

Cost-sharing methods in approximation algorithms

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Metric Facility location

Input:

- ▶ undirected graph $G = (V, E)$
- ▶ non-negative edge costs $c : E \rightarrow \mathbb{R}^+$
- ▶ set of **facilities** $F \subseteq V$
- ▶ facility i has facility opening cost f_i
- ▶ set of **demand points** $D \subseteq V$
- ▶ c_{ij} : cost of connecting demand point j to facility i .
Connection cost satisfy triangle inequality

Goal: Compute

- ▶ set $F' \subseteq F$ of opened facilities; and
- ▶ function $\phi : \mathcal{D} \rightarrow \mathcal{F}'$ assigning demand points to opened facilities that minimizes

$$\sum_{i \in F'} f_i + \sum_{j \in \mathcal{D}} c_{\phi(j)j}$$

LP formulation

$$\begin{aligned}
 \min \quad & \sum_{i \in F, j \in D} c_{ij} x_{ij} + \sum_{i \in F} f_i y_i \\
 \text{s.t.} \quad & \sum_{i \in F} x_{ij} \geq 1 && j \in D \\
 & y_i - x_{ij} \geq 0 && i \in F, j \in D \\
 & x_{ij} \in \{0, 1\} && i \in F, j \in D \\
 & y_i \in \{0, 1\} && i \in F
 \end{aligned}$$

- ▶ $y_i = 1$ if facility i is opened;
- ▶ $x_{ij} = 1$ if demand j connected to facility i .

Connected facility location

Input:

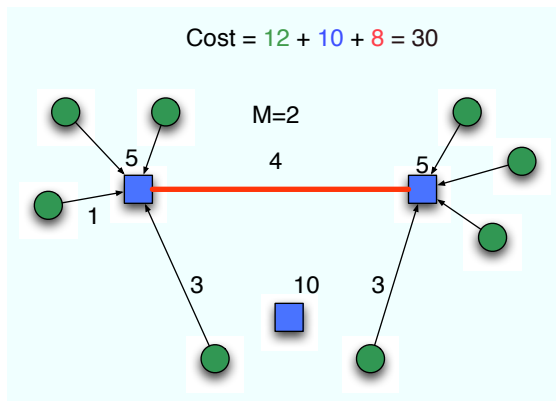
- ▶ Same as facility location; plus
- ▶ Parameter M

Goal: Compute

- ▶ set $F' \subseteq F$ of opened facilities; and
- ▶ function $\phi : \mathcal{D} \rightarrow \mathcal{F}'$ assigning cities to opened facilities; and
- ▶ Steiner tree T connecting the opened facilities that minimizes

$$\sum_{i \in F'} f_i + \sum_{j \in \mathcal{D}} c_{\phi(j)j} + M \sum_{e \in T} c_e$$

Example



Two facilities of cost 5 are opened and connected in a tree

Connected facility location

LP formulation:

Try all possible vertices facilities as root of the Steiner tree T .

$$\begin{array}{ll}
 \min & \sum_{i \in F} f_i y_i + \sum_{i \in F, j \in D} c_{ij} x_{ij} + M \sum_e c_e z_e \\
 \text{s.t.} & \sum_{i \in F} x_{ij} \geq 1 \quad j \in D \\
 & y_i - x_{ij} \geq 0 \quad i \in F, j \in D \\
 & y_v = 1 \\
 & \sum_{i \in S} x_{ij} \leq \sum_{e \in \delta(S)} z_e \quad \forall S \subseteq V, v \notin S, j \\
 & x_{ij}, y_i, z_e \in \{0, 1\} \quad i \in F, j \in D
 \end{array}$$

Primal-dual 9- apx [Swamy and Kumar, 2002].

Idea: Once a demand has contributed to open a facility, it starts paying for the Steiner cost.

