# Approximation Algorithms for the Fault-Tolerant Facility Placement Problem

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#### Outline

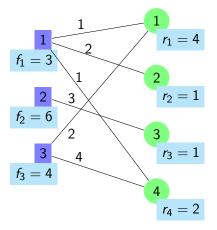
- The FTFP Problem
- Results in Dissertation
- Related Work
- 4 Techniques
- 5 Approximation Algorithms
- **6** Summary



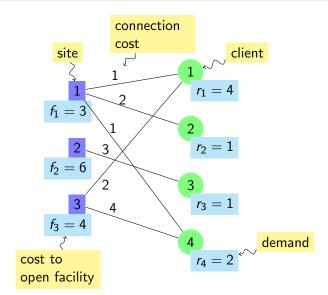
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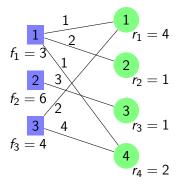
### Fault-Tolerant Facility Placement Problem (FTFP)



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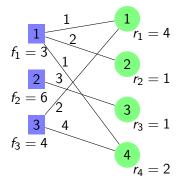


### Feasible Integral Solution

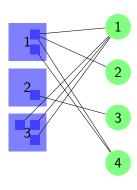


Instance

### Feasible Integral Solution

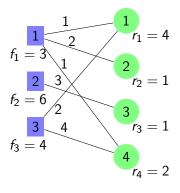


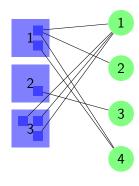
Instance



Solution

### Feasible Integral Solution





Instance

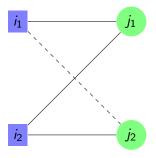
Solution

#### Cost

$$2f_1 + f_2 + 3f_3 + d_{11} + d_{12} + 2d_{14} + d_{23} + 3d_{31} = 38$$



### Metric Distances: Triangle Inequality



$$d(i_1, j_2) \le d(i_1, j_1) + d(i_2, j_1) + d(i_2, j_2)$$

Needed when estimating distances...

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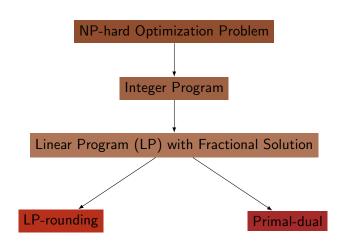
#### Hardness

## How hard is FTFP?

FTFP is NP-hard

FTFP is MaxSNP-hard

Best ratio  $\geq 1.463$  unless P = NP



### Results Highlight

- LP-rounding: 1.575-approximation
- LP-rounding: asymptotic ratio of 1 when all demands large
- Primal-dual:  $H_n$ -approximation
- Primal-dual: Example of  $\Omega(\log n / \log \log n)$  for dual-fitting

 $\begin{array}{ll} \mathsf{FTFP} & r_j \geq 1 & <\infty \text{ facility per site} \\ \mathsf{UFL} & r_j = 1 & \leq 1 \text{ facility per site} \\ \mathsf{FTFL} & r_j \geq 1 & \leq 1 \text{ facility per site} \end{array}$ 

$$\begin{array}{ll} \mathsf{FTFP} & r_j \geq 1 & <\infty \text{ facility per site} \\ \mathsf{UFL} & r_j = 1 & \leq 1 \text{ facility per site} \\ \mathsf{FTFL} & r_j \geq 1 & \leq 1 \text{ facility per site} \end{array}$$

$$\mathsf{UFL} \preceq \mathsf{FTFP} \preceq \mathsf{FTFL}$$

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$$\mathsf{UFL} \preceq \mathsf{FTFP} \preceq \mathsf{FTFL}$$

```
UFL 1.575
FTFL 1.7245
```

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$$\mathsf{UFL} \preceq \mathsf{FTFP} \preceq \mathsf{FTFL}$$

```
UFL 1.575
FTFP 1.7245
```

```
Primal-dual
UFL 1.52
FTFP O(log n)
```

