

# A Lottery Model for Center-type Problems With Outliers

David G. Harris   Thomas Pensyl   Aravind Srinivasan  
**Khoa Trinh**

Department of Computer Science, University of Maryland, College Park, USA

APPROX 2017

# The $k$ -center problem

- Input:
  - set  $V$  of vertices (or *clients*)
  - symmetric distance metric  $d$  on  $V$
  - a parameter  $k \in \mathbb{Z}$
- Goal: pick a set  $\mathcal{S} \subseteq V$  of size  $\leq k$  which *minimizes* the maximum connection cost  $R$  of any vertex:

$$R := \max_{j \in V} \min_{i \in \mathcal{S}} d(i, j)$$

- Known-results: 2-approximable; best possible unless  $P=NP$  [Hsu & Nemhauser '79, Hochbaum & Shmoys '86]

# The matroid center problem

- The cardinality constraint  $|S| \leq k$  is replaced by a matroid constraint:  $S$  must be an independent set of some given matroid  $\mathcal{M} = (V, \mathcal{I})$ .
- Known-results: 3-approximable; best possible unless  $P=NP$  [CLLW '13]

# The robust model

- Introduced by [Charikar, Khuller, Mount, Narasimhan '01]
- Given  $t \leq n$ , only need to cover  $\geq t$  clients (i.e., we can ignore  $\leq n - t$  “outliers” from the instance)
- The robust  $k$ -center problem:
  - Upper-bound: 3 [CKMN '01], 2 [CGK '16]
  - Lower-bound: 2
- The robust matroid center problem:
  - Upper-bound: 7 [CLLW '13]
  - Lower-bound: 3 [CLLW '13]

# The lottery model

## Motivations:

- We can think of the robust  $k$ -center as a facility-location problem: “unfairness” arises naturally as some clients might always be considered as outliers,
- In many contexts (e.g., unsupervised machine learning), each vertex  $j$  is “autonomous” – we do not want individuals excluded consistently.

# The lottery model

**The Lottery Model** for robust center problems: each client  $j \in \mathcal{C}$  submits a target probability  $p_j$  and we want a minimum radius  $R$  such that  $\exists$  distribution  $\mathcal{D}$  on subsets of  $V$ : for any  $\mathcal{S} \sim \mathcal{D}$  (such an  $\mathcal{S}$  should be efficiently sampleable),

- Feasibility constraint:  $\mathcal{S}$  is feasible with prob. 1
- Coverage constraint:  $\#\text{covered clients} \geq t$  with prob. 1
- *Fairness constraint*:  $\Pr[j \text{ is covered}] \geq p_j$  for all  $j$

# Our main results

- A simple tight 3-approximation algorithm for the robust matroid center problem,
- Our results for the lottery model:

Problem	Radius	Feasibility	Coverage	Fairness
$k$ -center	$2R$	$\leq k$	$\geq (1 - \epsilon)t$	$\geq (1 - \epsilon)p_j$ for all clients
MatCenter	$3R$ $3R$	Basis + "1" Basis	$\geq t$ $\geq t - \epsilon^2 n$	$\geq p_j$ for all clients $\geq p_j - \epsilon$ for $\geq (1 - \epsilon)n$ clients

# LP relaxation

- Notation: let  $n := |V|$ . For any  $z \in \mathbb{R}_+^V$  and  $S \subseteq V$ , let  $z(S) := \sum_{i \in S} z_i$  denote the *volume* or *mass* of  $S$ .



# LP relaxation

- Notation: let  $n := |V|$ . For any  $z \in \mathbb{R}_+^V$  and  $S \subseteq V$ , let  $z(S) := \sum_{i \in S} z_i$  denote the *volume* or *mass* of  $S$ .
- Can guess the optimal radius  $R$  in  $O(n^2)$  time.

