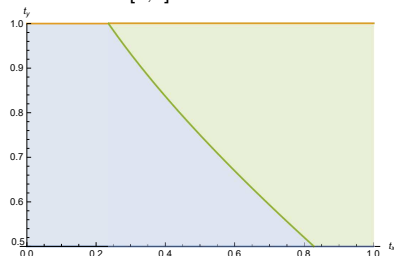
$$s_{[1,3]}^+(x_2) = \begin{cases} \frac{3t_3-1}{3} & \text{if } t_3 \geq \frac{1}{3} \\ 0 & \text{o/w} \end{cases}, \quad s_{[1,3]}^-(x_3) = \begin{cases} \frac{3t_3+1}{3} & \text{if } t_3 \leq \frac{2}{3} \\ 1 & \text{o/w} \end{cases}.$$



Algorithm

- 1: **for** $e_x \in E$ **do**
- 2: **for** $i \in \mathcal{I}_{e_x}$ **do**
- 3: **for** $e_y \in E$ **do**
- 4: **for** $j \in \mathcal{I}_{e_y}$ **do**
- 5: Generate the subdivision in the rectangle $[z^i, z^{i+1}] \times [z^j, z^{j+1}]$ by the arcs defining the conditions of the worst case demand realisation for any $e \in E$.

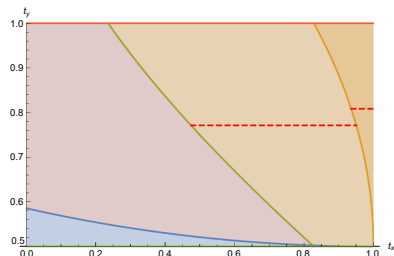
Let $e = [2, 3]$, $x = ([1, 2], t_x)$,
 $y = ([2, 3], t_y)$, $\frac{1}{2} \leq t_y \leq 1$.
 Cells for $r_{[2,3]}(x, y)$:



Algorithm

- 1: **for** $e_x \in E$ **do**
- 2: **for** $i \in \mathcal{I}_{e_x}$ **do**
- 3: **for** $e_y \in E$ **do**
- 4: **for** $j \in \mathcal{I}_{e_y}$ **do**
- 5: Let $\mathcal{C}_{e_x e_y}^{ij}$ be the family of arcs defined by:
 - boundaries of the cells previously obtained.
 - For any cell where $r(x, y)$ is concave, the intersection of the curve $\frac{\partial r}{\partial y}(x, y) = 0$ with that cell.

Let $x = ([1, 2], t_x)$ and $y = ([2, 3], t_y), \frac{1}{2} \leq t_y \leq 1$.
 Cells for $r(x, y)$,



Algorithm

- 1: **for** $e_x \in E$ **do**
- 2: **for** $i \in \mathcal{I}_{e_x}$ **do**
- 3: Obtain the upper envelope, $h_{e_x}^i(x)$, of $r(x, y(x))$ of the arcs contained in $\bigcup_{e_y \in E, j \in \mathcal{I}_{e_y}} C_{e_x e_y}^{ij}$.
- Find the minimum x_i^* of $h_{e_x}^i(x)$ over $[z^i, z^{i+1}]$.
- 4: **if** $h_{e_x}^i(x_i^*) < r(x^*)$ **then**
- 5: set $x^* := x_i^*$, $r(x^*) = h_{e_x}^i(x_i^*)$.

Algorithm

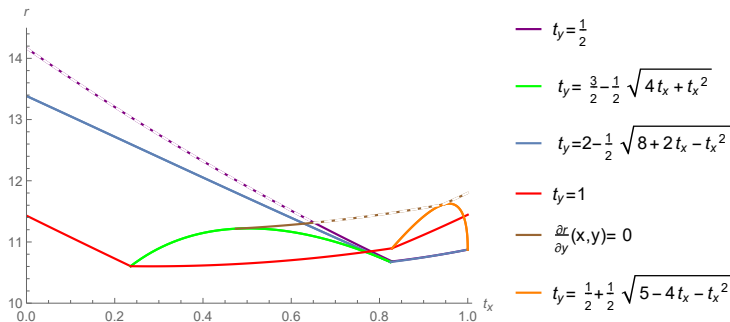


Figure: $r(x, y(x))$, $x = ([1, 2], t_x)$ and $y = ([2, 3], t_y)$, $\frac{1}{2} \leq t_y \leq 1$.

Algorithm

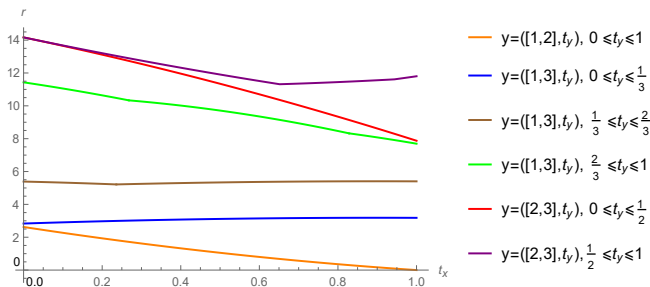


Figure: Upper envelope of $r(x, y(x))$, $x = ([1, 2], t_x)$.

The minimum value of r is 11.3153, where $x_{[1,2]}^* = ([1, 2], 0.6517)$.
 This should be repeated for each $x \in [z^i, z^{i+1}]$, where $i \in \mathcal{I}_{e_x}$ and $e_x \in E$.

Algorithm

Theorem

The single facility MinMax Regret Maximal Covering Location Problem on a network with edge linear demand realisations can be solved exactly in $\mathcal{O}(m^2 \lambda_{10}(m^4))$ time using the previous algorithm, where $\lambda_s(m) = \max\{|U| : U \text{ is a } DS(m, s) - \text{sequence}\}$.

Theorem

The single facility MinMax Regret Maximal Covering Location Problem on a network with edge constant demand realisations can be solved exactly in $\mathcal{O}(m^2 \lambda_3(m^4))$ time using the previous algorithm, where $\lambda_s(m) = \max\{|U| : U \text{ is a } DS(m, s) - \text{sequence}\}$.

Conclusions

- Although the majority of problems become NP-hard in the minmax regret version, we propose a polynomial time algorithm for solving the single-facility MinMax Regret Maximal Covering Location Problem on a network where the demand is
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Conclusions

- Although the majority of problems become NP-hard in the minmax regret version, we propose a polynomial time algorithm for solving the single-facility MinMax Regret Maximal Covering Location Problem on a network where the demand is
 - distributed along the edges,
 - constant or linear functions,
 - uncertain with only a known interval estimation.

Future work

Potential avenues for future research:

- Other kind of demand realisation functions.

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- Other kind of demand realisation functions.
- Multi-facility location version of the problem.
- Apply a different criterion of coverage, e.g. the gradual covering.

Thanks for your attention