

# Approximation Algorithms for the Fault-Tolerant Facility Placement Problem

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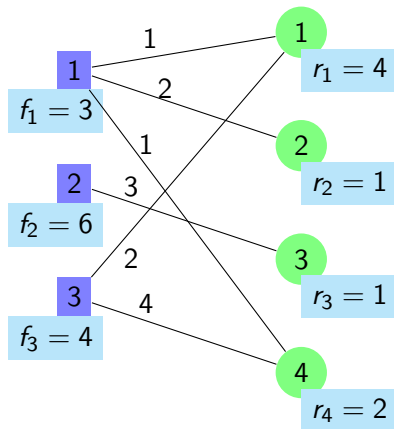
06/10/2013

- 1 The FTFP Problem
- 2 Related Work
- 3 Our Results
- 4 Techniques and Algorithms
- 5 Summary

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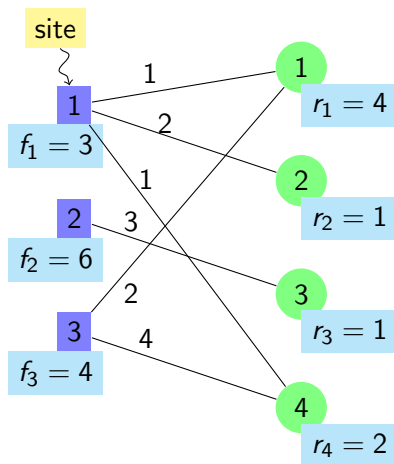
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# Fault-Tolerant Facility Placement Problem (FTFP)



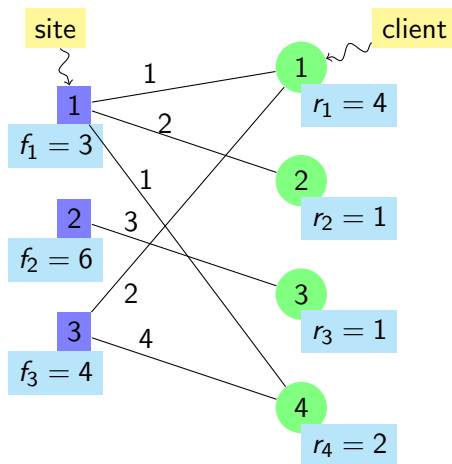
Instance

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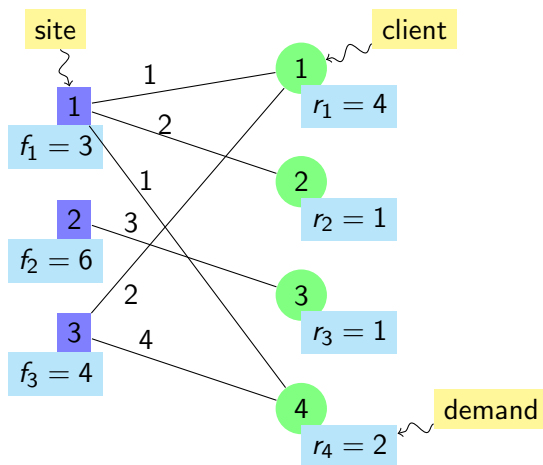
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# Fault-Tolerant Facility Placement Problem (FTFP)



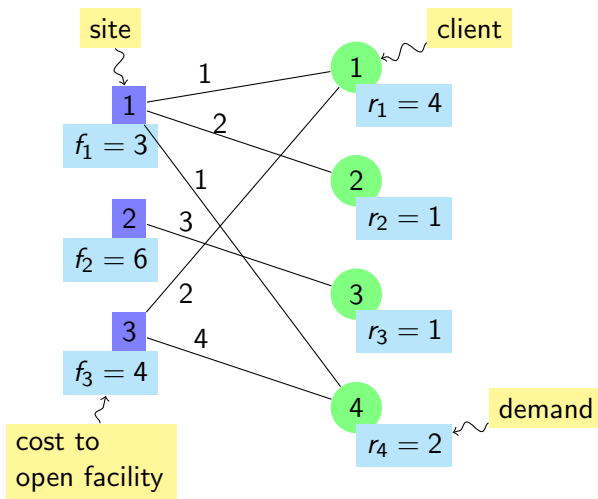
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# Fault-Tolerant Facility Placement Problem (FTFP)