# LP relaxation

- Notation: let  $n := |V|$ . For any  $z \in \mathbb{R}_+^V$  and  $S \subseteq V$ , let  $z(S) := \sum_{i \in S} z_i$  denote the *volume* or *mass* of  $S$ .
- Can guess the optimal radius  $R$  in  $O(n^2)$  time.
- Let  $y_i$  be the indicator for “center  $i$  is opened”
- Let  $x_{ij}$  be the indicator for “vertex  $j$  is connected to center  $i$ ”

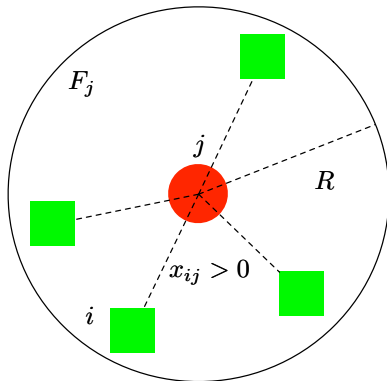
# LP relaxation

- Notation: let  $n := |V|$ . For any  $z \in \mathbb{R}_+^V$  and  $S \subseteq V$ , let  $z(S) := \sum_{i \in S} z_i$  denote the *volume* or *mass* of  $S$ .
- Can guess the optimal radius  $R$  in  $O(n^2)$  time.
- Let  $y_i$  be the indicator for “center  $i$  is opened”
- Let  $x_{ij}$  be the indicator for “vertex  $j$  is connected to center  $i$ ”
- Find a fractional solution  $(x, y)$  such that

$$\left\{ \begin{array}{ll} \sum_{i: d_{ij} \leq R} x_{ij} \leq 1, \forall j \in C & j \text{ connects to } \leq 1 \text{ center} \\ \sum_{j \in V} \sum_{i: d_{ij} \leq R} x_{ij} \geq t, & \geq t \text{ vertices are covered} \\ x_{ij} \leq y_i, \forall i \in V & j \text{ can only connect to } i \text{ if it's open} \\ y(U) \leq r_{\mathcal{M}}(U), \forall U \subseteq V & \text{matroid rank constraints} \\ 0 \leq x_{i,j}, y_i \leq 1 & \end{array} \right.$$

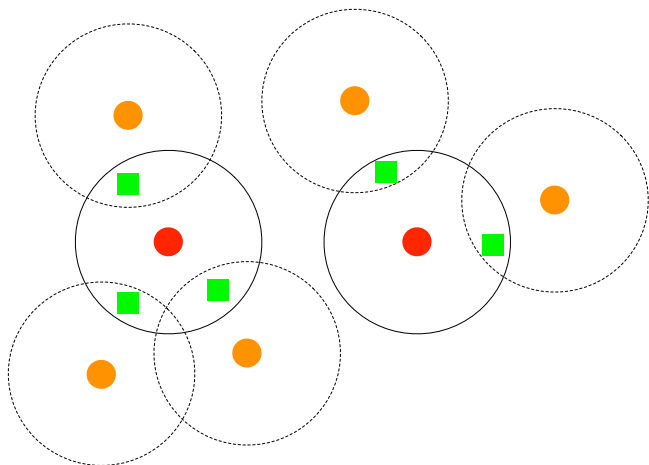
## Filtering step

- For each  $j \in V$ , define *cluster*  $F_j := \{i \in V : x_{ij} > 0\}$
- Let  $s_j := \sum_{i \in F_j} x_{ij}$  denote the “mass” of  $F_j$ . Think of  $s_j$  as the *extent* that  $j$  gets connected.