- Related Work
- 6 Approximation Algorithms



#### Related Work for UFL

### Approximation Results for UFL

1997	3.16	LP-rounding
1998	1.736	LP-rounding
2002	1.58	LP-rounding
2001	3	primal-dual
2002	1.61	greedy
2002	1.52	greedy
2004	3	local search
2007	1.5	hybrid
2011	1.488	hybrid
	1998 2002 2001 2002 2002 2004 2007	1998     1.736       2002     1.58       2001     3       2002     1.61       2002     1.52       2004     3       2007     1.5

#### Lower Bound

Guha and Khuller 1998 1.463



#### Related Work for FTFL

### Approximation Algorithms for FTFL

Jain and Vazirani	2000	3 In max <sub>j</sub> r <sub>j</sub>	primal-dual
Guha <i>et al.</i>	2001	4	LP-rounding
Swamy, Shmoys	2008	2.076	LP-rounding
Byrka <i>et al.</i>	2010	1.7245	LP-rounding

No primal-dual algorithms for FTFL with constant ratio.



### Work on FTFP (Dissertation Topic)

### Approximation Algorithms for FTFP

Xu and Shen	2009		Introduced FTFP
Liao and Shen	2011	1.861	Dual-fitting (for special case)
Yan and Chrobak	2011	3.16	LP-rounding
Yan and Chrobak	2012	1.575	LP-rounding
Yan and Chrobak	preliminary results		Dual-fitting (for general case)

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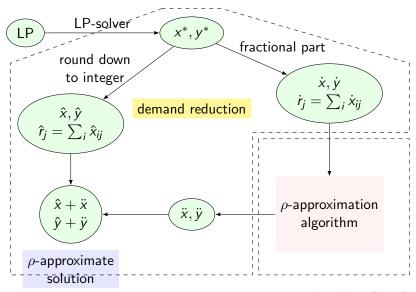
- Techniques

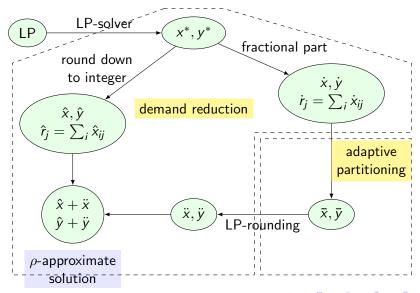
### Algorithm for FTFP — LP

- $v_i$  = number of facilities open at site  $i \in F$
- $x_{ii}$  = number of connections from client  $j \in \mathbb{C}$  to site  $i \in F$

(Primal) minimize 
$$\sum f_i y_i + \sum d_{ij} x_{ij}$$
  
subject to  $y_i - x_{ij} \ge 0$   $\forall i, j$   
 $\sum x_{ij} \ge r_j$   $\forall j$   
 $x_{ij} \ge 0, y_i \ge 0$   $\forall i, j$ 

(Dual) maximize 
$$\sum r_j \alpha_j$$
  
subject to  $\sum \beta_{ij} \leq f_i \quad \forall i$   
 $\alpha_j - \beta_{ij} \leq d_{ij} \quad \forall i, j$   
 $\alpha_j \geq 0, \beta_{ij} \geq 0 \quad \forall i, j$ 





#### **Techniques**

#### Demand Reduction

- Reduce all  $r_i$  to polynomial values (to ensure polynomial time of rounding)
- $\rho$ -approx for reduced instance  $\Rightarrow \rho$ -approx for original instance
- Adaptive Partitioning
  - Split sites into facilities and clients into unit demands
  - Split associated fractional values
  - Properties ensure rounding similar to UFL can be applied



#### **Demand Reduction**

#### **Implementation**

- Solving LP for (x\*, y\*)
- $(\hat{\mathbf{x}}, \hat{\mathbf{y}}) = (\mathbf{x}^*, \mathbf{y}^*)$  round down to integer
- $\bullet$   $(\dot{\mathbf{x}},\dot{\mathbf{y}})=(\mathbf{x}^*,\mathbf{y}^*)-(\hat{\mathbf{x}},\hat{\mathbf{y}})$ , fractional part
- $\hat{r}_j = \sum_j \hat{x}_{ij}$  for  $\hat{\mathcal{I}}$ ,  $\dot{r}_j = r_j \hat{r}_j$  for  $\dot{\mathcal{I}}$
- ullet  $(\hat{\mathbf{x}},\hat{\mathbf{y}})$  (integral) feasible and optimal for  $\hat{\mathcal{I}}$
- ullet  $(\dot{x},\dot{y})$  (fractional) feasible and optimal for  $\dot{\mathcal{I}}$