Instance

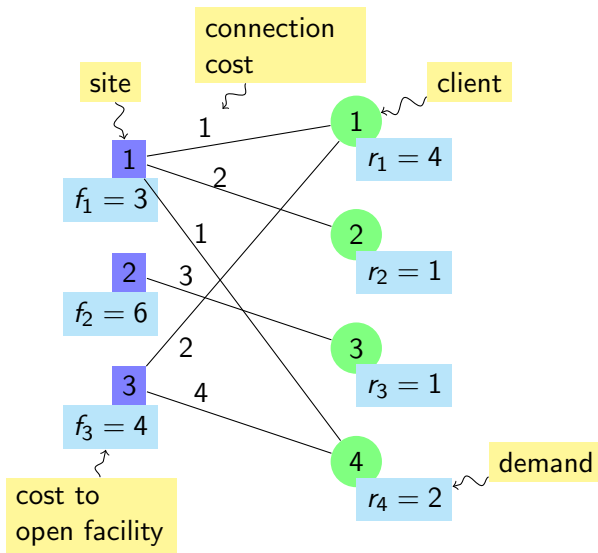
# Fault-Tolerant Facility Placement Problem (FTFP)



Instance

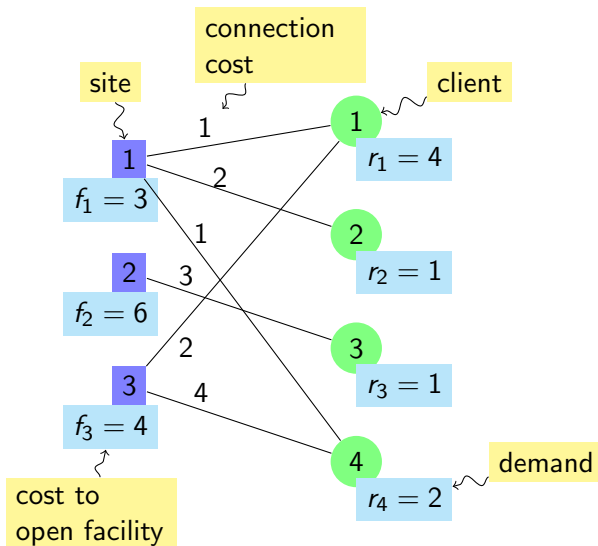


# Fault-Tolerant Facility Placement Problem (FTFP)



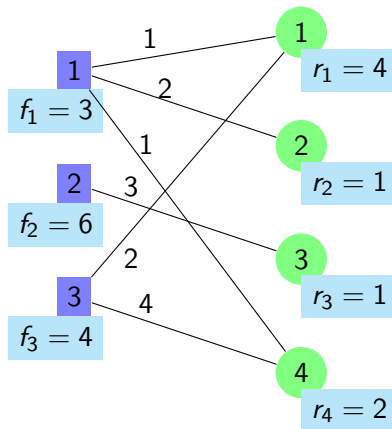
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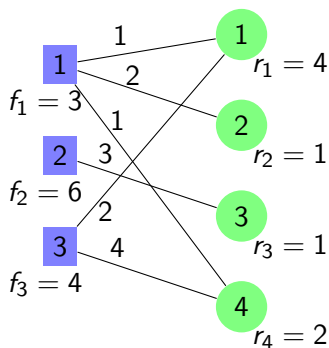
Instance