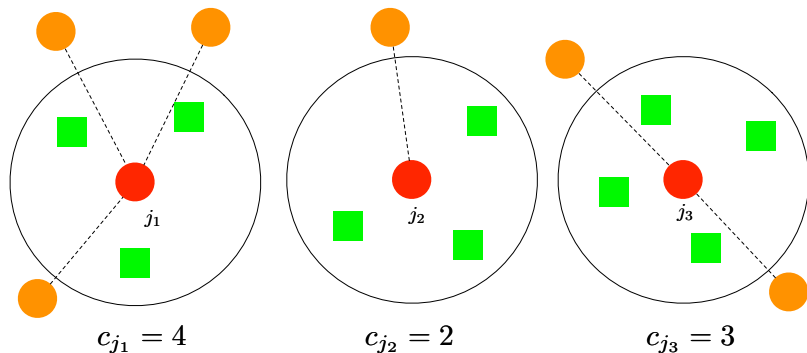



## Filtering step


- Similar to the method by [CGTS '02] for  $k$ -median
- **Sort** all clusters in **decreasing order** of their mass
- Create a **maximal** set of pairwise **disjoint** clusters:
  - for each  $F_j$ : take  $F_j$  and remove all  $F_k$  that  $F_j \cap F_k \neq \emptyset$



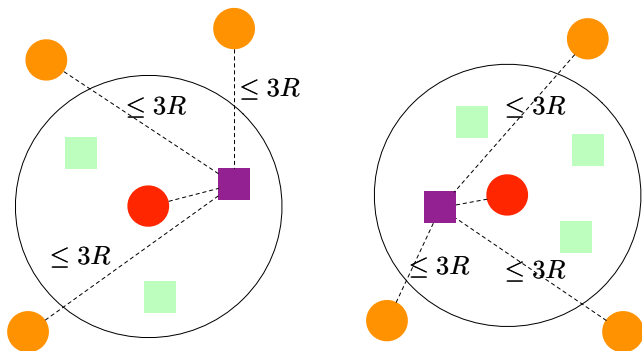
## Filtering step



Let  $V'$  be the set of 

For each  $j \in V'$ , let  $c_j := 1 + \#$   removed by  $j$

# Algorithm



**Idea:** for each  $j \in V'$ , opening 1 center in  $F_j$  will cover  $c_j$  clients within radius  $3R$ .

# Algorithm

1. Define the rounding polytope:

$$\mathcal{P} := \left\{ z \in [0, 1]^V : \underbrace{z(U) \leq r_{\mathcal{M}}(U) \quad \forall U \subseteq V}_{\text{original matroid constraint}} \wedge \underbrace{z(F_i) \leq 1 \quad \forall i \in V'}_{\text{partition matroid constraint}} \right\}$$

→  $\mathcal{P}$  has integral extreme points.



# Algorithm

1. Define the rounding polytope:

$$\mathcal{P} := \left\{ z \in [0, 1]^V : \underbrace{z(U) \leq r_{\mathcal{M}}(U) \forall U \subseteq V}_{\text{original matroid constraint}} \wedge \underbrace{z(F_i) \leq 1 \forall i \in V'}_{\text{partition matroid constraint}} \right\}$$

→  $\mathcal{P}$  has integral extreme points.

2. Find an extreme point  $Z \in \mathcal{P}$  which maximizes  $f$  defined as

$$f(z) := \underbrace{\sum_{j \in V'} c_j \cdot z(F_j)}_{\text{\# clients covered by } z}$$

3. Output  $\mathcal{S} := \{i \in V : Z_i = 1\}$

## Analysis (sketch)

- Claim:  $\mathcal{P} \neq \emptyset$ . Define  $z'_i := \begin{cases} x_{ij} & \text{if } i \in F_j \text{ for some } j \in V' \\ 0 & \text{otherwise} \end{cases}$   
 $\implies z' \in \mathcal{P}$ .

## Analysis (sketch)

- Claim:  $\mathcal{P} \neq \emptyset$ . Define  $z'_i := \begin{cases} x_{ij} & \text{if } i \in F_j \text{ for some } j \in V' \\ 0 & \text{otherwise} \end{cases}$   
 $\implies z' \in \mathcal{P}$ .
- At least  $t$  vertices are covered:

$$\begin{aligned} f(Z) &\geq f(z') && \text{by choice of } Y \\ &= \sum_{j \in V'} c_j \cdot z'(F_j) && \text{by def. of } f \\ &\geq \sum_{j \in V} s_j && \text{by greedy choice of } V' \\ &\geq t && \text{by LP constraints} \end{aligned}$$

# The lottery model

Each client  $j \in \mathcal{C}$  submits a target probability  $p_j$ .

