

# On Lower Bounds for the Online Facility Location Problem and its Multi-commodity extension

Seminar Talk for Methods of Algorithm Design

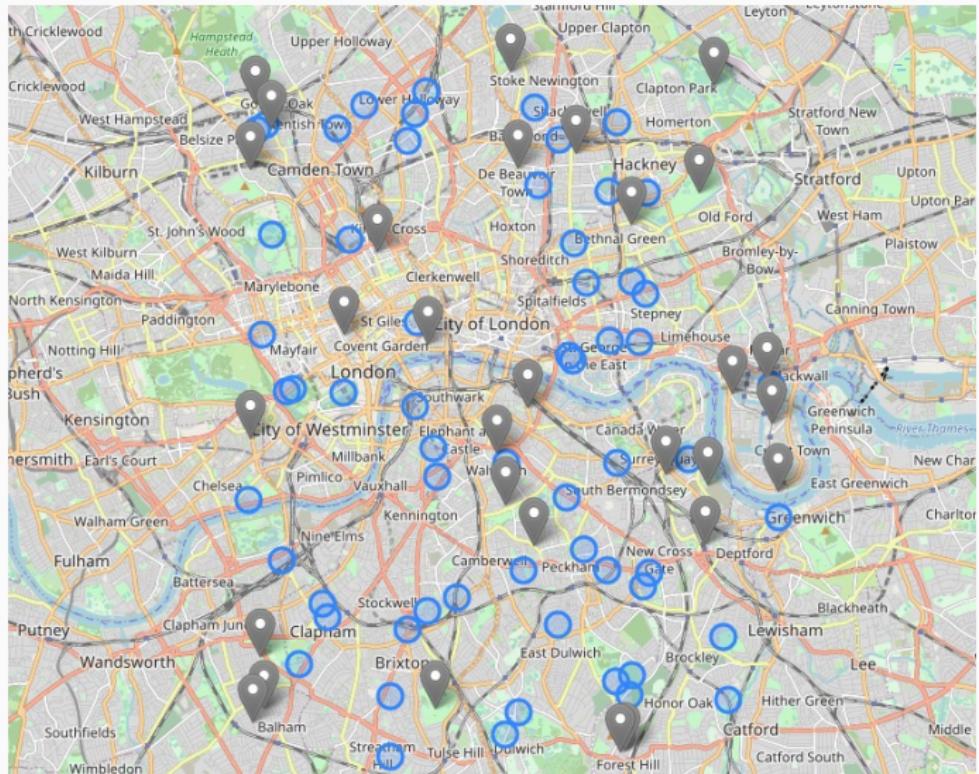
---

Marcus Rottschäfer

20.07.2021

Department of Informatics, University of Hamburg

# What is Facility Location good for?



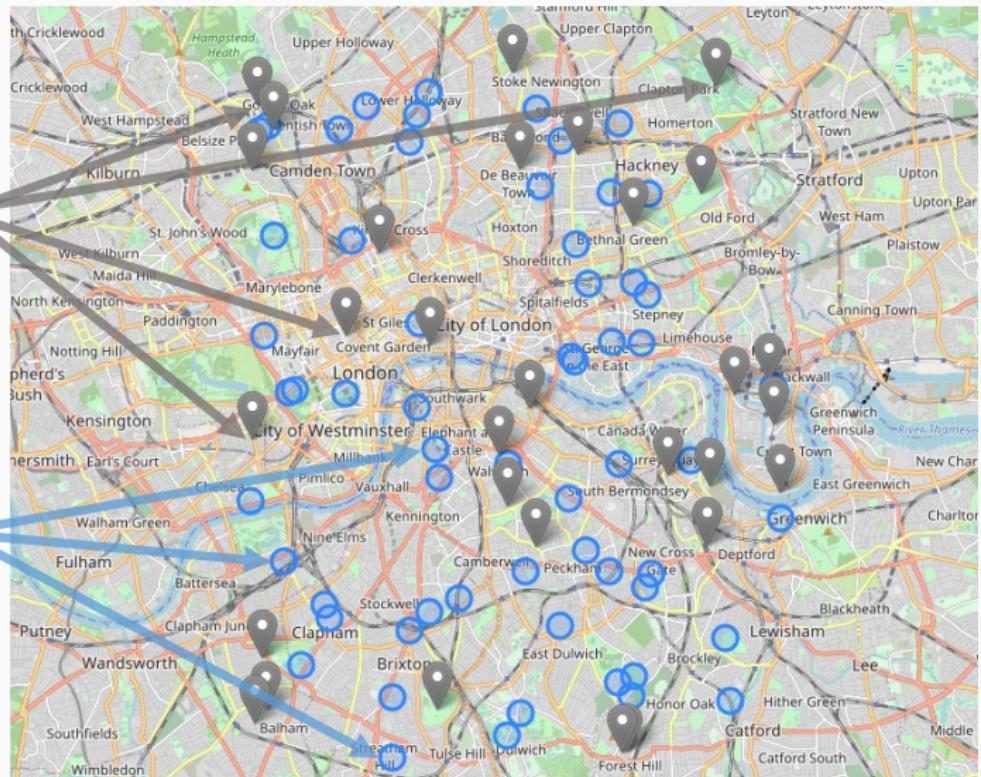
# What is Facility Location good for?