# Fault-Tolerant Facility Placement Problem (FTFP)



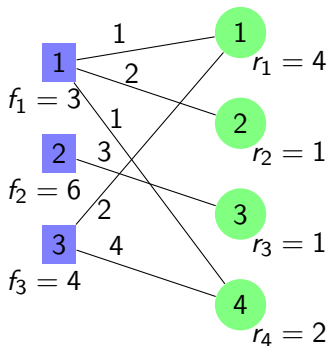
Instance

# Feasible Integral Solution

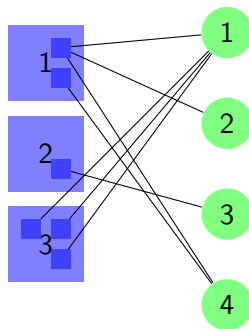


Instance

# Feasible Integral Solution

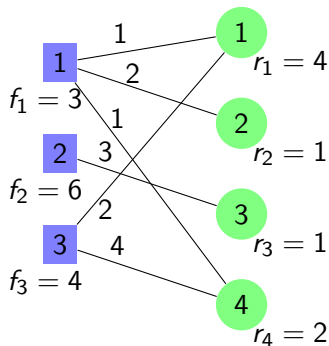


Instance

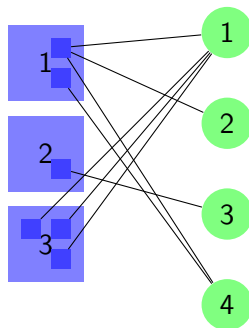


Solution

# Feasible Integral Solution



Instance

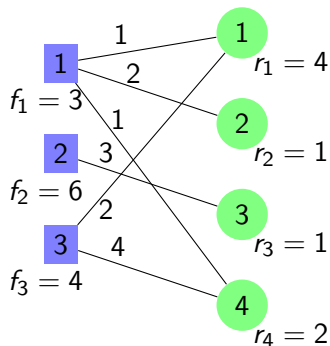


Solution

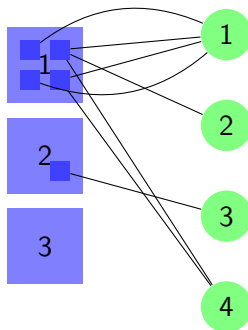
## Cost

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

# Optimal Integral Solution

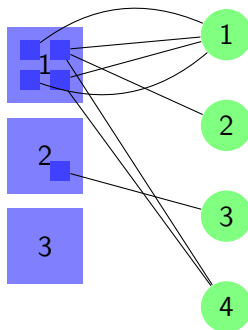
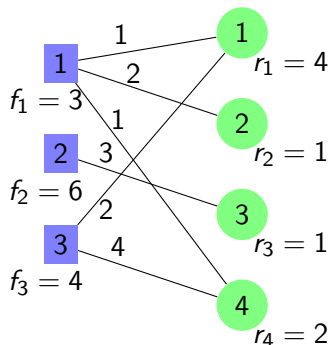


Instance



Solution

# Optimal Integral Solution



Cost

$$4f_1 + 1f_2 + 0f_3 + 4d_{11} + d_{12} + 2d_{14} + d_{23} = 29$$



# Relation between Problems

FTFP	$r_j \geq 1$	$\geq 1$ facility per site
UFL	$r_j = 1$	$\leq 1$ facility per site
FTFL	$r_j \geq 1$	$\leq 1$ facility per site

# Relation between Problems

FTFP	$r_j \geq 1$	$\geq 1$ facility per site
UFL	$r_j = 1$	$\leq 1$ facility per site
FTFL	$r_j \geq 1$	$\leq 1$ facility per site

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

# Relation between Problems

FTFP	$r_j \geq 1$	$\geq 1$ facility per site
UFL	$r_j = 1$	$\leq 1$ facility per site
FTFL	$r_j \geq 1$	$\leq 1$ facility per site

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

LP-rounding

UFL	1.575
FTFP	
FTFL	1.7245

# Relation between Problems

FTFP	$r_j \geq 1$	$\geq 1$ facility per site
UFL	$r_j = 1$	$\leq 1$ facility per site
FTFL	$r_j \geq 1$	$\leq 1$ facility per site

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

LP-rounding

UFL	1.575
FTFP	
FTFL	1.7245

Primal-dual

UFL	1.52
FTFP	
FTFL	$O(\log n)$

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# Uncapacitated Facility Location Problem (UFL)

All demands are 1, each site can open only one facility

1

1

$$r_1 = 1$$

2

2

$$r_2 = 1$$

3

3

$$r_3 = 1$$

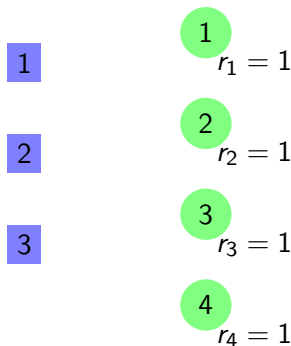
4

$$r_4 = 1$$

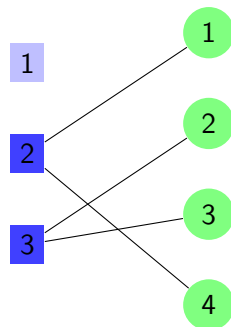
Instance

# Uncapacitated Facility Location Problem (UFL)

All demands are 1, each site can open only one facility



Instance



Solution

# Related Work for UFL

## Approximation Results for UFL

Shmoys, Tardos and Aardal	1997	3.16	LP-rounding
Chudak	1998	1.736	LP-rounding
Sviridenko	2002	1.58	LP-rounding
Jain and Vazirani	2001	3	primal-dual
Jain <i>et al.</i>	2002	1.61	greedy
Mahdian <i>et al.</i>	2002	1.52	greedy
Arya <i>et al.</i>	2004	3	local search
Byrka	2007	1.5	hybrid
Li	2011	1.488	hybrid

## Lower Bound

Guha and Khuller	1998	1.463
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# Fault-Tolerant Facility Location Problem (FTFL)