WANT: a minimum radius  $R$  such that  $\exists$  distribution  $\mathcal{D}$  on subsets of  $V$ : for any  $\mathcal{S} \sim \mathcal{D}$ , we have

- Feasibility constraint:  $\mathcal{S}$  is a basis of  $\mathcal{M}$  with prob. 1
- Coverage constraint:  $\#\text{covered clients} \geq t$  with prob. 1
- *Fairness constraint*:  $\Pr[j \text{ is covered}] \geq p_j$  for all  $j$

Our goal: a randomized algorithm which can sample from  $\mathcal{D}$ .

# Modifying LP relaxation

- Again, assume WLOG that  $R$  is the optimal radius.
- Re-define

$$y_i := \Pr[\text{"center } i \text{ is opened"}]$$

$$x_{ij} := \Pr[\text{"vertex } j \text{ is connected to center } i"]$$

- Find a fractional solution  $(x, y)$  such that

$$\left\{ \begin{array}{ll} \sum_{i:d_{ij} \leq R} x_{ij} \geq p_j, \forall j \in \mathcal{C} & \text{fairness constraint} \\ \sum_{i:d_{ij} \leq R} x_{ij} \leq 1, \forall j \in \mathcal{C} & j \text{ connects to } \leq 1 \text{ center} \\ \sum_{j \in V} \sum_{i:d_{ij} \leq R} x_{ij} \geq t, & \geq t \text{ clients are covered} \\ x_{ij} \leq y_i, \forall i \in V & j \text{ can only connect to } i \text{ if it's open} \\ y(U) \leq r_{\mathcal{M}}(U), \forall U \subseteq V & \text{matroid rank constraints} \\ 0 \leq x_{i,j}, y_i \leq 1 & \end{array} \right.$$

# First attempt

Recall that  $z'_i := \begin{cases} x_{ij} & \text{if } i \in F_j \text{ for some } j \in V' \\ 0 & \text{otherwise} \end{cases}$ , and

$$\mathcal{P} := \left\{ z \in [0, 1]^V : \underbrace{z(U) \leq r_{\mathcal{M}}(U) \forall U \subseteq V}_{\text{original matroid constraint}} \wedge \underbrace{z(F_i) \leq 1 \forall i \in V'}_{\text{partition matroid constraint}} \right\}$$

**A natural approach:** decompose  $z' \in \mathcal{P}$  into a convex combination of vertices of  $\mathcal{P}$  and randomly choose a basis  $Y$  according to this distribution

- $Y$  is always a basis of  $\mathcal{M}$
- preserving marginals  $\implies$  fairness constraint is satisfied
- coverage constr. is only satisfied in expectation

# Matroid polytopes

Let  $\mathcal{M} = (\Omega, \mathcal{I})$  be any matroid. The matroid base polytope of  $\mathcal{M}$  is

$$\mathcal{P}_{\mathcal{M}} := \{x \in [0, 1]^{\Omega} : x(\mathcal{S}) \leq r_{\mathcal{M}}(\mathcal{S}) \forall \mathcal{S} \subseteq \Omega; \quad x(\Omega) = r_{\mathcal{M}}(\Omega)\}.$$

# Matroid polytopes

Let  $\mathcal{M} = (\Omega, \mathcal{I})$  be any matroid. The matroid base polytope of  $\mathcal{M}$  is

$$\mathcal{P}_{\mathcal{M}} := \{x \in [0, 1]^{\Omega} : x(S) \leq r_{\mathcal{M}}(S) \forall S \subseteq \Omega; \quad x(\Omega) = r_{\mathcal{M}}(\Omega)\}.$$

## Definition

Suppose  $Ax \leq b$  is a valid inequality of  $\mathcal{P}_{\mathcal{M}}$ . A face  $D$  of  $\mathcal{P}_{\mathcal{M}}$  (w.r.t this inequality) is the set  $D := \{x \in \mathcal{P}_{\mathcal{M}} : Ax = b\}$ .