Hospitals  
(Facilities)



Ambulances  
(Clients)



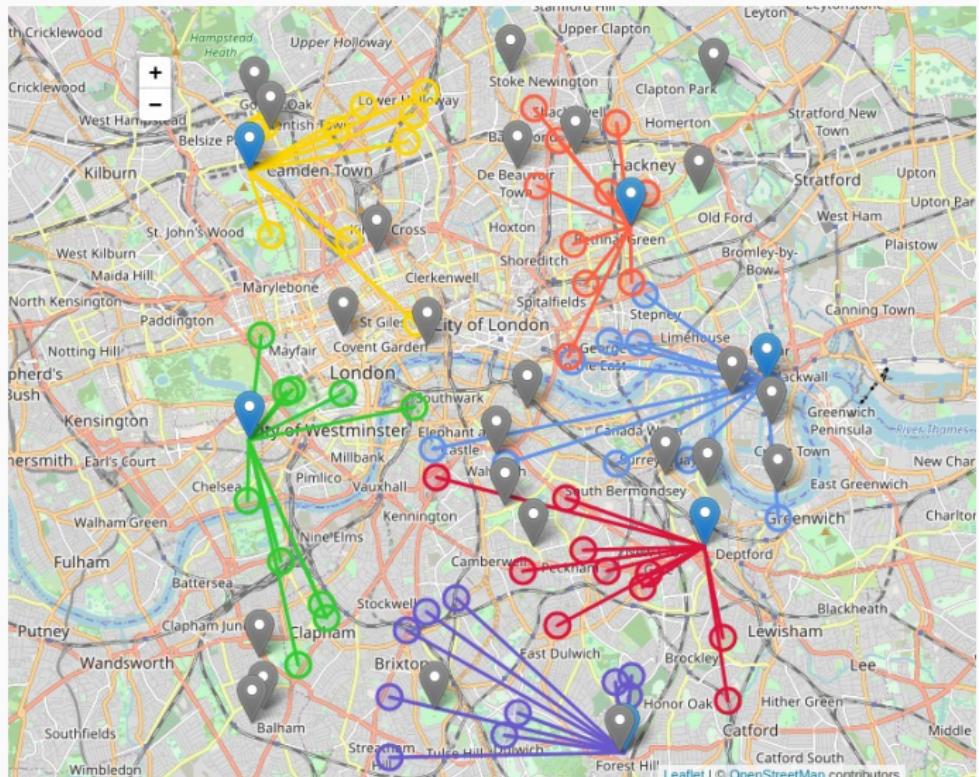
# What is Facility Location good for?



Hospitals  
(Facilities)



Ambulances  
(Clients)



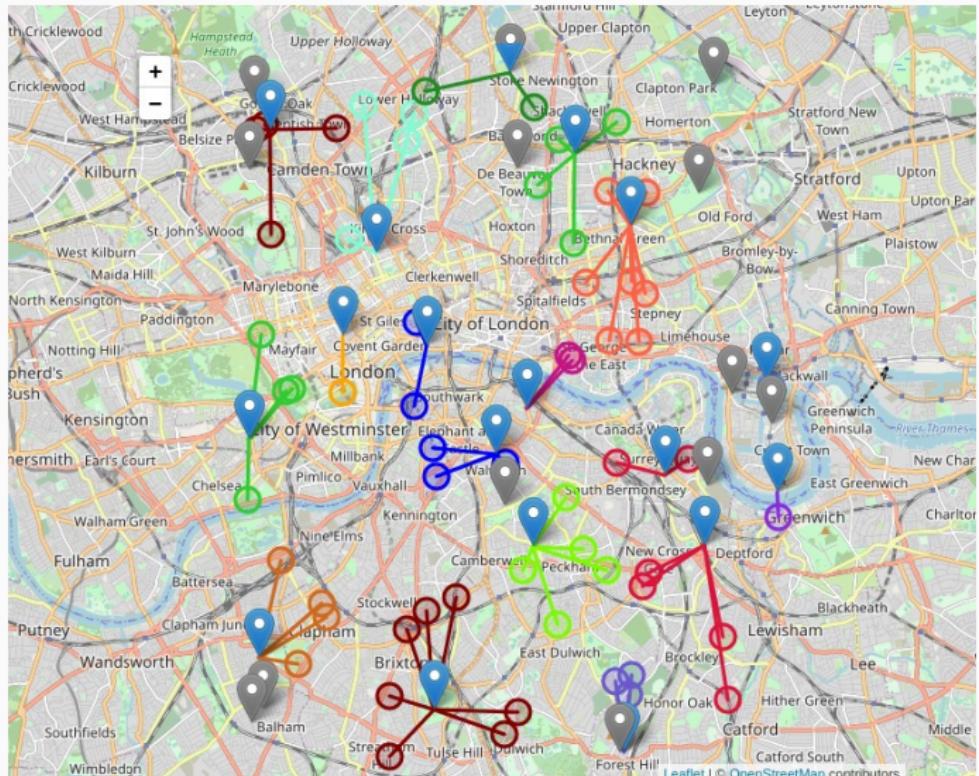
# What is Facility Location good for?