#### **Properties**

- $\dot{r}_i = \text{poly}(|F|)$
- $\rho$ -approx for  $\dot{\mathcal{I}}$  implies  $\rho$ -approx for  $\mathcal{I}$



### Demand Reduction: Consequences

#### FTFP to FTFL, 1.7245-approximation

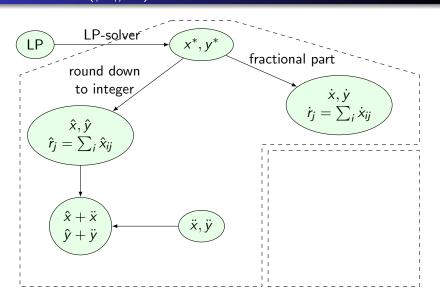
- Sites into facilities
- Clients with demand r<sub>i</sub>
- FTFL size polynomial because of demand reduction

Ratio 
$$1 + O(|F|/Q)$$
 for  $Q = \min_j r_j$ , approaches 1 when  $Q$  is large

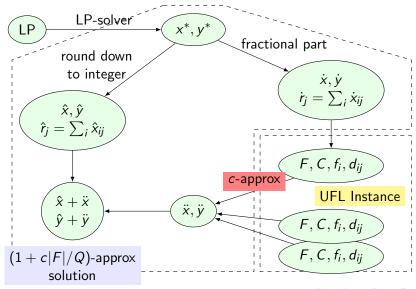
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### Ratio 1 + O(|F|/Q) for FTFP



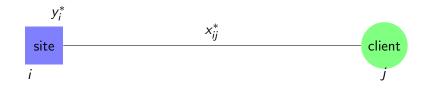
### Ratio 1 + O(|F|/Q) for FTFP

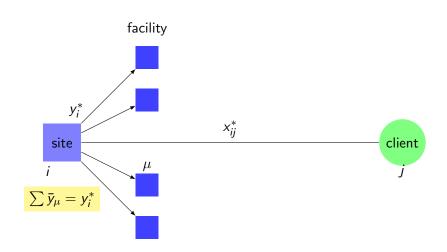


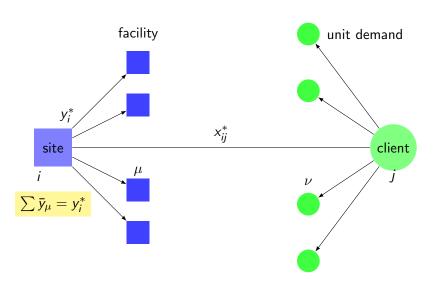
### **Techniques**

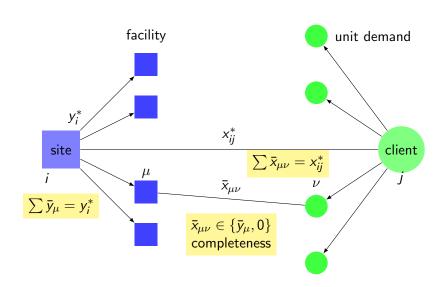
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  - $\rho$ -approx for reduced instance  $\Rightarrow \rho$ -approx for original instance
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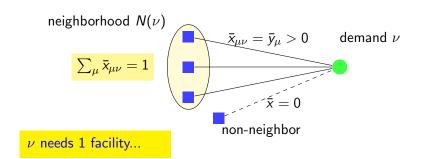








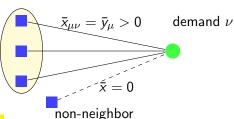
### Neighborhood of Demand



# Neighborhood of Demand

neighborhood  $N(\nu)$ 





 $\nu$  needs 1 facility...

Strategy 1: for each  $\nu$ , open one  $\mu \in N(\nu)$  with prob.  $\bar{y}_{\mu}$ 

- optimal connection cost
- large facility cost

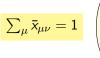
Strategy 2: open facility only for demands with disjoint neighborhoods

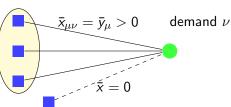
- optimal facility cost
- large connection cost



## Neighborhood of Demand

neighborhood  $N(\nu)$ 





non-neighbor

 $\nu$  needs 1 facility...