# Matroid polytopes

Let  $\mathcal{M} = (\Omega, \mathcal{I})$  be any matroid. The matroid base polytope of  $\mathcal{M}$  is

$$\mathcal{P}_{\mathcal{M}} := \{x \in [0, 1]^{\Omega} : x(S) \leq r_{\mathcal{M}}(S) \forall S \subseteq \Omega; \quad x(\Omega) = r_{\mathcal{M}}(\Omega)\}.$$

## Definition

Suppose  $Ax \leq b$  is a valid inequality of  $\mathcal{P}_{\mathcal{M}}$ . A face  $D$  of  $\mathcal{P}_{\mathcal{M}}$  (w.r.t this inequality) is the set  $D := \{x \in \mathcal{P}_{\mathcal{M}} : Ax = b\}$ .

## Theorem (Edmonds)

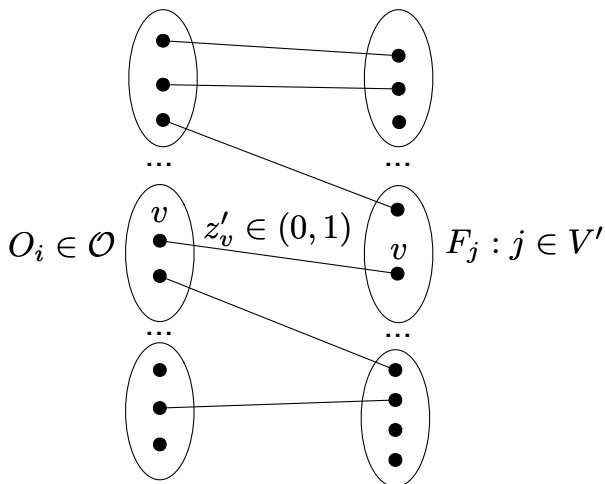
*Any face  $D$  of  $\mathcal{P}_{\mathcal{M}}$  can be characterized by*

$$D = \{x \in \mathcal{P}_{\mathcal{M}} : x(T) = b_T \quad \forall T \in \mathcal{O}; \quad x_i = 0 \quad \forall i \in J\},$$

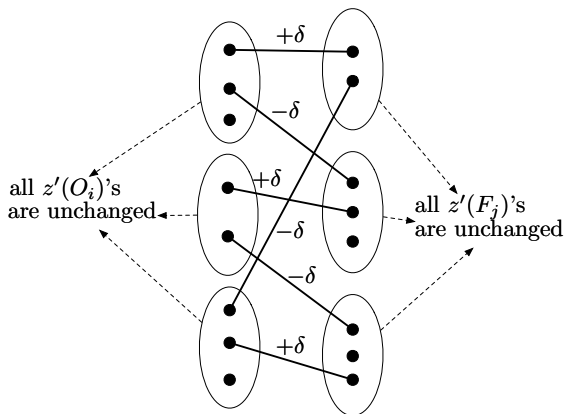
*where  $J \subseteq \Omega$  and  $\mathcal{O}$  is a family of pairwise disjoint sets:  
 $O_1, O_2, \dots, O_t \subseteq \Omega$ , and  $b_{O_1}, \dots, b_{O_t}$  are some integers.*

## Visualization of $z'$ and $\mathcal{P}$

Compute disjoint sets  $O_1, O_2, \dots, O_t \subseteq V$  and integers  $b_{O_1}, \dots, b_{O_t}$  that characterize the face  $D$  of  $\mathcal{P}$  w.r.t all tight constraints on  $z'$ . Let  $O_0 := V \setminus \bigcup_{i=1}^t O_i$ .

