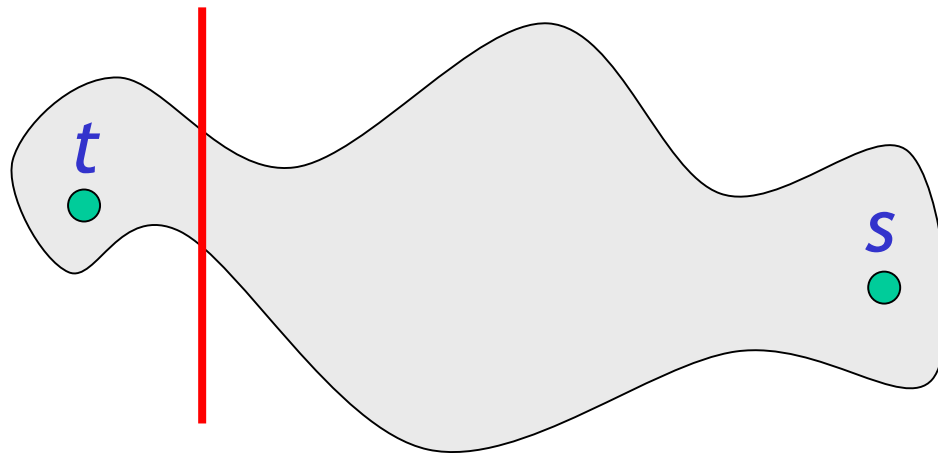


Unbalanced graph cuts

Zoya Svitkina
Cornell University

Joint work with Ara Hayrapetyan,
David Kempe and Martin Pál

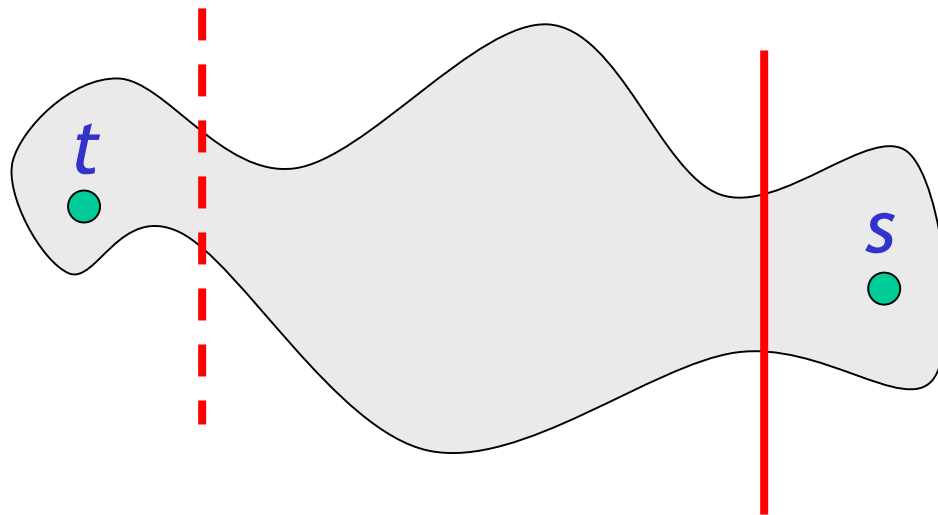
Graph cuts



Min-cut: smallest capacity

Graph cuts

Small capacity and
small size of s-side



Min-cut: smallest capacity

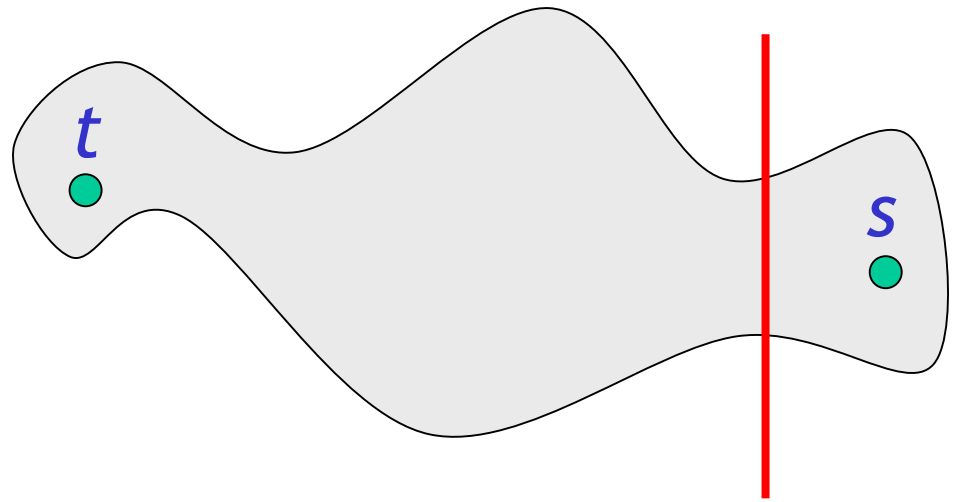
Minimum size bounded capacity cut

Given:

Graph G , nodes s, t , bound B

Find:

- s - t cut (S, T)
 - minimize $|S|$
 - capacity $\leq B$



Applications

- Direct
 - Disaster/military/crime containment
 - Epidemiology (vaccination, quarantine)
 - Computer virus containment
- Indirect (as subroutine)
 - Communities in graphs
 - Others

Related work

- **Balanced/sparsest cuts**
 - T. Leighton, S. Rao (1988)
 - N. Garg, V. Vazirani, M. Yannakakis (1996)
 - S. Arora, S. Rao, U. Vazirani (2004)
- **Cuts of specified size**
 - U. Feige, R. Krauthgamer (2000)
- **Minimum size cuts**
 - S. Eubank, V.S.A. Kumar, M.V. Marathe, A. Srinivasan, N.Wang (2005)