Definition: Single-sink rent-or-buy

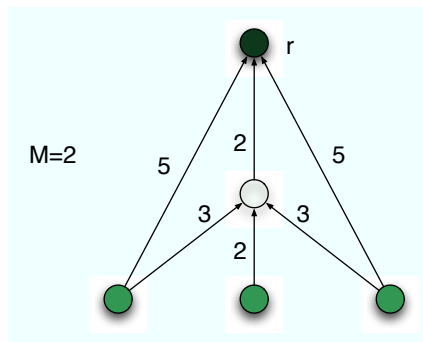
Input:

- ▶ Graph $G = (V, E)$, edge costs $c_e \geq 0$ for all $e \in E$
- ▶ root r
- ▶ Demand points $D = \{v_1, \dots, v_k\} \subseteq V$
- ▶ Flows f_1, \dots, f_k
(here: assume $f_i = 1$ for all i)
- ▶ Economies of scale parameter $M \geq 1$

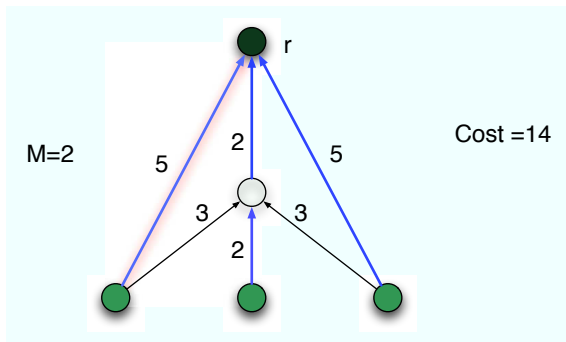
Goal: Find $E_b, E_r \subseteq E$ s.t.

- ▶ $F = E_b \cup E_r$ has an v_i, r -path for all i ,
- ▶ $\sum_{e \in E_r} \lambda(e) \cdot c_e + \sum_{e \in E_b} M \cdot c_e$ is minimum

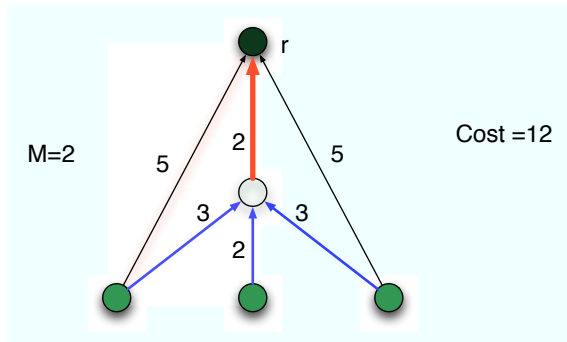
Example



Example



Example



Single-source Rent-or-buy network design

- ▶ SROB is a special case of Connected Facility Location
- ▶ Facilities have 0 opening cost
- ▶ Facilities can be opened at all vertices of the graph
- ▶ A 4.55 approximation primal-dual algorithm given in [Swamy and Kumar, 2002]
- ▶ Simple and elegant solution given in [Gupta, Kumar and Roughgarden, 2003] with 3.55 approximation
- ▶ Also applies to Multi-commodity rent-or-buy (later in this talk), CFL, Virtual Private Network design, Single-sink buy at bulk.

Special Cases

Steiner tree ($M = 1$):

Given a graph $G = (V, E)$, root r , k terminals v_1, \dots, v_k and non-negative edge costs c_e for all $e \in E$.

Find a minimum-cost tree T in G that contains an v_i, r -path for all i .

Shortest Paths ($M = \infty$):

An optimum solution will never buy any edge. Cheapest way of renting capacity f_i between s_i and t_i is along shortest path.

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The GKR approach

Sample-Augment algorithm:

- ▶ **Sample step** Mark each demand with probability $1/M$
- ▶ **Subproblem step** Buy a forest F connecting the set of marked demands R
- ▶ **Augmentation step** Greedily rent capacity to produce a feasible solution.

GKR applied to SROB

Sample-Augment for SROB

W.l.o.g, Consider demand flow $f_j = 1$.

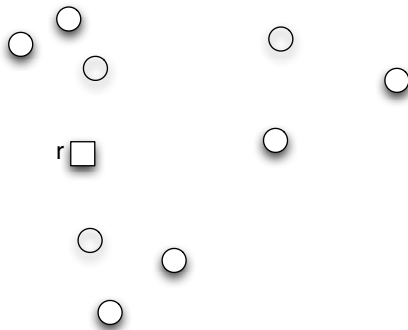
- ▶ **Sample step** Mark each demand with probability $1/M$
- ▶ **Subproblem step** Buy a tree T connecting the set of marked demands R to the root r .
- ▶ **Augmentation step** Connect each demand in D/R to the closest vertex in T .