Strategy 1: for each  $\nu$ , open one  $\mu \in N(\nu)$  with prob.  $\bar{y}_{\mu}$ 

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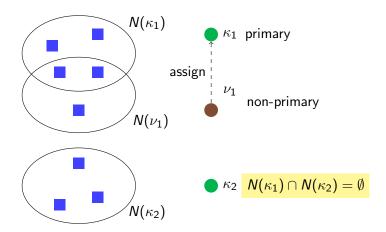
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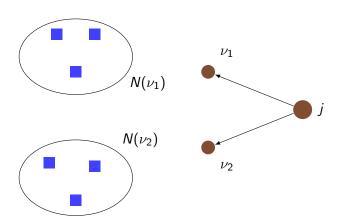
How to balance these two costs?



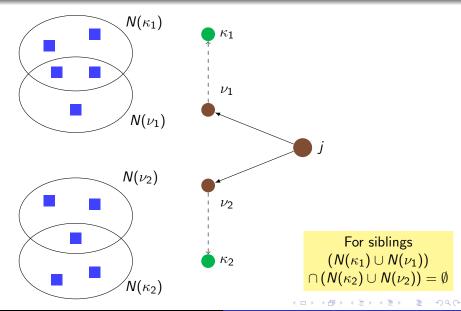
# Two Types of Demands: Primary and Non-primary



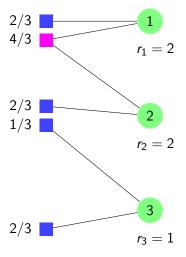
# Neighborhood Structure for Siblings



# Neighborhood Structure for Siblings



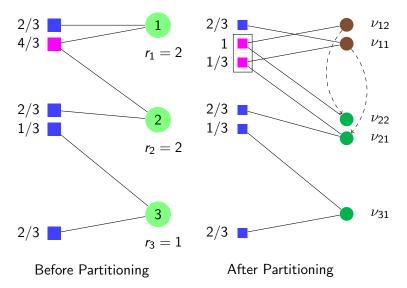
# **Example of Partitioning**



Before Partitioning



## Example of Partitioning





### Partitioning:

- Clients → demands
- Sites → facilities
- $(x^*, y^*) \rightarrow (\bar{x}, \bar{y})$
- $\bullet \sum_{\mu} \bar{x}_{\mu\nu} = 1$
- $\bar{x}_{\mu\nu} = \bar{y}_{\mu}$  or 0

#### Structure:

• If  $\kappa_1, \kappa_2$  primary then  $N(\kappa_1) \cap N(\kappa_2) = \emptyset$ 

- Each non-primary  $\nu$  assigned to  $\kappa$  with
  - $N(\kappa) \cap N(\nu) \neq \emptyset$
  - priority( $\kappa$ ) < priority ( $\nu$ )

 $\bullet$   $(N(\kappa_1) \cup N(\nu_1)) \cap$  $(N(\kappa_2) \cup N(\nu_2)) = \emptyset$ 



### Partitioning:

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#### small facility cost

- Each non-primary  $\nu$  assigned to  $\kappa$  with
  - $N(\kappa) \cap N(\nu) \neq \emptyset$
  - priority( $\kappa$ )  $\leq$  priority ( $\nu$ )

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  - $N(\kappa) \cap N(\nu) \neq \emptyset$
  - priority( $\kappa$ ) < priority ( $\nu$ )

### small connection cost of $\nu$

 $\bullet$   $(N(\kappa_1) \cup N(\nu_1)) \cap$  $(N(\kappa_2) \cup N(\nu_2)) = \emptyset$ 



### Partitioning:

- Clients → demands
- Sites → facilities
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- $\sum_{\mu} \bar{x}_{\mu\nu} = 1$
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- Each non-primary  $\nu$  assigned to  $\kappa$  with
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 $(N(\kappa_1) \cup N(\nu_1)) \cap (N(\kappa_2) \cup N(\nu_2)) = \emptyset$ 