NP-hardness

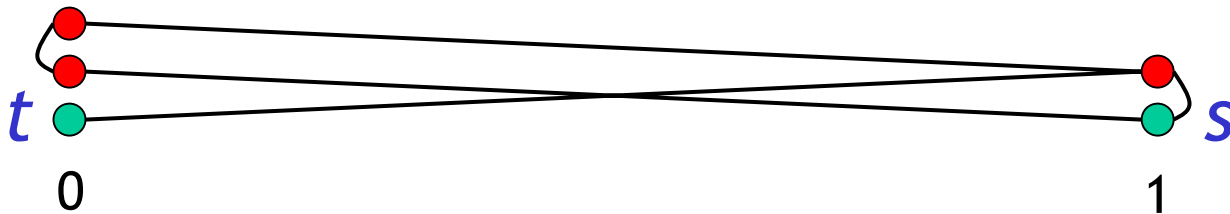
- Weighted problem on trees
 - reduction from Knapsack
- Unweighted problem on general graphs
 - reduction from Clique

Two algorithms

- Bicriteria approximation:
 - find a cut of capacity at most $2 \cdot B$,
 - with at most $2 \cdot OPT$ nodes on s -side
- Techniques:
 - 1) randomized rounding of a linear program
 - 2) parametric minimum cuts,
analysis based on Lagrangian relaxation

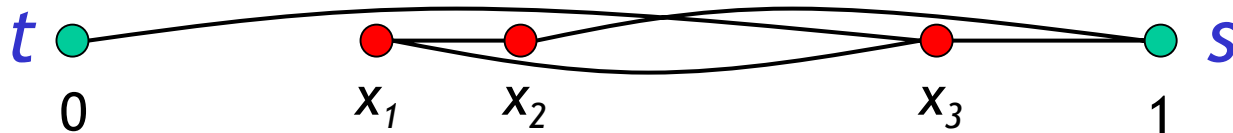
LP used for the algorithms

$$\begin{array}{ll} \text{Minimize} & \sum_{v \in V} x_v \\ \text{subject to} & x_s = 1 \\ & x_t = 0 \\ & y_e \geq |x_u - x_v| \text{ for all } e = (u, v) \in E \\ & \sum_{e \in E} y_e \cdot c_e \leq B \\ & x_v, y_e \geq 0 \end{array}$$



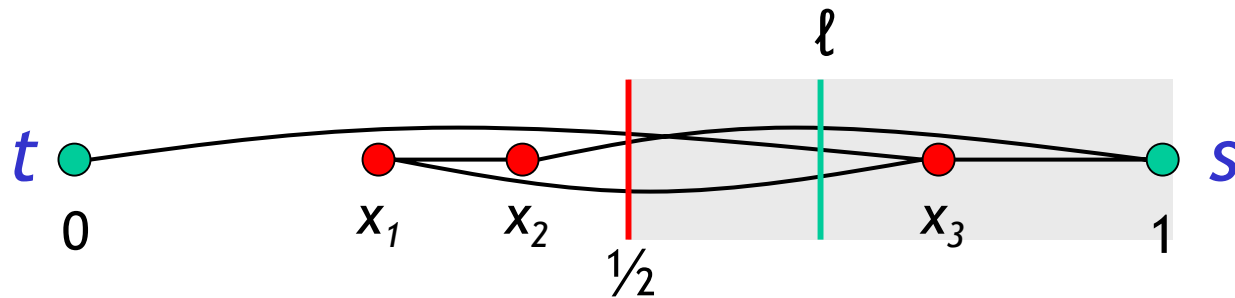
LP used for the algorithms

$$\begin{array}{ll} \text{Minimize} & \sum_{v \in V} x_v \\ \text{subject to} & x_s = 1 \\ & x_t = 0 \\ & y_e \geq |x_u - x_v| \text{ for all } e = (u, v) \in E \\ & \sum_{e \in E} y_e \cdot c_e \leq B \\ & x_v, y_e \geq 0 \end{array}$$