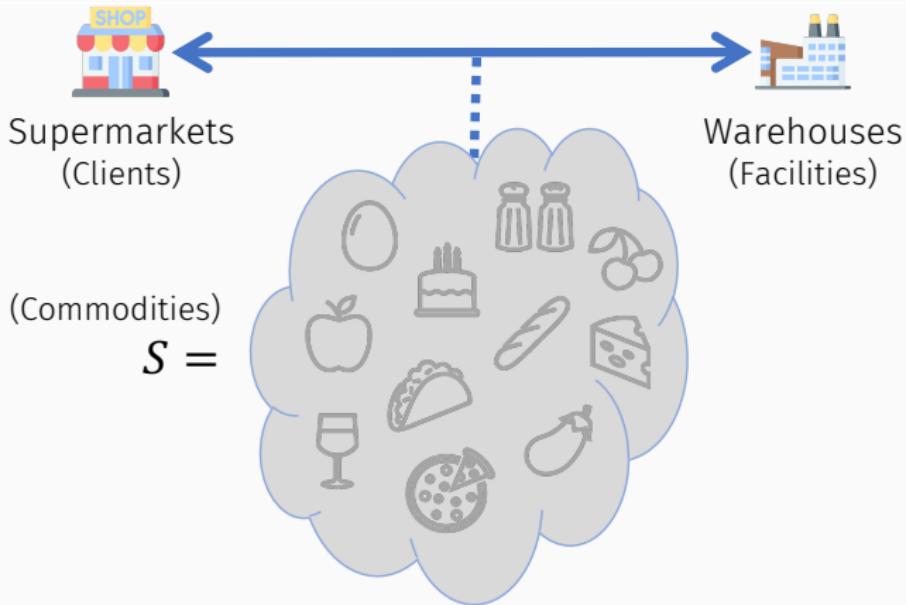
Hospitals  
(Facilities)



Ambulances  
(Clients)



# Why should we extend it to Multi-commodities?



## Multi-commodity Facility Location vs. Facility Location

- Clients can have **multiple** needs  $s \subseteq S$
- Facilities provide **commodities**  $\sigma \subseteq S$

# Table of contents

1. Introduction

2. Lower Bounds

    Lower Bound for OFL

    Lower Bound for OMCFL

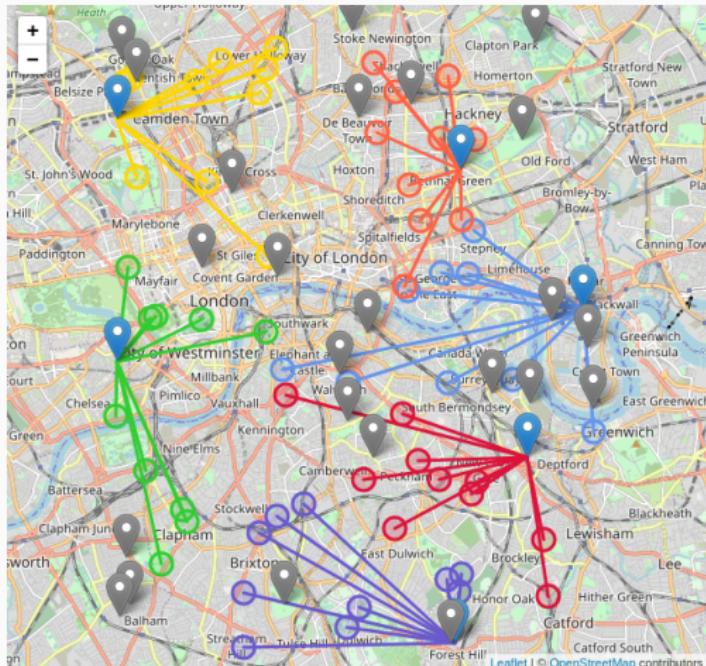
3. What more is there?