Demands may be more than 1, each site can open only one facility

1

1

$$r_1 = 2$$

2

2

$$r_2 = 1$$

3

3

$$r_3 = 1$$

4

$$r_4 = 2$$

Instance

# Fault-Tolerant Facility Location Problem (FTFL)

Demands may be more than 1, each site can open only one facility

1

2

3

1

$$r_1 = 2$$

2

$$r_2 = 1$$

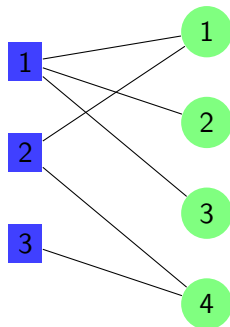
3

$$r_3 = 1$$

4

$$r_4 = 2$$

Instance



Solution

# Related Work for FTFL

## Approximation Algorithms for FTFL

Jain and Vazirani	2000	$3 \ln \max_j r_j$	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.

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# 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

This talk:

Yan and Chrobak	2012	1.575	LP-rounding
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### Highlights

- Matches the best LP-based ratio for UFL
- Better than 1.7245 for FTFL
- Technique to extend LP-rounding algorithms for UFL to FTFP

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# LP Formulation for FTFP

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

$$\begin{aligned} & \text{minimize} && \sum f_i y_i + \sum d_{ij} x_{ij} && (1) \\ & \text{subject to} && y_i - x_{ij} \geq 0 && \forall i, j \\ & && \sum x_{ij} \geq r_j && \forall j \\ & && x_{ij} \geq 0, y_i \geq 0 && \forall i, j \end{aligned}$$

$$\begin{aligned} (\text{Dual}) \quad & \text{maximize} && \sum r_j \alpha_j && (2) \\ & \text{subject to} && \sum \beta_{ij} \leq f_i && \forall i \\ & && \alpha_j - \beta_{ij} \leq d_{ij} && \forall i, j \\ & && \alpha_j \geq 0, \beta_{ij} \geq 0 && \forall i, j \end{aligned}$$

# Techniques

- Demand Reduction

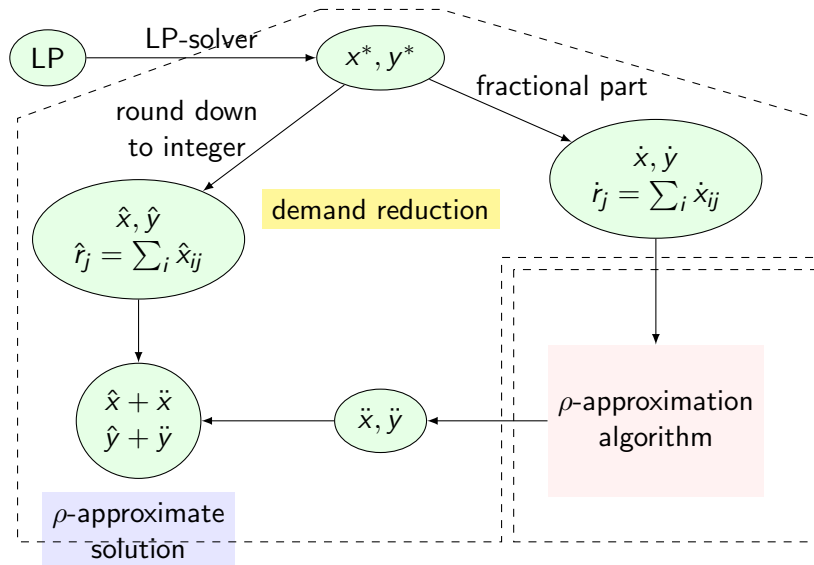
- Reduce all  $r_j$  to polynomial values (to ensure polynomial time of rounding)
- $\rho$ -approx for reduced instance  $\Rightarrow$   $\rho$ -approx for original instance