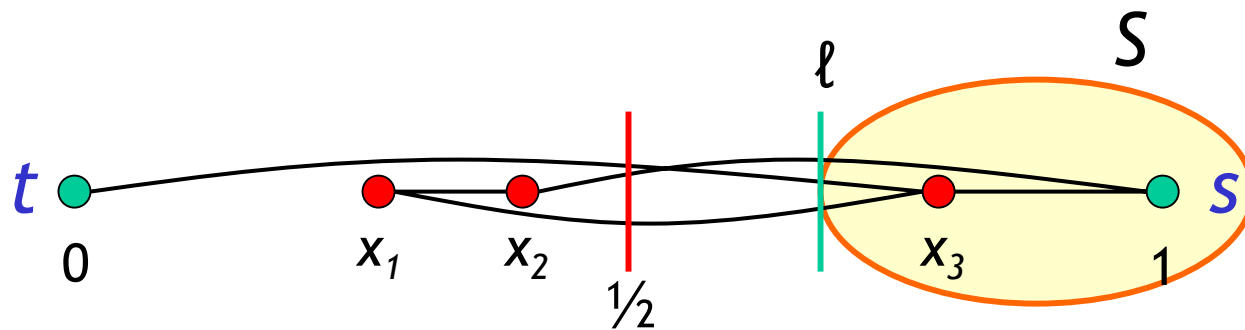
LP rounding algorithm

1. Solve the LP
2. Choose random $\ell \in [1/2, 1]$



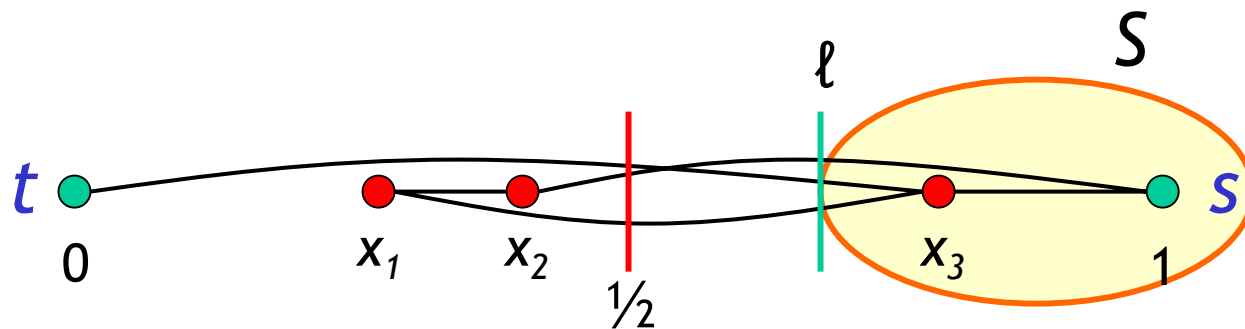
LP rounding algorithm

1. Solve the LP
2. Choose random $\ell \in [1/2, 1]$
3. Return set S of nodes with $x_i \geq \ell$



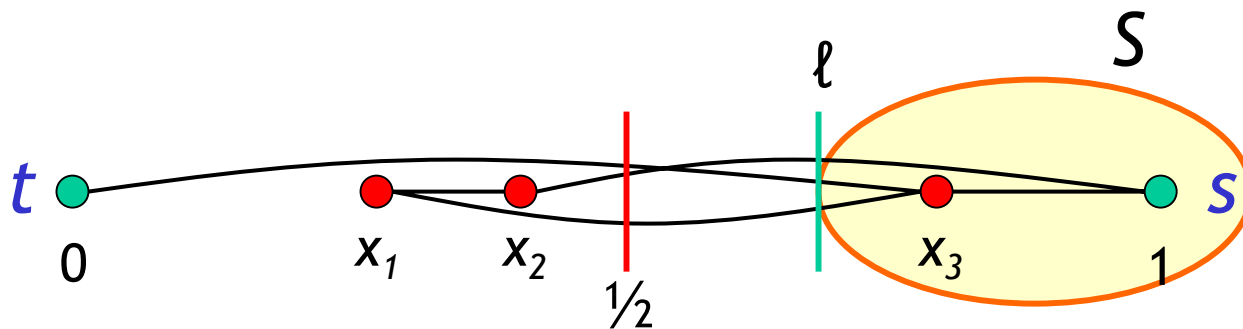
Analysis: (2,2)-bicriteria

- Claim 1: Size of $S \leq 2 \cdot LP \leq 2 \cdot OPT$
 - Proof: For each chosen i , $x_i \geq \frac{1}{2}$



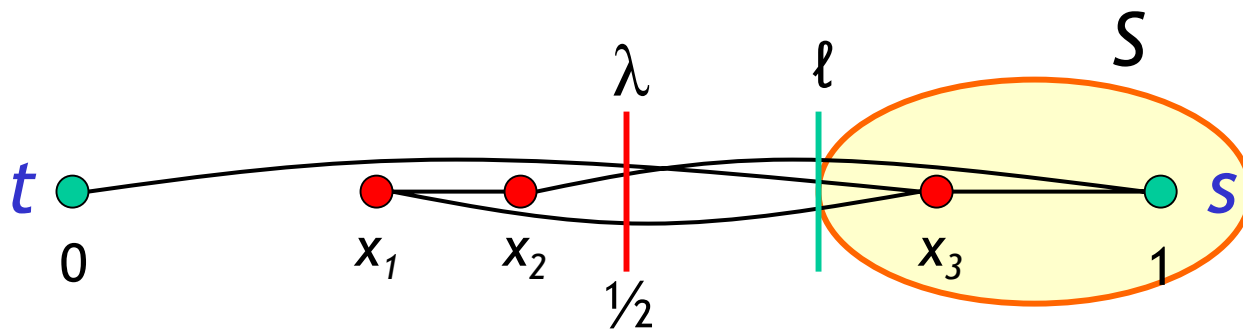
Analysis: (2,2)-bicriteria

- Claim 1: Size of $S \leq 2 \cdot LP \leq 2 \cdot OPT$
 - Proof: For each chosen i , $x_i \geq \frac{1}{2}$
- Claim 2: Expected edges cut $\leq 2 \cdot B$
 - Proof: Choose random ℓ on segment of length $\frac{1}{2}$
 \Rightarrow Prob. cut $e \leq 2 \cdot$ length of e



Extensions to LP rounding algorithm

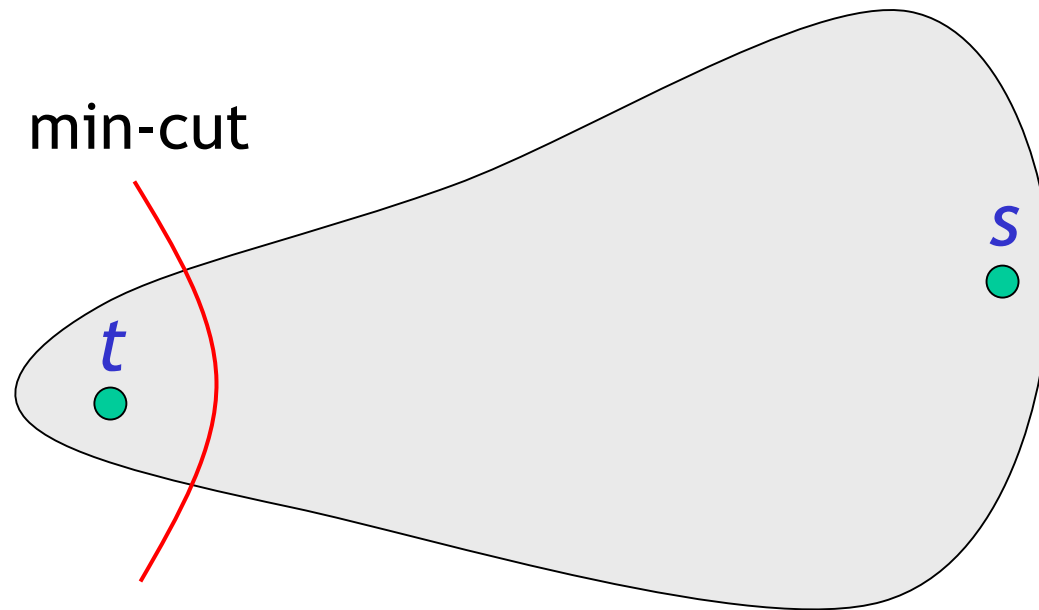
- Derandomize by trying all ℓ
- By using any $\lambda \in (0, 1)$ instead of $\frac{1}{2}$, get a bicriteria algorithm with
 - size of set $\leq \text{OPT}/\lambda$
 - capacity of cut $\leq B/(1-\lambda)$