4. Conclusion

## Formalities

---

# Online Facility Location Problem (OFLP)

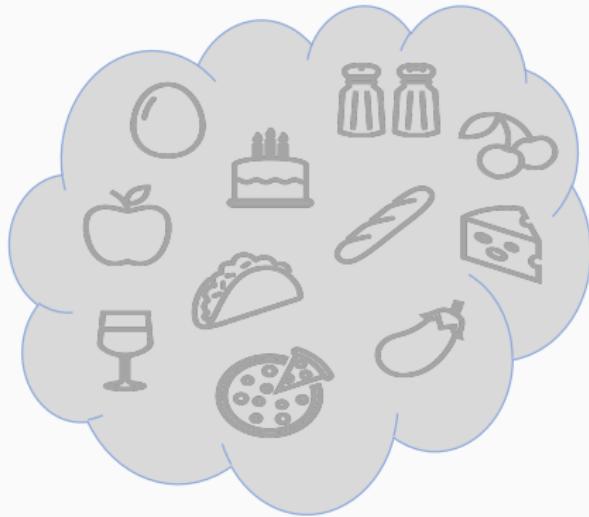


- Metric space  
 $\mathcal{M} = (M, d)$   
 $d : M \times M \rightarrow \mathbb{R}^+$
- Requests & Facilities  
 $r \in R, f \in F$ , anywhere in  $M$   
Request sequence  $\sigma$
- Construction costs  
 $f_m$ : Open  $f$  at  $m \in M$
- Provisioning costs  
 $d(r, f)$ : Serve  $r$  with  $f$

Serve  $\sigma = (r_1, r_2, \dots)$  and construct facilities  $F$  such that  
 $\mathcal{A}(\sigma) := \sum_{f_m \in F} f_m + \sum_{r \in \sigma} d(r, f)$  is minimized.

# Online Multi-commodity Facility Location Problem (OMCFLP)

What changes with the Multi-commodity extension?



- **Commodities**  
predefined set  $S$
- **Requests**  $r \in R$   
additional  $s_r \subseteq S$
- **Facilities**  
 $f^\psi$  with  $\psi \subseteq S$   
(configuration of  $f$ )

Construction costs:

- depend on configuration  $\psi : f_m^\psi$ :

Provisioning costs:

- Sum of sums (all facilities jointly serving  $s_r$ )

## Further Remarks

- Start **without** facilities
- Can't **close** facilities
- Requests only connected to facilities **capable of serving** them
- Always serve  $r$  with the **closest** facility
- Consider the **metric, uncapacitated** case only

# Related Work

---

## Offline Cases

- NP-hard on general graphs :(
- 1.463-approximation lower bound [3]
- Li [4] gave 1.488-approximation algorithm
- Multi-commodity:  $\Theta(\log |S|)$ -approximation [6], but constant approximation [7] (linear facility cost function)

## Lower Bounds

---

# Overview of the Results

What **Competitive Ratio** can we expect?

- **Bad news:** we can't hope for a *constant CR* :(

## Theorem (Online Facility Location Problem)

No randomized algorithm for the OFLP can achieve a CR better than  $\Omega(\frac{\log n}{\log \log n})$  against an oblivious adversary, even if the metric space is a line segment.

## Theorem (Online Multi-commodity Facility Location Problem)

No randomized algorithm for the OMCFLP can achieve a CR better than  $\Omega(\sqrt{|S|} + \frac{\log n}{\log \log n})$  against an oblivious adversary, even if the metric space is a line segment.