- Adaptive Partitioning

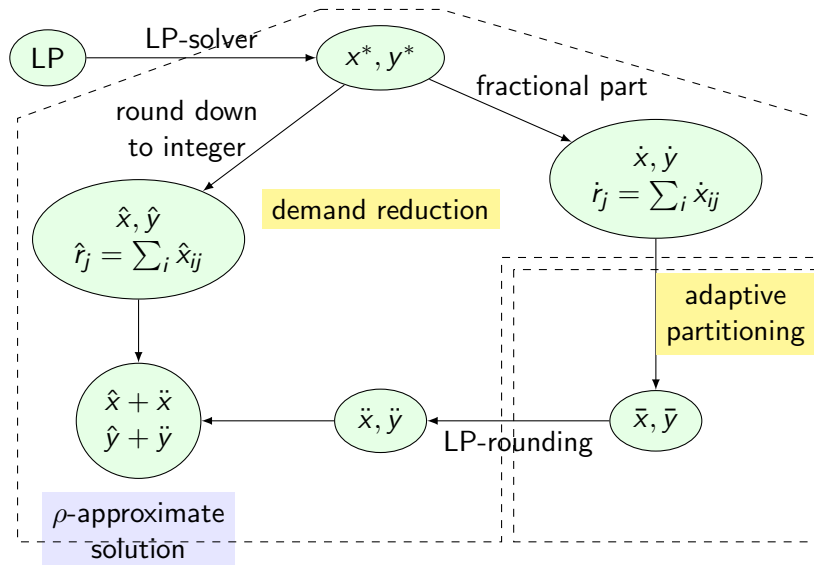
- Split sites into facilities and clients into unit demands
- Split associated fractional values
- Properties ensure rounding similar to UFL can be applied



# Algorithm for FTFP



# Algorithm for FTFP



# Demand Reduction

## Implementation

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

## Properties

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

# Demand Reduction: Consequences

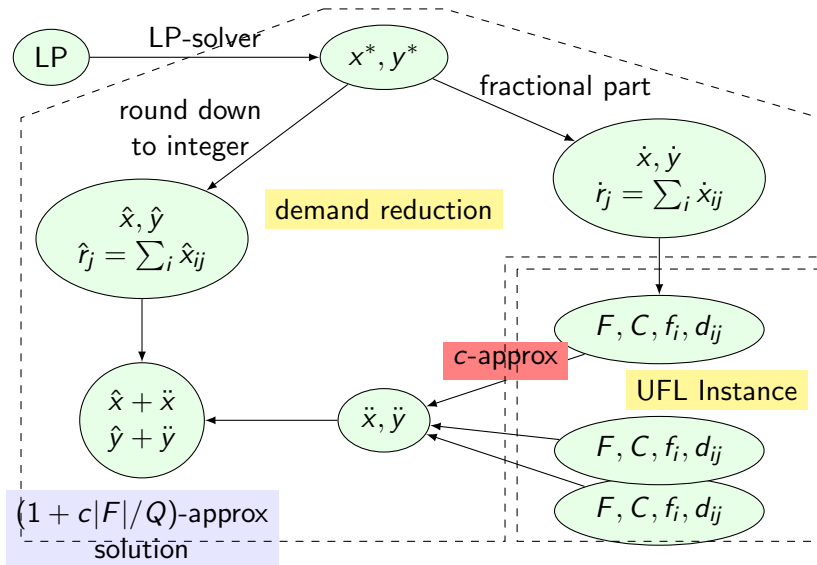
FTFP to FTFL, 1.7245-approximation

- sites into facilities
- clients with demand  $r_j$

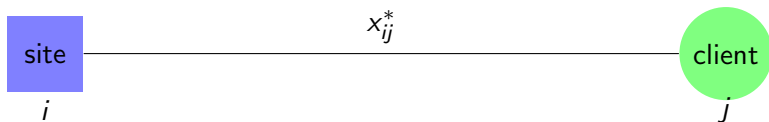
Ratio  $1 + O(|F|/Q)$  for  $Q = \min_j r_j$ , approaches 1 when  $Q$  is large