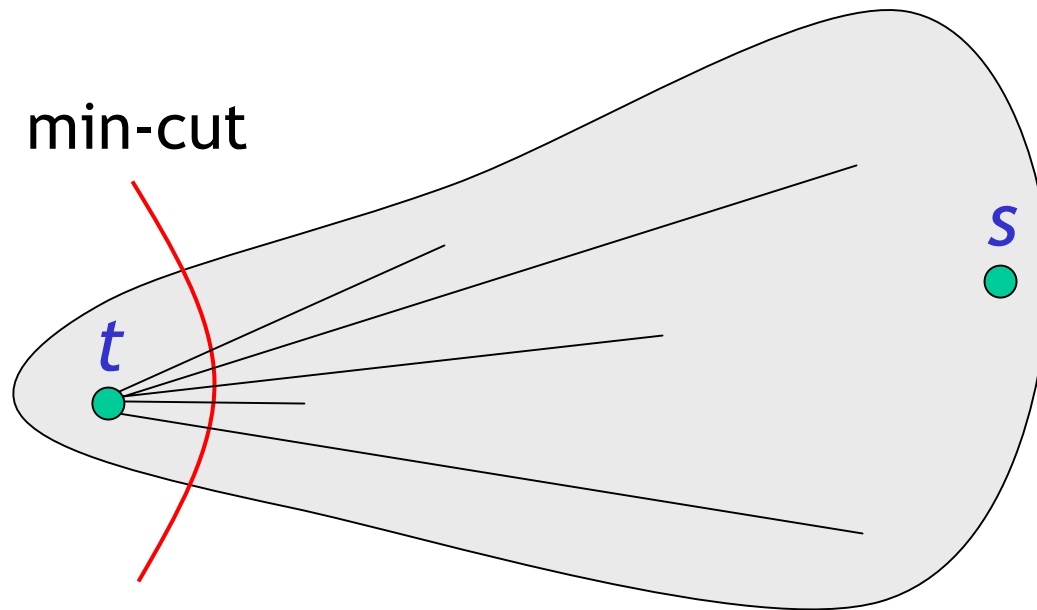
Parametric cut algorithm

- Does not solve the LP
 - running time like min-cut
 - Lagrangian relaxation used for analysis
- Approximates only one criterion:
for a given $\lambda \in (0, 1)$, finds a cut that
 - *either* has size $\leq \text{OPT}/\lambda$ (and capacity $\leq B$)
 - *or* has capacity $\leq B/(1-\lambda)$ (and size optimal)