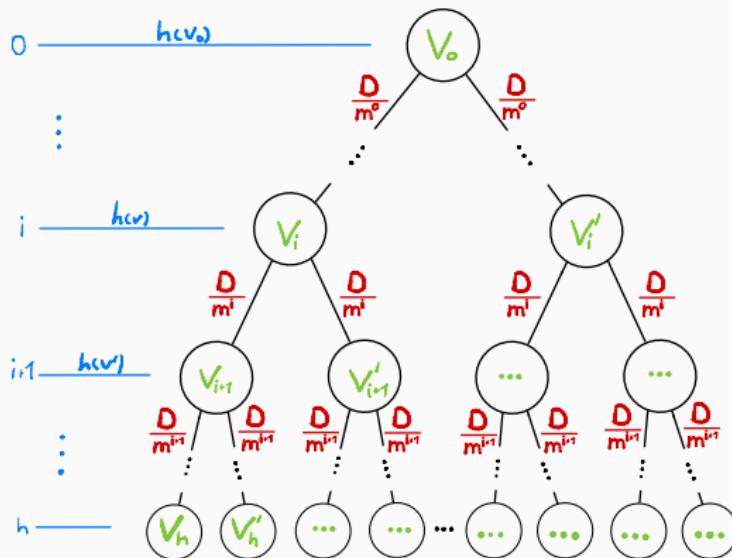
# Lower Bound for OFL: Proof Overview

## Structure of Proof:

- Limit to uniform facility costs & single-facility solutions only
- Construct special tree-structure metric space  $T$
- Provide input  $\sigma$  s.t. lower bound holds on  $T$
- Project  $T$  on any line segment

## Lower Bound for OFL: Proof (1/5))

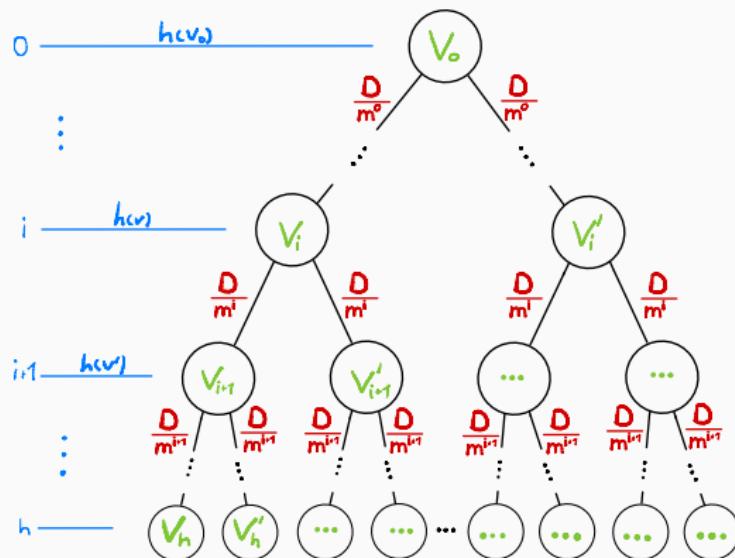
(Tree-structure) Metric space  $T$



Let our metric space  $T$  be a full, binary  $m$ -HST.

# Lower Bound for OFL: Proof (1/5)

(Tree-structure) Metric space  $T$



Properties

- Both children have same distance
- Distance drops by factor  $m$
- All leaves at height  $h$
- Represents finite metric space

Let our metric space  $T$  be a full, binary  $m$ -HST.

## Lower Bound for OFL: Proof (2/5)

### Claim (Distances in $T$ )

Let  $T_v$  be the subtree rooted at node  $v \in T$ . Then the distance from  $v$  to any node  $u \notin T_v$  is at least  $\frac{D}{m^{h(v)} - 1}$ . The distance from  $v$  to any node  $w \in T_v$  is not larger than  $\frac{m}{m-1} \frac{D}{m^{h(v)}}$ .

## Lower Bound for OFL: Proof (2/5)

### Claim (Distances in $T$ )

Let  $T_v$  be the subtree rooted at node  $v \in T$ . Then the distance from  $v$  to any node  $u \notin T_v$  is at least  $\frac{D}{m^{h(v)-1}}$ . The distance from  $v$  to any node  $w \in T_v$  is not larger than  $\frac{m}{m-1} \frac{D}{m^{h(v)}}$ .

### Proof.

1. See picture.
2. The longest path must end at leaf node. Hence

$$\max_{w \in T_v} d(v, w) = \sum_{i=h(v)}^{h-1} \frac{D}{m^i} = \frac{D}{m^{h(v)}} \sum_{i=0}^{h-1-h(v)} \frac{1}{m^i} \leq \frac{D}{m^{h(v)}} \frac{m}{m-1}$$

## Lower Bound for OFL: Proof (2/5)

### Claim (Distances in $T$ )

Let  $T_v$  be the subtree rooted at node  $v \in T$ . Then the distance from  $v$  to any node  $u \notin T_v$  is at least  $\frac{D}{m^{h(v)} - 1}$ . The distance from  $v$  to any node  $w \in T_v$  is not larger than  $\frac{m}{m-1} \frac{D}{m^{h(v)}}$ .