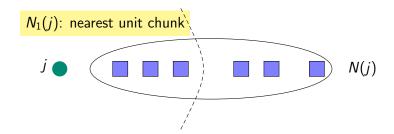
#### fault-tolerance

## Partitioning Implementation

## Partitioning implementation: two phases

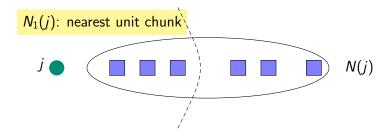
- Phase 1, the partitioning phase
  - Define demands
  - Allocate facilities
- Phase 2, the augmenting phase
  - Add facilities to make neighborhood unit

## In each iteration, create one demand for best client



## Phase 1, Step 1

## In each iteration, create one demand for best client

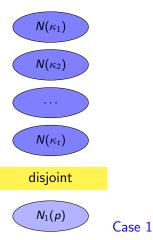


- $\mathsf{bid}(j) = \mathsf{avgdist}(N_1(j)) + \alpha_i^*(\mathsf{dual\ value})$
- Best bid client p selected to create a demand



# Phase 1, Step 2

Best client p creates demand  $\nu$ , to decide  $N(\nu)$ , two cases:



 $N(\kappa)$   $N_1(p)$ 

 $N_1(p)$  overlaps some  $N(\kappa)$ 

Case 2

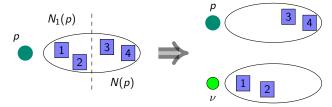
## Phase 1, Step 2 Contd.

Best client p creates demand  $\nu$ , to decide  $N(\nu)$ , two cases:

# Phase 1, Step 2 Contd.

Best client p creates demand  $\nu$ , to decide  $N(\nu)$ , two cases:

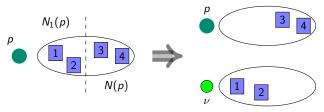
• Case 1: disjoint,  $N(\nu)$  gets  $N_1(p)$ 



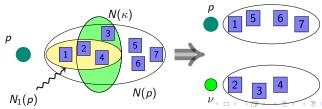
# Phase 1, Step 2 Contd.

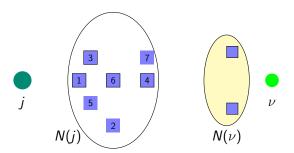
### Best client p creates demand $\nu$ , to decide $N(\nu)$ , two cases:

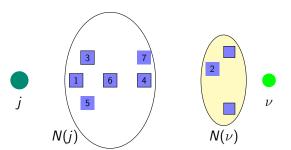
• Case 1: disjoint,  $N(\nu)$  gets  $N_1(p)$ 

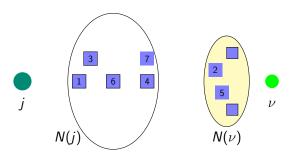


• Case 2: overlap,  $N(\nu)$  gets  $N(p) \cap N(\kappa)$ 

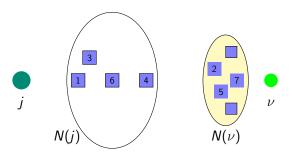


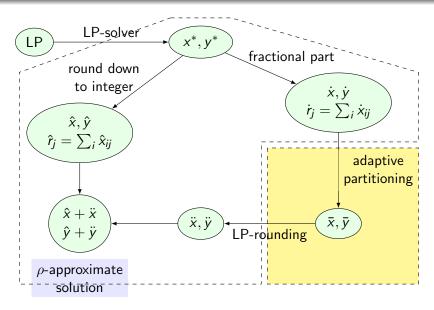






## Phase 2





Done with partitioning, next to rounding



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# 3-Approximation for FTFP

## Rounding: round each $\bar{y}_{\mu}$ and $\bar{x}_{\mu\nu}$ to 0 or 1

- Facilities: each primary  $\kappa$  opens one  $\mu \in N(\kappa)$
- Connections: non-primary demands  $\nu$  assigned to  $\kappa$ connect to  $\mu$