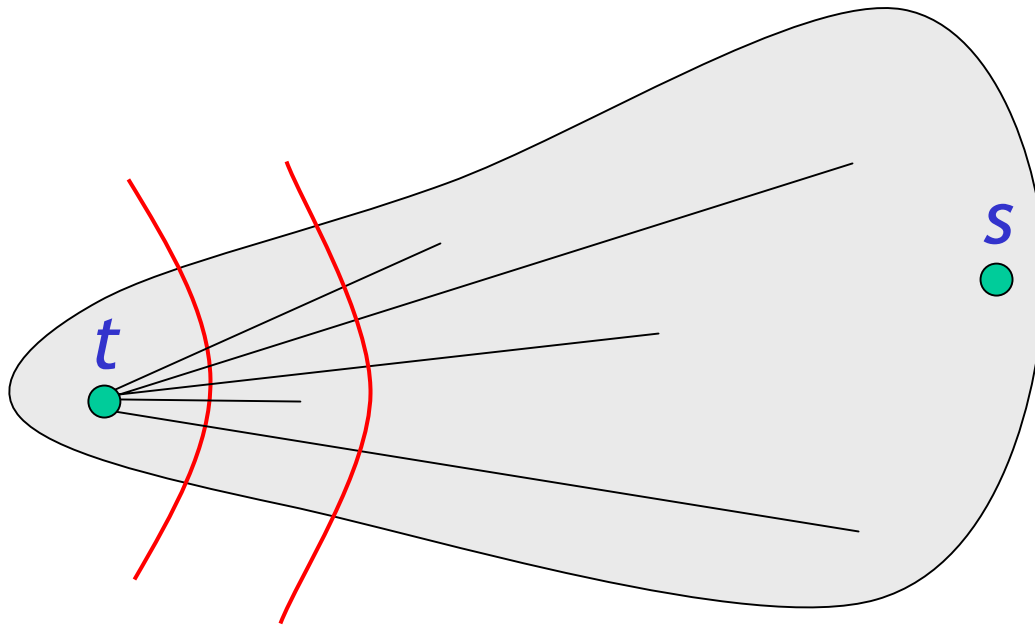
Parametric cut algorithm



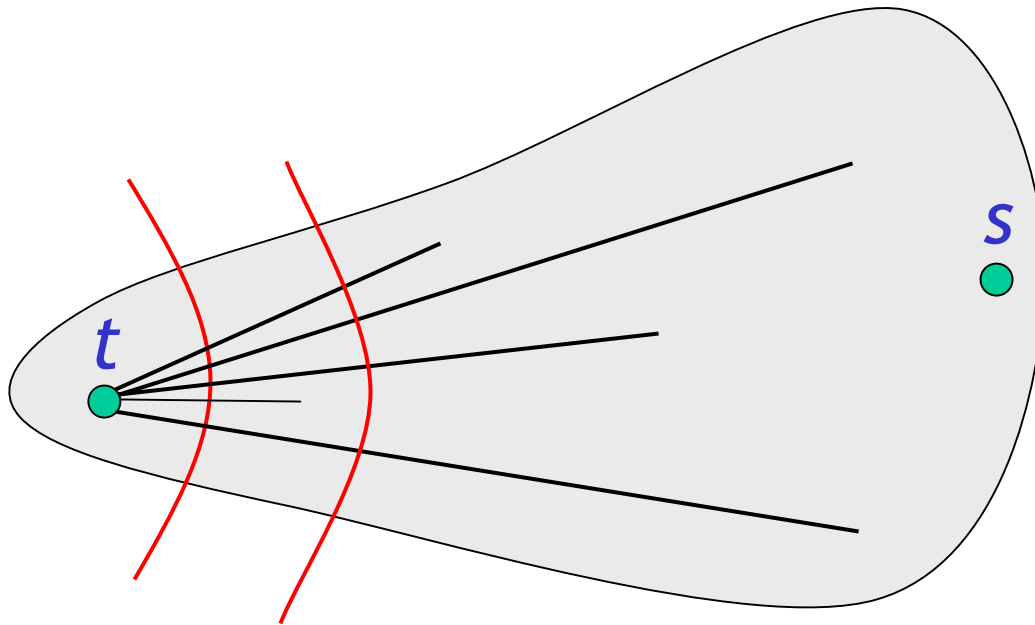
Parametric cut algorithm



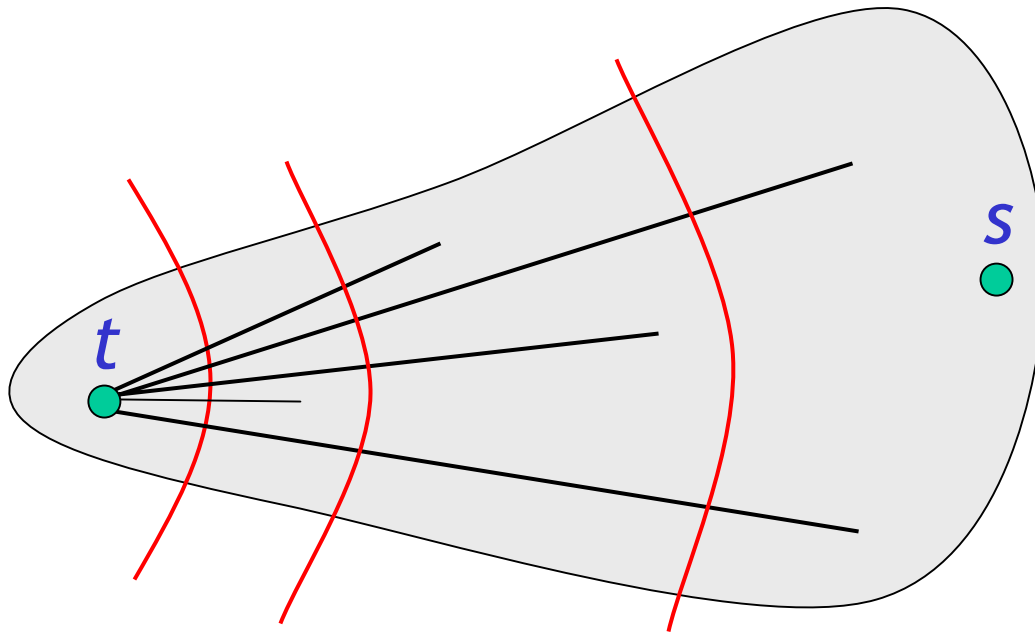
Parametric cut algorithm



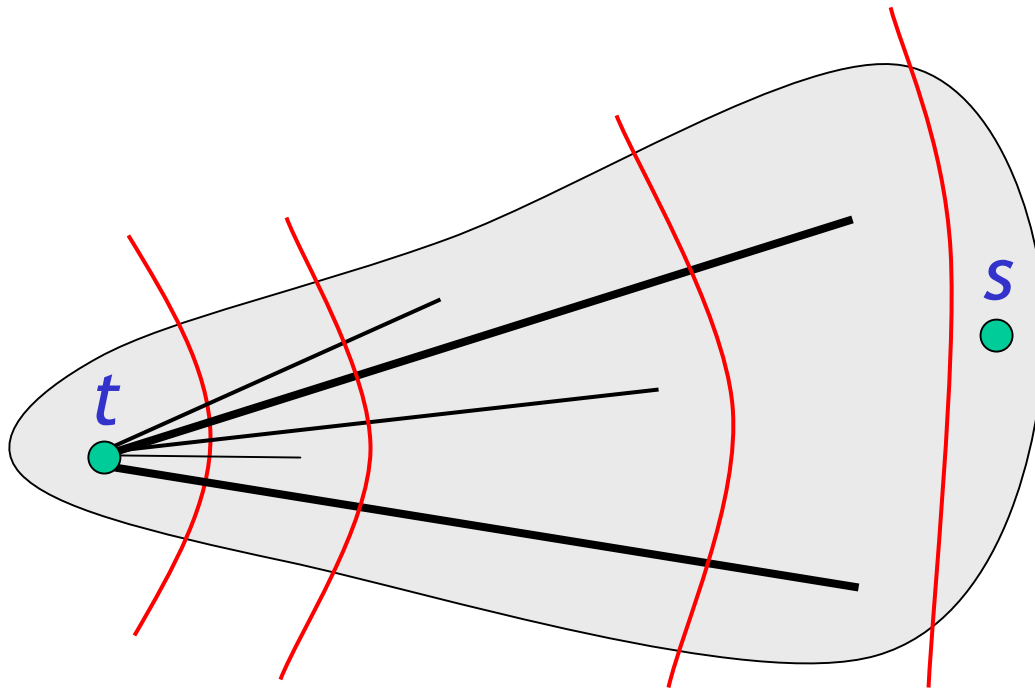
Parametric cut algorithm



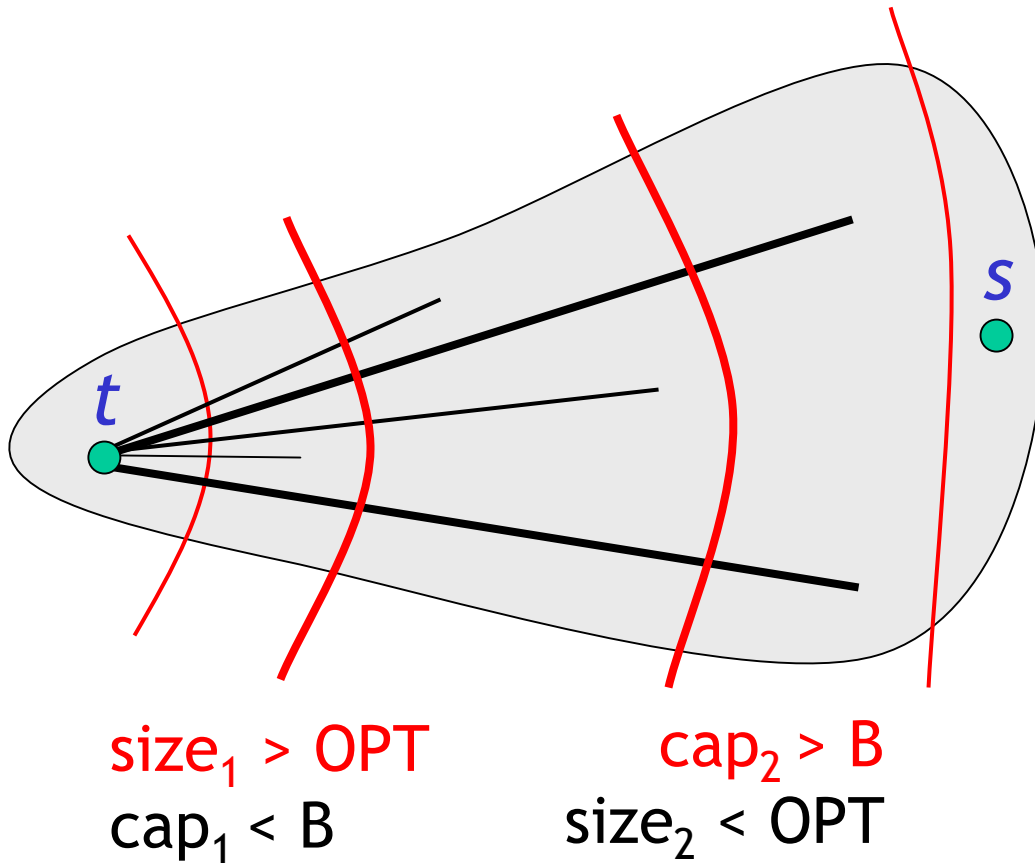
Parametric cut algorithm



Parametric cut algorithm

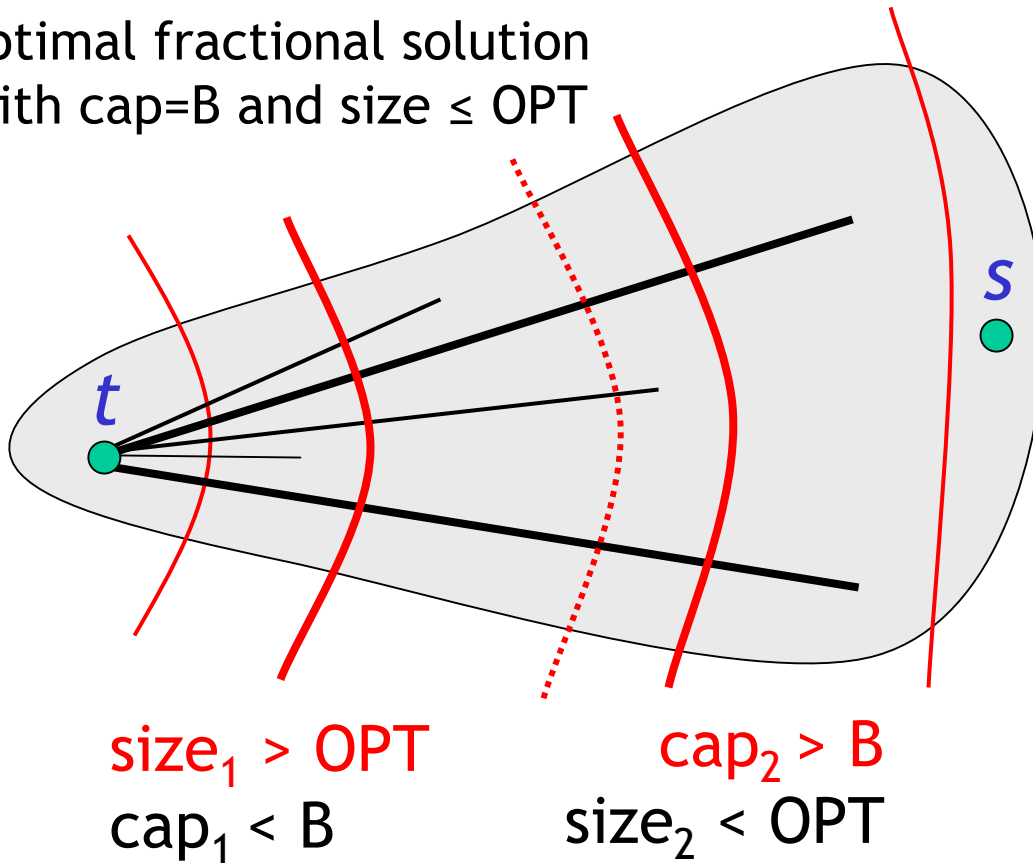