### Proof.

1. See picture.
2. The longest path must end at leaf node. Hence

$$\max_{w \in T_v} d(v, w) = \sum_{i=h(v)}^{h-1} \frac{D}{m^i} = \frac{D}{m^{h(v)}} \sum_{i=0}^{h-1-h(v)} \frac{1}{m^i} \leq \frac{D}{m^{h(v)}} \frac{m}{m-1}$$

- Now we'd construct **bad** (for any randomized algorithm) input sequence  $\sigma$ .
- Instead we apply **Yao's Principle**.

## Lower Bound for OFL: Proof (3/5)

### Bad Request Sequence

$$\sigma = \left( \underbrace{v_0}_{1} \underbrace{v_1 \dots v_1}_m \dots \underbrace{v_i \dots v_i}_{m^i} \underbrace{v_{i+1} \dots v_{i+1}}_{m^{i+1}} \dots \underbrace{v_h \dots v_h}_{m^h} \right)$$

- At phase transition  $v_i \rightarrow v_{i+1}$ : Choose  $v_{i+1}$  u.i.r from  $v_i$ 's children
- Yields request sequence distribution  $\mathbb{Q}$

## Lower Bound for OFL: Proof (3/5)

### Bad Request Sequence

$$\sigma = \left( \underbrace{v_0}_{1} \underbrace{v_1 \dots v_1}_m \dots \underbrace{v_i \dots v_i}_{m^i} \underbrace{v_{i+1} \dots v_{i+1}}_{m^{i+1}} \dots \underbrace{v_h \dots v_h}_{m^h} \right)$$

- At phase transition  $v_i \rightarrow v_{i+1}$ : Choose  $v_{i+1}$  u.i.r from  $v_i$ 's children
- Yields request sequence distribution  $\mathbb{Q}$

### Optimal Algorithm's Costs (Upper Bound):

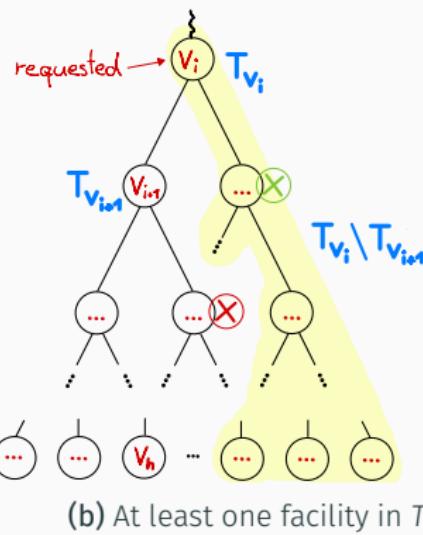
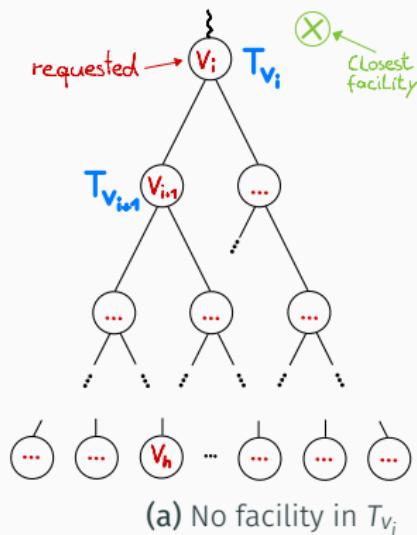
- $\mathcal{A}$ : (1) Open facility at leaf  $v_h$  (2) Serve  $\sigma$ .

$$\begin{aligned} OPT(\sigma) &\leq \mathcal{A}(\sigma) = f_{v_h} + \sum_{i=0}^h \sum_{j=1}^{m^i} d(v_h, \sigma_{ij}) \stackrel{3}{\leq} f_{v_h} + \sum_{i=0}^{h-1} m^i \frac{D}{m^i} \frac{m}{m-1} \\ &= f + hD \frac{m}{m-1} \end{aligned}$$

# Lower Bound for OFL: Proof (4/5)

## Deterministic Algorithm's Cost (Lower Bound):

- Let  $\mathcal{ALG}$  be ANY deterministic algorithm run on  $\mathbb{Q}$
- Right before processing phase  $i + 1$  (fixes  $T_{v_i}$ ):
- Consider **expected costs** for  $T_{v_i}$ , but not for  $T_{v_{i+1}}$ :  $\mathbb{E}[\mathcal{ALG}(\sigma_i)|T_{v_i}]$