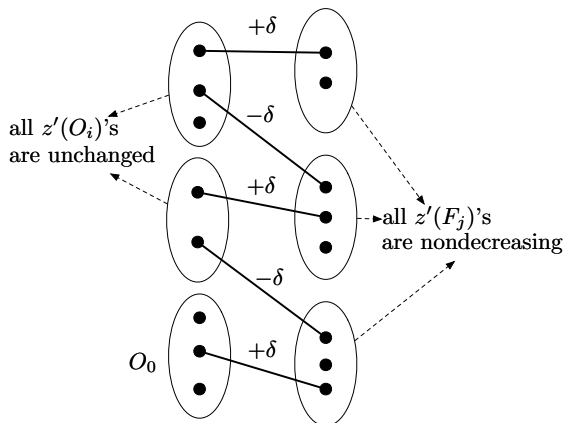


## Case 1: $\exists$ a cycle



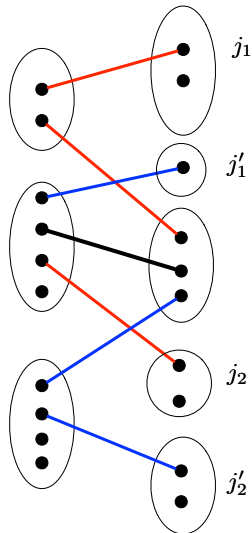
- $\delta$  is maximal so that either some  $z_v$  hits  $\{0, 1\}$  or some rank constraint of  $\mathcal{P}_{\mathcal{M}}$  becomes tight.
- $f(z') = \sum_{j \in V'} c_j \cdot z'(F_j)$  is **unchanged**

## Case 2: $\exists$ a maximal path $O_0$



- Again,  $\delta$  is maximal so that either some  $z_v$  hits  $\{0, 1\}$  or some rank constraint of  $\mathcal{P}_{\mathcal{M}}$  becomes tight.
- $f(z') = \sum_{j \in V'} c_j \cdot z'(F_j)$  is **nondecreasing**

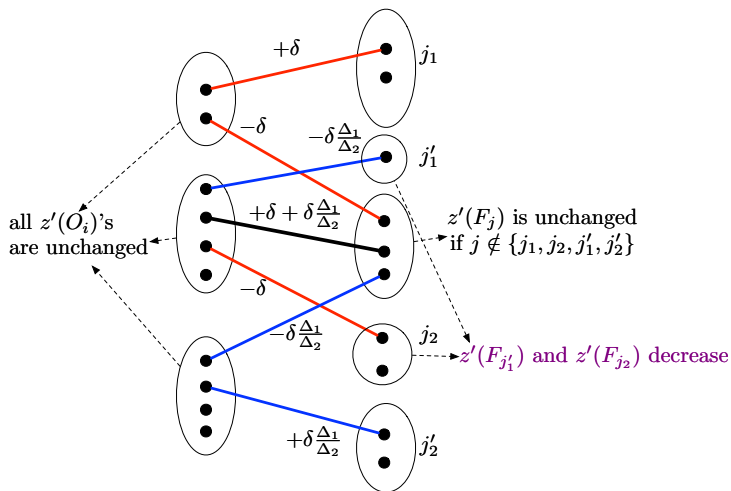
### Case 3: $\exists \geq 2$ maximal paths



WLOG, assume  $c_{j_1} \geq c_{j_2}$   
and  $c_{j'_1} \geq c_{j'_2}$

Define  $\Delta_1 := c_{j_1} - c_{j_2}$   
and  $\Delta_2 := c_{j'_1} - c_{j'_2}$

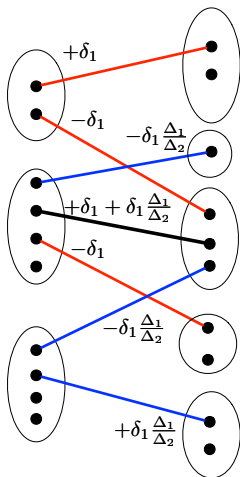
### Case 3: $\exists \geq 2$ maximal paths