- next slide

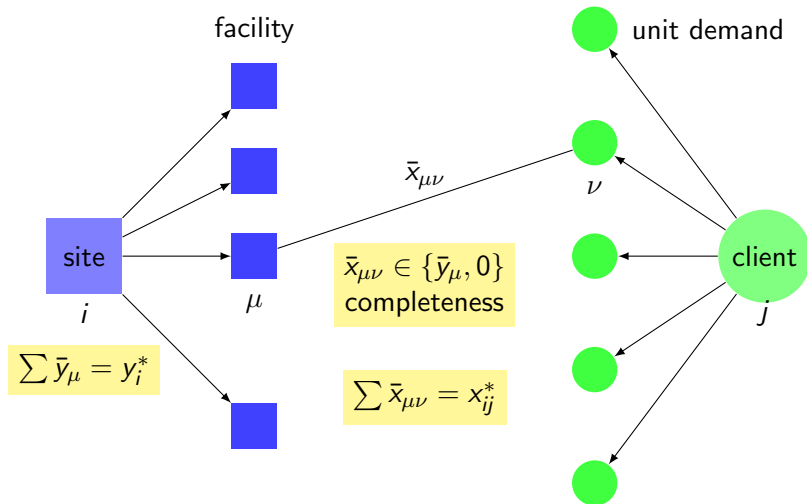
# Ratio $1 + O(|F|/Q)$ for FTFP



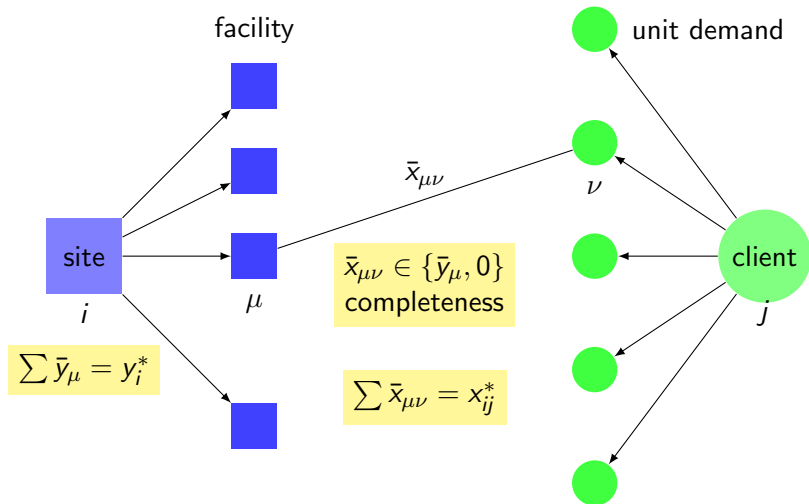
# Adaptive Partitioning



# Adaptive Partitioning



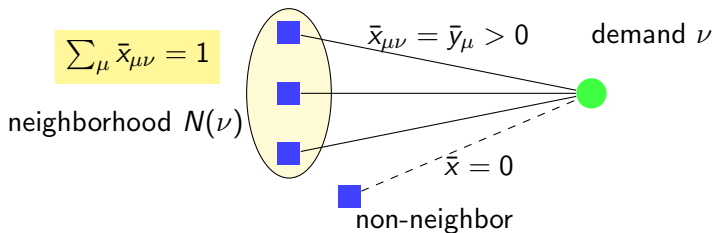
# Adaptive Partitioning



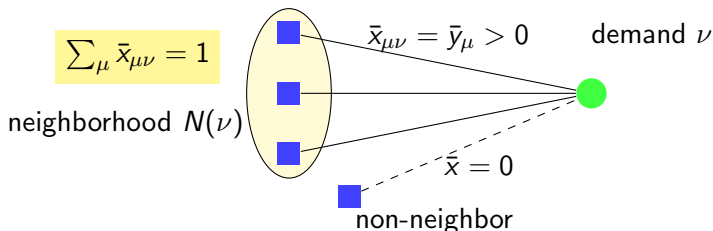
Partition must satisfy several properties  
needed for rounding to work...



# Neighborhood of a demand



# Neighborhood of a demand



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

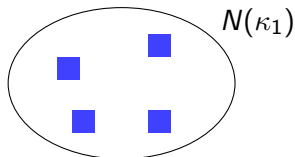
- optimal connection cost
- large facility cost

**Strategy 2:** do this for demands with disjoint neighborhoods

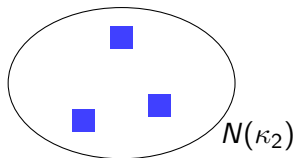
- optimal facility cost
- large connection cost

How to balance these strategies?

# Two Types of Demands



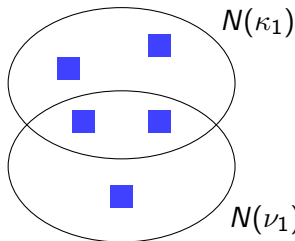
●  $\kappa_1$  primary



●  $\kappa_2$

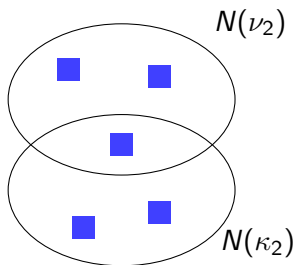
$$N(\kappa_1) \cap N(\kappa_2) = \emptyset$$