### **Analysis**

- Fault-Tolerance:  $\nu$  uses only facilities in  $N(\nu) \cup N(\kappa)$
- Cost:  $< 3 \cdot LP^*$ , because
  - Facility cost < F\*</li>
  - Connection cost  $< C^* + 2 \cdot LP^*$

## 1.736-Approximation for FTFP

## Rounding: round each $\bar{y}_{\mu}$ and $\bar{x}_{\mu\nu}$ to 0 or 1

- Facilities:
  - Each primary  $\kappa$  opens random  $\mu \in N(\kappa)$
  - Other facilities open randomly independently
- Connections:
  - If a neighbor open, connect to nearest neighbor
  - Else connect via assigned primary demand

### **Analysis**

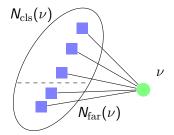
- Fault-Tolerance:  $\nu$  uses only facilities in  $N(\nu) \cup N(\kappa)$
- Cost:  $< (1+2/e) LP^*$ , because
  - Facility cost < F\*</li>
  - Connection cost  $\leq C^* + (2/e) \cdot LP^*$



# 1.575-Approximation for FTFP — Partitioning

### More intricate neighborhood structure

- Two neighborhoods: close and far,  $N(\nu) = N_{\rm cls}(\nu) \cup N_{\rm far}(\nu)$
- $N_{\rm cls}(\nu) = \text{nearest } (1/\gamma) \text{fraction of } N(\nu)$
- $N_{\rm cls}(\nu) \cap N_{\rm cls}(\kappa) \neq \emptyset$ , if  $\nu$  assigned to  $\kappa$
- For siblings  $\nu_1, \nu_2, N_{\rm cls}(\kappa_1) \cup N(\nu_1)$  and  $N_{\rm cls}(\kappa_2) \cup N(\nu_2)$ disjoint



## 1.575-Approximation for FTFP — Rounding

Rounding: boost  $(\mathbf{x}^*, \mathbf{y}^*)$  by  $\gamma$  and apply demand reduction and adaptive partitioning, then round by

- Facilities:
  - Each primary  $\kappa$  opens random  $\mu \in N_{cls}(\kappa)$
  - Other facilities open randomly independently
- Connections:
  - If a neighbor open, connect to nearest neighbor
  - Else connect via assigned primary demand

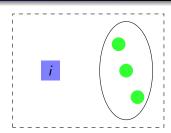
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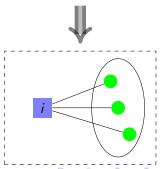
- Fault-Tolerance:  $\nu$  uses only facilities in  $N(\nu) \cup N_{\rm cls}(\kappa)$
- Cost:  $\langle \gamma \cdot LP \text{ for } \gamma = 1.575, \text{ because}$ 
  - Facility cost  $< \gamma \cdot F^*$
  - Connection cost  $< \gamma \cdot C^*$



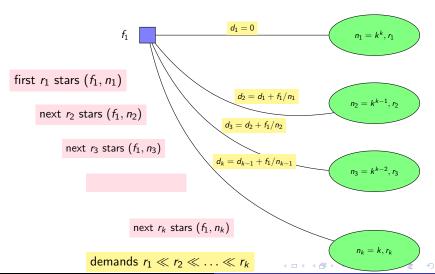
# Greedy and Dual-fitting

- Greedy in polynomial time
  - Best star can be found quickly
  - Best star remains best
- Ratio  $H_n$  (Wolsey's result): Greedy is  $H_n$ -approx for
  - Minimizing a linear function
  - Subject to submodular constraints
- Lower bound  $\Omega(\log n / \log \log n)$  for dual-fitting
  - Example has k groups,  $n = k^k$
  - Shrinking factor is k/2





Dual feasibility forces a ratio of k/2, number of clients  $n = k^k$ 



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## Summary

#### Results

- 1.575-approximation algorithm for FTFP
- Technique for extending LP-rounding algorithms for UFL to FTFP

# Summary

#### Results

- 1.575-approximation algorithm for FTFP
- Technique for extending LP-rounding algorithms for UFL to FTFP

### Open Problems

- Can FTFL be approximated with the same ratio?
- LP-free algorithms for FTFP or FTFL with constant ratio?
- Close the 1.463 − 1.488 gap for UFL!