(a) exp. costs  $\geq \frac{D}{m^{i-1}} m^i + 0 = Dm$

(b) exp. costs  $\geq 0 + \frac{1}{2}f + \frac{1}{2}0 = \frac{f}{2}$

## Lower Bound for OFL: Proof (5/5)

---

- Hence for the expected costs of  $\mathcal{ALG}$ :

$$\mathbb{E}[\mathcal{ALG}(\sigma_i) | T_{v_i}] \geq \min(Dm, \frac{f}{2}) + \mathbb{E}[\mathcal{ALG}(\sigma_{i-1}) | T_{v_{i-1}}]$$

- We go through  $h + 1$  phases:  $0 \leq i \leq h - 1 + \text{last phase } h$

$$\mathbb{E}[\mathcal{ALG}(\sigma)] \geq h \min(Dm, \frac{f}{2}) + \min(Dm, f)$$

## Lower Bound for OFL: Proof (5/5)

- Hence for the expected costs of  $\mathcal{ALG}$ :

$$\mathbb{E}[\mathcal{ALG}(\sigma_i) | T_{v_i}] \geq \min(Dm, \frac{f}{2}) + \mathbb{E}[\mathcal{ALG}(\sigma_{i-1}) | T_{v_{i-1}}]$$

- We go through  $h+1$  phases:  $0 \leq i \leq h-1$  + last phase  $h$

$$\mathbb{E}[\mathcal{ALG}(\sigma)] \geq h \min(Dm, \frac{f}{2}) + \min(Dm, f)$$

### Competitive Ratio

- We choose  $m = h$  and  $D = \frac{f}{h}$ :

$$\mathbb{E}[\mathcal{ALG}(\sigma)] \geq h \min\left(\frac{f}{h}h, \frac{f}{2}\right) + \min\left(\frac{f}{h}h, f\right) = h\frac{f}{2} + f = hD\frac{h+2}{2}$$

$$OPT(\sigma) \stackrel{12}{\leq} f + hD\frac{m}{m-1} = hD + hD\frac{h}{h-1} = hD\frac{2h-1}{h-1}$$

- Since  $|\sigma| \leq m^h \frac{m}{m-1} \leq n \stackrel{m=h}{\implies} \frac{h^{h+1}}{h-1} \leq n$ : Choose  $h = \left\lfloor \frac{\log n}{\log \log n} \right\rfloor$  to obtain competitive ratio of  $\Omega\left(\frac{\log n}{\log \log n}\right)$ .

## Lower Bound for OMCFL: Overview

### Theorem (Online Multi-commodity Facility Location Problem)

No randomized algorithm for the Online Multi-commodity Facility Location Problem can achieve a **competitive ratio** better than  $\Omega(\sqrt{|S|})$ , even on a single point.

### Corollary (Online Multi-commodity Facility Location Problem)

No randomized algorithm for the Online Multi-commodity Facility Location Problem can achieve a **competitive ratio** better than  $\Omega(\sqrt{|S|} + \frac{\log n}{\log \log n})$ , even on a line metric.

- No proof here
- Corollary follows from combining theorem with OFLP lower bound

What more is there?

---

# Remarks on the Lower Bounds

## Lower bounds matched?

- Fotakis [2] gave  $\mathcal{O}(\frac{\log n}{\log \log n})$ -competitive deterministic algorithm (OFLP)
- Meyerson [5] gave  $\mathcal{O}(\frac{\log n}{\log \log n})$ -competitive randomized algorithm (OFLP)
- $\mathcal{O}(\sqrt{|S|} * \log n)$ -competitive deterministic algorithm [1] (OMCFLP)
- $\mathcal{O}(\sqrt{|S|} * \frac{\log n}{\log \log n})$ -competitive randomized algorithm [1] (OMCFLP)

## Concluding

---

# Conclusion

---

(Metric Spaces, Uncapacitated Problems)

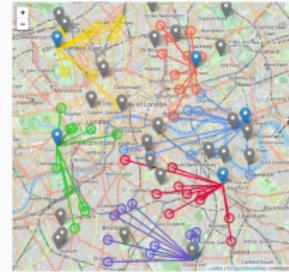
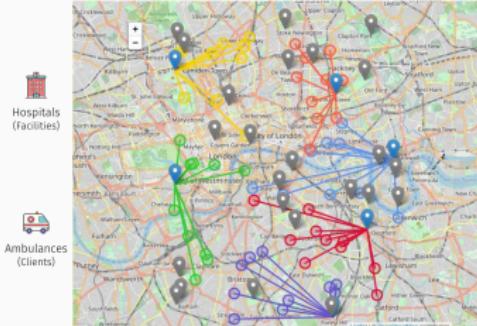
## OFLP