We separately bound:

- ▶ **Buying cost** incurred in the Subproblem step
- ▶ **Renting cost** incurred in the Augmentation step

Sample-augment: Example

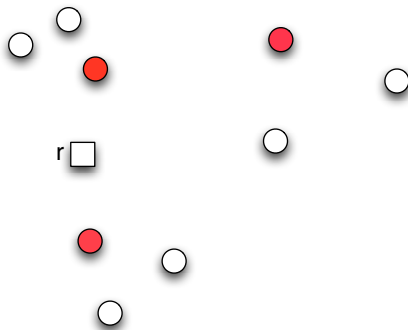
$M=3$



Demands sampled with pb $1/M$

Sample-augment:Example

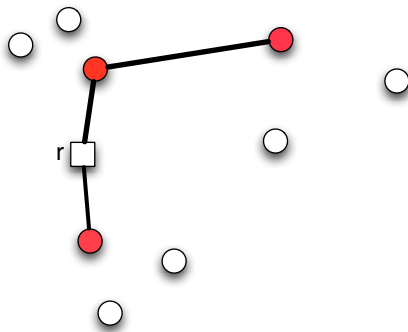
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Three "facilities" opened

Sample-augment: Example

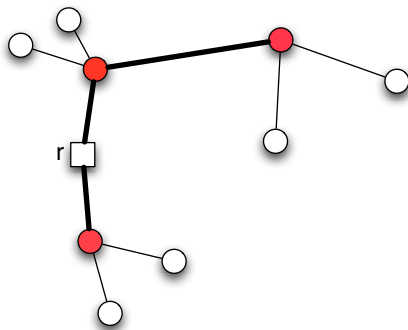
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Build a Steiner tree over sampled demands

Sample-augment:Example

$M=3$



Connect demands to closest facilities

Approximation of SROB

Bounding the buying cost.

- ▶ T^{OPT} : Steiner tree in OPT spanning R^{OPT} .
- ▶ $OPT = M T^{OPT} + \sum_{v \in D/R^{OPT}} c(v, T^{OPT})$

Lemma

$$E_R[T(R)] \leq OPT(D)$$

Proof.

$$E_R[T(R)] \leq M T^{OPT} + \sum_{v \in D/R^{OPT}} \frac{1}{M} M c(v, T^{OPT}) = OPT(D)$$

□

Strict cost-shares

- ▶ We like to distribute in a fair manner between the demands the cost of the subproblem solution
- ▶ Every player should be charged proportionally to its contribution to the cost.

β -strictness

$\xi(v, R)$: cost share of vertex v on sapled set R .

Definition

Cost-shares $\xi(v, R)$ are β -strict if:

- ▶ $\sum_{v \in R} \xi(v, R) \leq T(R)$ **competitiveness**
- ▶ $c(v, T(R/v)) \leq \beta \xi(v, R)$ **strictness**

Strict cost-shares for SROB

Theorem

There exists 2-strict cost shares for Steiner tree.

- ▶ Let us run the Prim algorithm on the set of sampled demands
- ▶ MST is a 2-apx for Steiner tree.
- ▶ Let T_i be the tree constructed on the first i vertices selected by Prim's algorithm.
- ▶ If vertex v is connected by Prim at the $i + 1$ -th iteration, define $\xi_v(R) = \frac{1}{2}c(v, T_i)$.
- ▶ Prim's cost-shares are 2-strict for Steiner tree since:

$$c(v, T(R/v)) \leq c(v, T_i) \leq 2\xi_v(R).$$

Bounding the Renting cost

Proof.

- ▶ Renting cost $R_v = c(v, R)$ ($R_v = 0$ if $v \in R$.)
- ▶ Buying cost $B_v = M\xi(v, R)$ if $v \in R$ ($B_v = 0$ if $v \notin R$.)

Total buying cost: $\sum_{v \in D} B_v = \sum_{v \in R} M\xi(v, R) \leq M T(R)$

- ▶ Renting cost of v : $E[R_v | R] = (1 - \frac{1}{M})c(v, R)$
- ▶ Buying cost of v : $E[B_v | R] = \frac{1}{M}M\xi(v, R \cup v) = \xi(v, R \cup v)$
- ▶ It follows from β strictness: $E[R_v | R] \leq \beta E[B_v | R]$, and
 $E[\sum_{v \in D} R_v] \leq \beta E[\sum_{v \in D} B_v] \leq \beta E[M T(R)] \leq \beta \text{OPT}(D)$