Parametric cut algorithm



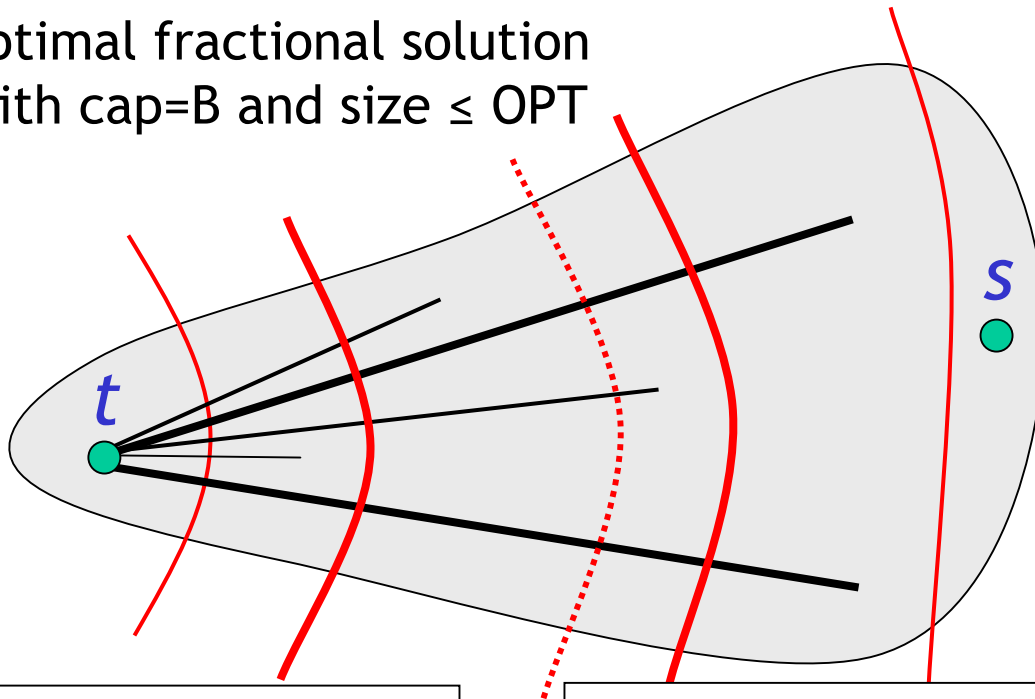
Parametric cut algorithm

a linear combination is an optimal fractional solution with $\text{cap} = B$ and $\text{size} \leq \text{OPT}$



Parametric cut algorithm

a linear combination is an optimal fractional solution with $\text{cap} = B$ and $\text{size} \leq \text{OPT}$



either