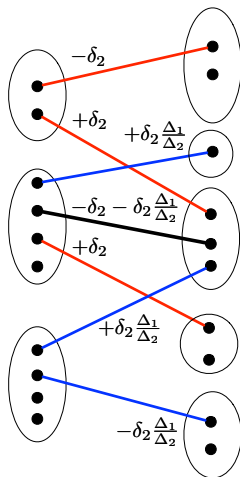
- $\Delta f = \delta c_{j_1} - \delta \frac{\Delta_1}{\Delta_2} c_{j'_1} - \delta c_{j_2} + \delta \frac{\Delta_1}{\Delta_2} c_{j'_2} = 0$  by choice of  $\Delta_1, \Delta_2$

### Case 3: $\exists \geq 2$ maximal paths

With probability  $\frac{\delta_2}{\delta_1 + \delta_2}$ ,

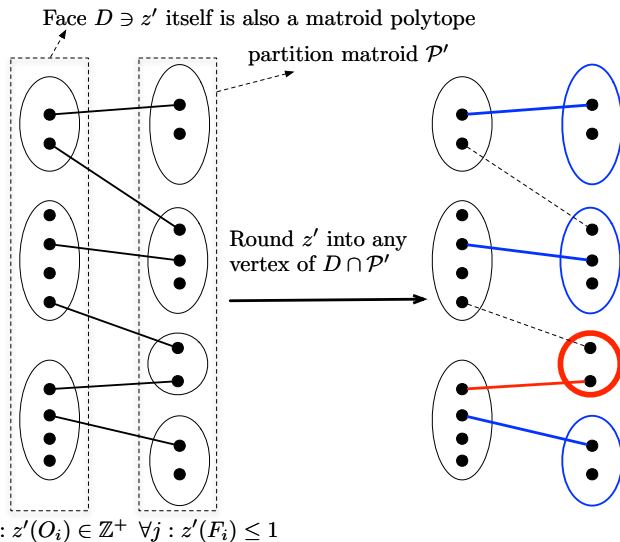


With probability  $\frac{\delta_1}{\delta_1 + \delta_2}$ ,



- E.g.,  $\mathbf{E}[\Delta z'(F_{j_1})] = \delta_1 \cdot \frac{\delta_2}{\delta_1 + \delta_2} - \delta_2 \cdot \frac{\delta_1}{\delta_1 + \delta_2} = 0$ .  
Similarly,  $\mathbf{E}[\Delta z'(F_{j_2})] = \mathbf{E}[\Delta z'(F_{j'_1})] = \mathbf{E}[\Delta z'(F_{j'_2})] = 0$ .

# Remaining case



- $f(z') = \sum_{j \in V'} c_j \cdot z'(F_j)$  is **nondecreasing** if we open one “extra” center



# Analysis (sketch)

- **Feasibility:** Basis + 1 extra center

# Analysis (sketch)

- **Feasibility:** Basis + 1 extra center
- **Coverage:**  $\geq t$  clients are covered with prob. 1 since  $f$  is nondecreasing during the process

# Analysis (sketch)

- **Feasibility:** Basis + 1 extra center
- **Coverage:**  $\geq t$  clients are covered with prob. 1 since  $f$  is nondecreasing during the process
- **Fairness:**
  - For any  $j \in V'$ :

$$\Pr[j \text{ is covered}] = \mathbf{E}[z'(F_j)] \geq \text{original mass of } F_j \geq p_j$$

- For any  $j \notin V'$ : let  $k \in V'$  be the center that removed  $j$

$$\begin{aligned}\Pr[j \text{ is covered}] &= \mathbf{E}[z'(F_k)] \\ &\geq \text{original mass of } F_k \\ &\geq \text{original mass of } F_j \quad \text{by greedy choice} \\ &\geq p_j\end{aligned}$$

Thank you for listening!  
Questions?