Approximation via Cost-sharing

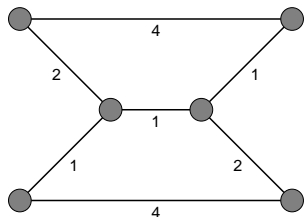
- ▶ One way to obtain strict cost shares is to add extra edges the solution of the subproblem.
- ▶ However, we like to obtain cost shares that are strict for a solution of good quality for the subproblem
- ▶ The approximation we achieve depends on the trade-off between the quality of the approximation to the subproblem and strictness

The strictness theorem:

Theorem

If there exist cost-shares that are competitive and β -strict for an α -approximate algorithm, then Sample-augment is $\alpha + \beta$ -approximated.

Multi-commodity rent-or-buy (MROB)

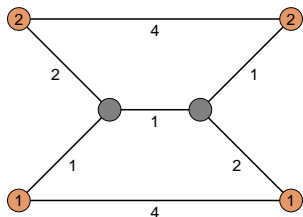


Given:

- ▶ Network $G = (V, E)$ with edge costs c_e for all $e \in E$
- ▶ Terminal pairs $(s_1, t_1), \dots, (s_k, t_k)$
- ▶ Each terminal pair (s_i, t_i) wants to send f_i units of flow from s_i to t_i

Goal: Install capacities on edges such that all flows f_i can be routed simultaneously

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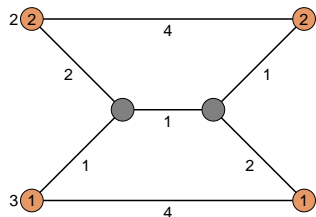


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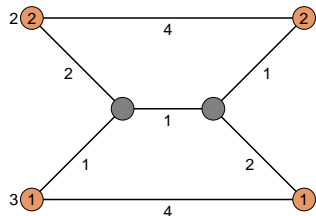


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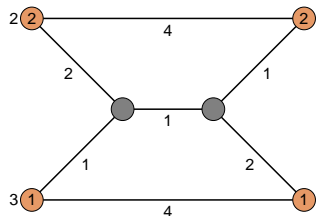


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MROB



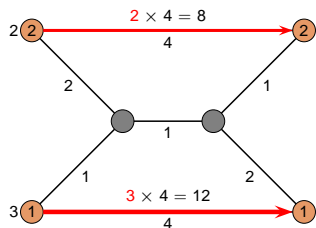
Rent-or-Buy: On each edge e

- ▶ we can either **rent** capacity $\lambda(e)$ at cost $\lambda(e) \cdot c_e$,
- ▶ or **buy** infinite capacity at cost $M \cdot c_e$

Example: $M = 4$

- ▶ Cost of capacity installation: 20
- ▶ Cost of capacity installation: 19

MROB



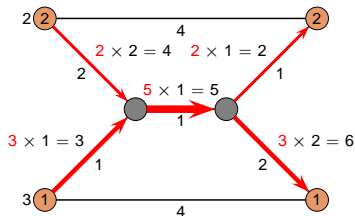
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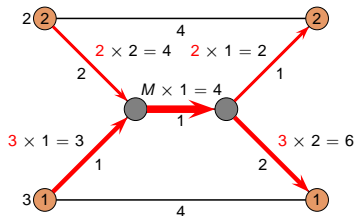
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Definition: Multicommodity Rent-or-Buy

Input:

- ▶ Graph $G = (V, E)$, edge costs $c_e \geq 0$ for all $e \in E$
- ▶ Terminal pairs $R = \{(s_1, t_1), \dots, (s_k, t_k)\} \subseteq V \times V$
- ▶ Flows f_1, \dots, f_k
(here: assume $f_i = 1$ for all i)
- ▶ Economies of scale parameter $M \geq 1$

Goal: Find $E_b, E_r \subseteq E$ s.t.

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Special Cases

Steiner Forests ($M = 1$):

Given a graph $G = (V, E)$, k terminal pairs $(s_1, t_1), \dots, (s_k, t_k)$ and non-negative edge costs c_e for all $e \in E$.

Find a minimum-cost forest F in G that contains an s_i, t_i -path for all i .