$$\begin{aligned} \text{OPT}/\lambda &\geq \text{size}_1 > \text{OPT} \\ \text{cap}_1 &< B \end{aligned}$$

or

$$\begin{aligned} B/(1-\lambda) &\geq \text{cap}_2 > B \\ \text{size}_2 &< \text{OPT} \end{aligned}$$

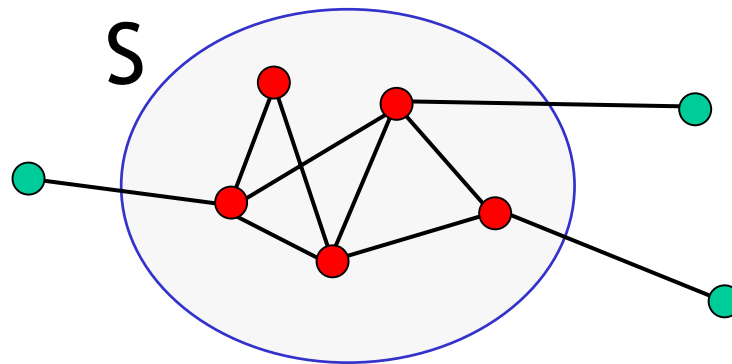
Summary of parametric cut algorithm

- Running time
 - Does not solve the LP
 - Can find all cuts in time of push-relabel algorithm
- Analysis uses Lagrangian relaxation
- Approximates only one criterion