# Two Types of Demands



●  $\kappa_1$  primary

●  $\nu_1$  non-primary

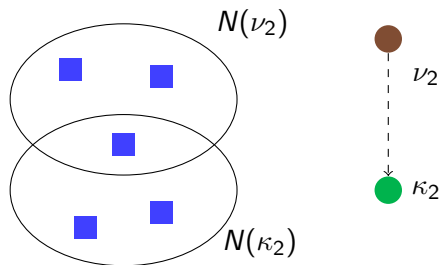
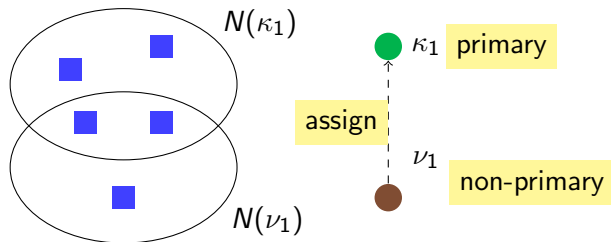


●  $\nu_2$

●  $\kappa_2$

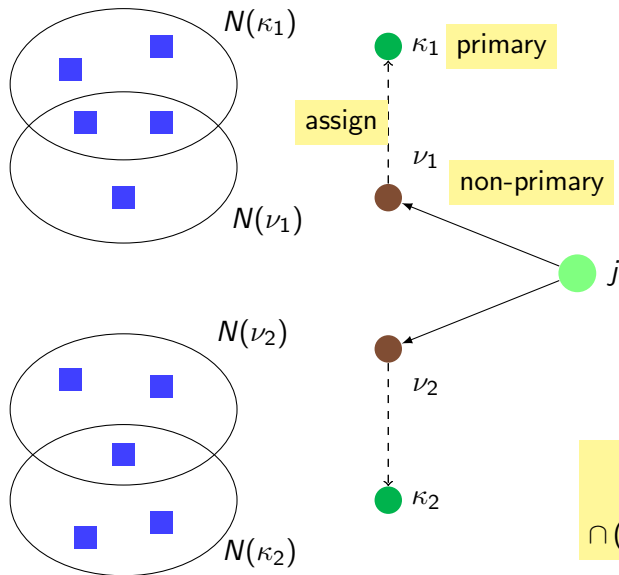
$$N(\kappa_1) \cap N(\kappa_2) = \emptyset$$

# Two Types of Demands



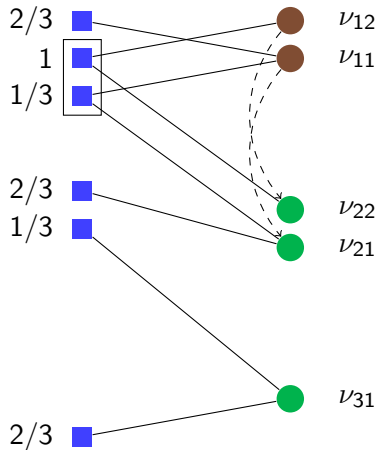
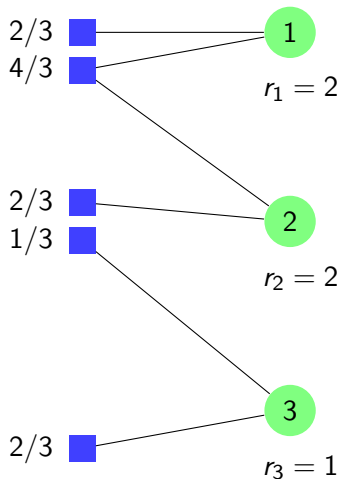
$$N(\kappa_1) \cap N(\kappa_2) = \emptyset$$

# Neighborhood Structure for Siblings



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

# Example of Partitioning



# Summary of Partitioning

## Partitioning:

- Clients  $\rightarrow$  demands
- Sites  $\rightarrow$  facilities  
(not yet opened)
- $(x^*, y^*) \rightarrow (\bar{x}, \bar{y})$
- $\sum_{\mu} \bar{x}_{\mu\nu} = 1$
- $\bar{x}_{\mu\nu} = \bar{y}_{\mu}$  or 0



# Summary of Partitioning

## Partitioning:

- Clients  $\rightarrow$  demands
- Sites  $\rightarrow$  facilities  
(not yet opened)
- $(x^*, y^*) \rightarrow (\bar{x}, \bar{y})$
- $\sum_{\mu} \bar{x}_{\mu\nu} = 1$
- $\bar{x}_{\mu\nu} = \bar{y}_{\mu}$  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
  - $N(\kappa) \cap N(\nu) \neq \emptyset$
  - $\text{priority}(\kappa) \leq \text{priority}(\nu)$   
(rough estimate of demand's cost)
- if  $\nu_1, \nu_2$  are siblings and  $\nu_i$  assigned to  $\kappa_i$ , then  
 $[N(\kappa_1) \cup N(\nu_1)] \cap [N(\kappa_2) \cup N(\nu_2)] = \emptyset$