- $\Theta(\frac{\log n}{\log \log n})$ -competitive lower bound (Theorem 1)
- Matched by both deterministic and randomized algorithm

## OMCFLP

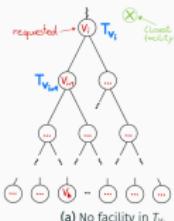
- $\Omega(\sqrt{|S|} + \frac{\log n}{\log \log n})$  lower bound on CR (Theorem 2)
- $\mathcal{O}(\sqrt{|S|} * \log n)$ -competitive deterministic algorithm
- $\mathcal{O}(\sqrt{|S|} * \frac{\log n}{\log \log n})$ -competitive randomized algorithm

# Your Questions

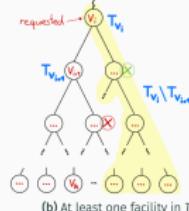


- Metric space  $\mathcal{M} = (M, d)$   
 $d : M \times M \rightarrow \mathbb{R}^+$
- Requests & Facilities  
 $r \in R, f \in F$ , anywhere in Request sequence  $\sigma$
- Construction costs  
 $f_m$ : Open  $f$  at  $m \in M$
- Provisioning costs  
 $d(r, f)$ : Serve  $r$  with  $f$

Serve  $\sigma = (r_1, r_2, \dots)$  and construct facilities  $F$  such that  
 $\mathcal{A}(\sigma) := \sum_{f_m \in F} f_m + \sum_{r \in \sigma} d(r, f)$  is minimized.



(a) exp. costs  $\geq \frac{p}{m^{j-1}} m^j + 0 = Dm$



(b) exp. costs  $\geq 0 + \frac{1}{2}f + \frac{1}{2}0 = \frac{f}{2}$

## OFLP

- $\Theta(\frac{\log n}{\log \log n})$ -competitive lower bound (Theorem 1)
- Matched by both deterministic and randomized algorithm

## OMCFLP

- $\Omega(\sqrt{|S|} + \frac{\log n}{\log \log n})$  lower bound on CR (Theorem 2)
- $\mathcal{O}(\sqrt{|S|} * \log n)$ -competitive deterministic algorithm
- $\mathcal{O}(\sqrt{|S|} * \frac{\log n}{\log \log n})$ -competitive randomized algorithm

## References i

---

-  J. Castenow, B. Feldkord, T. Knollmann, M. Malatyali, and F. Meyer  
Auf der Heide.  
**The online multi-commodity facility location problem.**  
In *Proceedings of the 32nd ACM Symposium on Parallelism in Algorithms and Architectures*, pages 129–139, 2020.
-  D. Fotakis.  
**On the competitive ratio for online facility location.**  
*Algorithmica*, 50(1):1–57, 2008.
-  S. Guha and S. Khuller.  
**Greedy strikes back: Improved facility location algorithms.**  
*Journal of algorithms*, 31(1):228–248, 1999.

## References ii

---

-  S. Li.  
A 1.488 approximation algorithm for the uncapacitated facility location problem.  
In *International Colloquium on Automata, Languages, and Programming*, pages 77–88. Springer, 2011.
-  A. Meyerson.  
**Online facility location.**  
In *Proceedings 42nd IEEE Symposium on Foundations of Computer Science*, pages 426–431. IEEE, 2001.
-  R. Ravi and A. Sinha.  
**Approximation algorithms for multicommodity facility location problems.**  
*SIAM Journal on Discrete Mathematics*, 24(2):538–551, 2010.

## References iii

-  D. B. Shmoys, C. Swamy, and R. Levi.  
**Facility location with service installation costs.**  
In *Proceedings of the fifteenth annual ACM-SIAM symposium on Discrete algorithms*, pages 1088–1097, 2004.
-  A. C.-C. Yao.  
**Probabilistic computations: Toward a unified measure of complexity.**  
In *18th Annual Symposium on Foundations of Computer Science (sfcs 1977)*, pages 222–227. IEEE Computer Society, 1977.

## Back-Up Slides

---

## Back up - Yao's Principle

- $\Sigma$ : set of all input sequences
- $\mathbb{Q}$ : distribution over input sequences
- $\mathcal{R}$ : any randomized algorithm
- $\mathcal{D}$ : set of all deterministic algorithms

### Yao's Principle [8]

Lower bound on the (worst-case) expected cost of any randomized algorithm  $\mathcal{R}$ .

$$\max_{\sigma \in \Sigma} \mathbb{E}[\mathcal{R}(\sigma)] \geq \min_{d \in \mathcal{D}} \mathbb{E}[d(\mathbb{Q})]$$