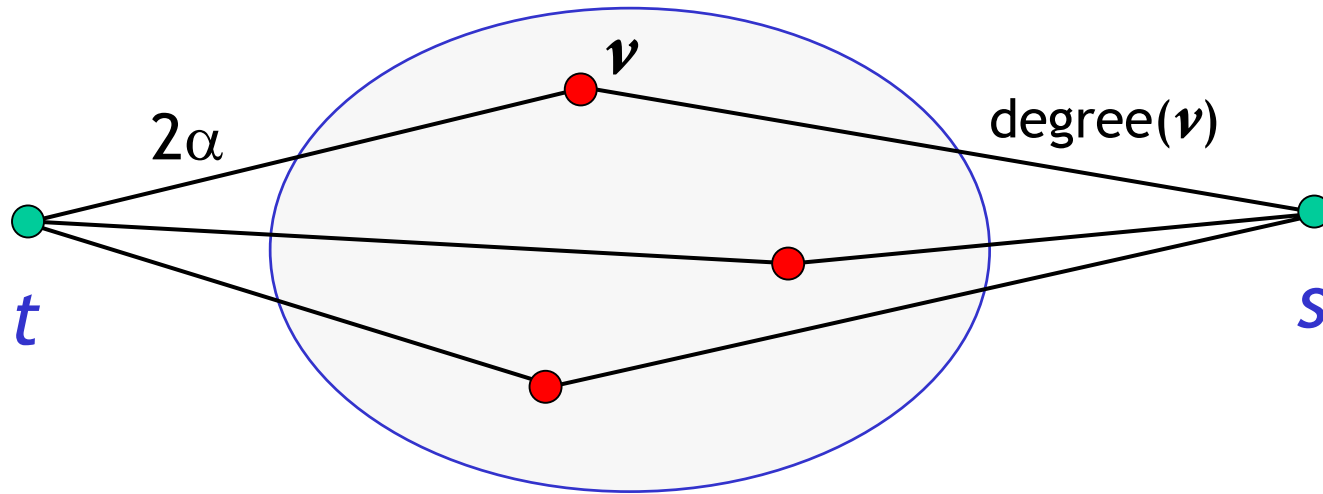
Application: Graph communities

- Define α -community = subgraph S with

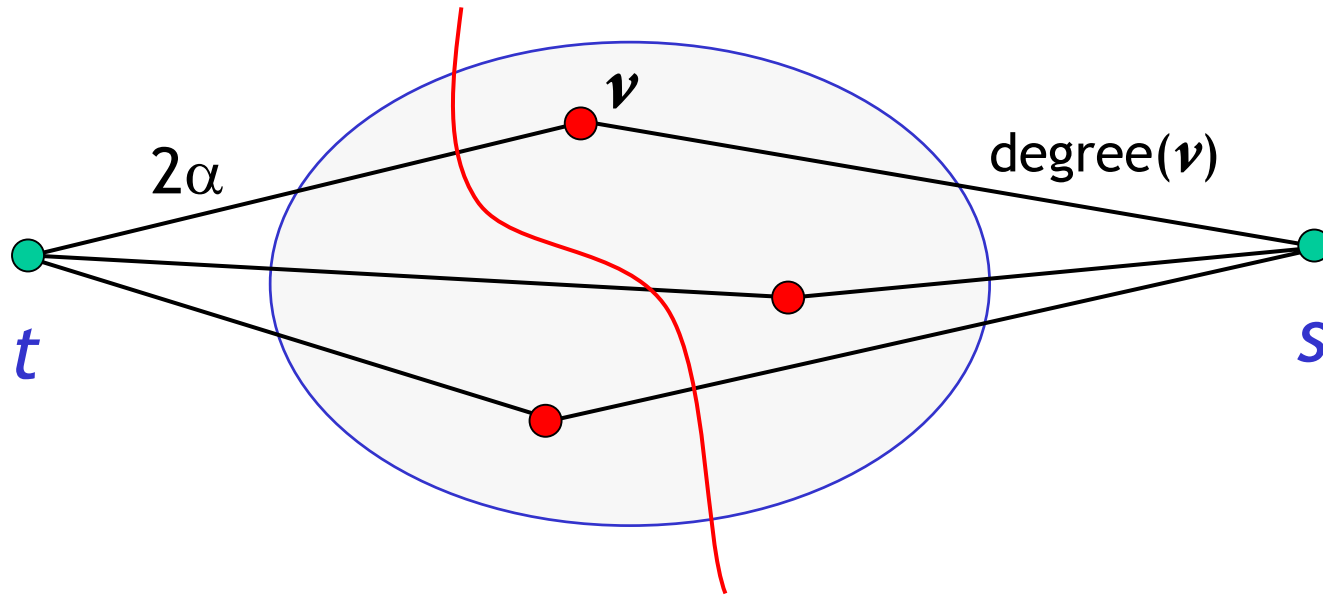
$$\frac{\text{edges inside } S}{\text{total degree in } S} \geq \alpha$$



Finding graph communities

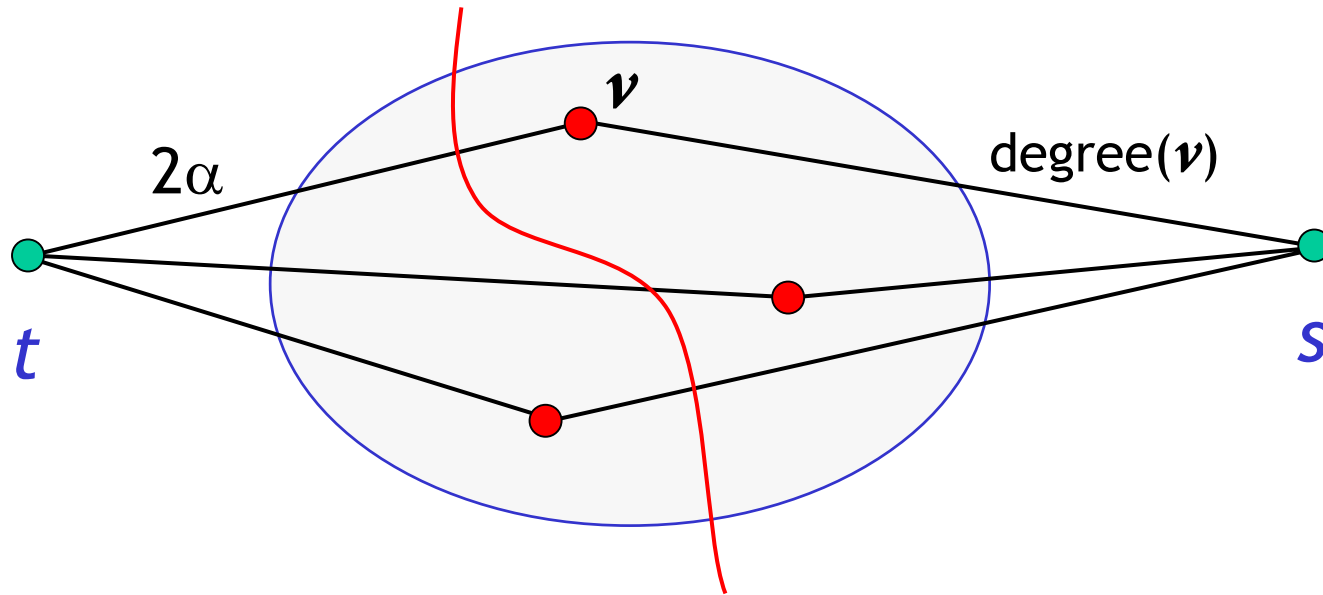


Finding graph communities



Theorem: S is an α -community iff $(S \cup \{s\}, V \setminus S \cup \{t\})$ is a cut of capacity $\leq 2E$

Finding graph communities



Theorem: S is an α -community iff $(S \cup \{s\}, V \setminus S \cup \{t\})$ is a cut of capacity $\leq 2E$

Problem: Whole graph is the best community

Goal: Given α , find the smallest α -community

Other results

- Exact algorithm for unweighted trees and graphs of bounded treewidth (PTAS if weighted)
- Extension to node cuts
- Algorithm producing optimal size and polylogarithmic violation of capacity

Open problems

- Algorithm not violating capacity bound
- Good approximation for maximization version