# Summary of Partitioning - Intuition

## Structure:

small facility cost



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

small connection  
cost of  $\nu$



- Each non-primary  $\nu$  assigned to  $\kappa$  with
  - $N(\kappa) \cap N(\nu) \neq \emptyset$
  - $\text{priority}(\kappa) \leq \text{priority}(\nu)$   
(rough estimate of demand's cost)

fault tolerance



- if  $\nu_1, \nu_2$  are siblings and  $\nu_i$  assigned to  $\kappa_i$ , then  
 $[N(\kappa_1) \cup N(\nu_1)] \cap [N(\kappa_2) \cup N(\nu_2)] = \emptyset$

# 3-Approximation for FTFP

## Client priority values

- $\text{tcc}(j) + \alpha_j^*$   
(average connection cost + dual value)

## Rounding

- **Facilities:** Each primary  $\kappa$  opens random  $\mu \in N(\kappa)$
- **Connections:** All demands assigned to  $\kappa$  connect to  $\mu$

## Analysis

- **Fault-Tolerance:**  $\nu$  uses only facilities in  $N(\nu) \cup N(\kappa)$
- **Cost:**  $\leq 3 \cdot \text{LP}^*$ , because
  - Facility cost  $\leq F^*$
  - Connection cost  $\leq C^* + 2 \cdot \text{LP}^*$

# 1.736-Approximation for FTFP

## Client priority values

- $\text{tcc}(j) + \alpha_j^*$   
(average connection cost + dual value)

## Rounding

- **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

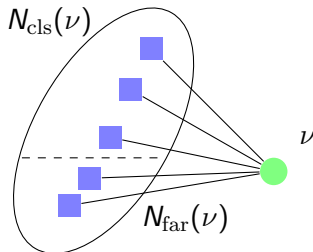
## Analysis

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

# 1.575-Approximation for FTFP – Idea

## More intricate neighborhood structure

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



# 1.575-Approximation for FTFP

## Client priority values

- $\text{tcc}_{\text{cls}}(j) + \text{dmax}_{\text{cls}}(j)$   
(average + worst connection cost to close neighborhood)

## Rounding (extension of Byrka's)

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

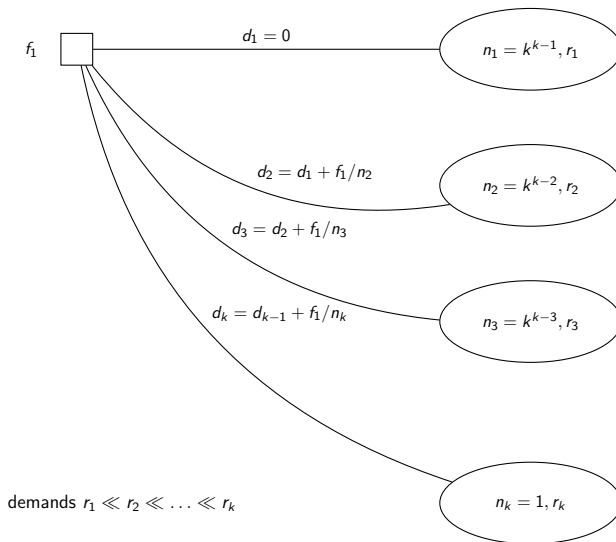
## Analysis

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

# 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 constraint
- Lower bound  $O(\log n / \log \log n)$  for dual-fitting
  - Example has  $k$  groups,  $n = k^k$
  - Shrinking factor is  $k/2$

# Dual-fitting Example





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# Q and A about FTFP

- Q: Is there a simple reduction from FTFP to UFL?
- Q: Which one is easier, FTFP or FTFL?
- Q: Can FTFP have a better ratio than FTFL?
- Q: When all  $r_j$  are large, do you get a ratio 1?
- Q: Does greedy have  $O(1)$  ratio or not?
- Q: What is the best possible ratio for FTFP?

# Q and A about FTFP

- Q: Is there a simple reduction from FTFP to UFL?
- A: Not sure, for the uniform demand case yes.
- Q: Which one is easier, FTFP or FTFL?
- Q: Can FTFP have a better ratio than FTFL?
- Q: When all  $r_j$  are large, do you get a ratio 1?
- Q: Does greedy have  $O(1)$  ratio or not?
- Q: What is the best possible ratio for FTFP?

# Q and A about FTFP

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- Q: Which one is easier, FTFP or FTFL?
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- 1.575-approximation algorithm for FTFP
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## 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!