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Results on MROB

Multicommodity Rent-or-Buy

Kumar, Gupta, Roughgarden '02	O(1)
Gupta, Kumar, Pál, Roughgarden '03	12
Becchetti, Könemann, L., Pál '05	6.82

Theorem

*There is a 5- apx for the multicommodity rent-or-buy problem.
Fleischer, Könemann, L., Schäfer'06*

Key features:

- ▶ Use the framework of Gupta et al. '03.
- ▶ Alternate view of Steiner forest algorithm by Agrawal, Klein and Ravi '95 gives much simpler analysis.

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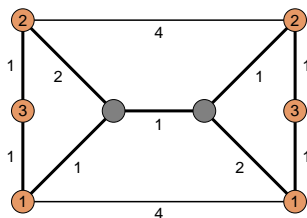
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Sample Augment for MROB[Gupta et al. '03]



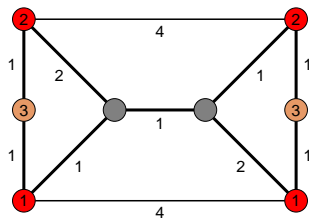
$$M = 3$$

- 1: Mark each terminal pair with probability $1/M$. Marked terminal pairs: D .
- 2: Buy the edges of a Steiner forest E_b for D .
- 3: Rent cheapest set E_r s.t. $F = E_b \cup E_r$ is feasible.

Total cost

$$M \cdot c(E_b) + \sum_{e \in E_r} \lambda(F, e) c_e = 23.$$

Sample Augment for MROB[Gupta et al. '03]



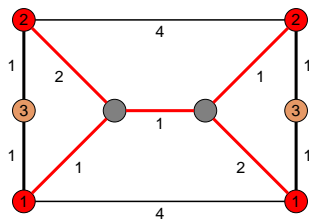
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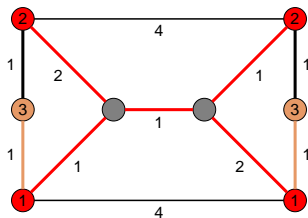
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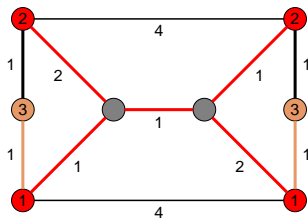
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Sample Augment for MRoB

SimpleMRoB [Gupta et al. '03]:

- 1: Mark each terminal pair with probability $1/M$. Let D be set of marked terminal pairs.
- 2: Compute (approximate) Steiner forest $F' = E_b$ for D and buy all edges in E_b .
- 3: For all terminal pairs $(s, t) \notin D$: Rent unit capacity on shortest s, t -path in contracted graph $G|F'$.

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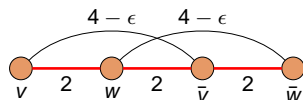
A Randomized Framework for MRoB

Theorem

Given an α -approximate and β -strict Steiner forest algorithm, *SimpleMRoB* returns a feasible solution $F = E_r \cup E_b$ such that

$$E \left[\sum_{e \in E_r} \lambda(e) \cdot c_e + \sum_{e \in E_b} M \cdot c_e \right] \leq (\alpha + \beta) \cdot \text{opt}.$$

Concept: Cost-Sharing



- ▶ An example with 2 terminal pairs $R = \{(v, \bar{v}), (w, \bar{w})\}$.
- ▶ Steiner forest returned by standard primal-dual algorithm AKR is v, \bar{w} -path.

$$\xi(v, \bar{v}) =$$

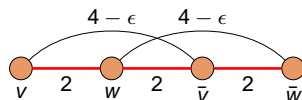
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Cost-Sharing Method:

Want algorithm to compute **cost-share** $\xi(u, \bar{u})$ for all $(u, \bar{u}) \in R$ s.t.

$$\sum_{(u, \bar{u}) \in R} \xi(u, \bar{u}) \leq \text{opt}_R$$

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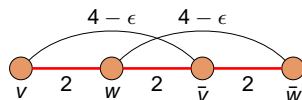
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Cost-Sharing Method:

Want algorithm to compute **cost-share** $\xi(u, \bar{u})$ for all $(u, \bar{u}) \in R$ s.t.

$$\sum_{(u, \bar{u}) \in R} \xi(u, \bar{u}) \leq \text{opt}_R$$

Concept: Cost-Sharing



$$\begin{aligned}\xi(v, \bar{v}) &= 3 \\ \xi(w, \bar{w}) &= 3\end{aligned}$$

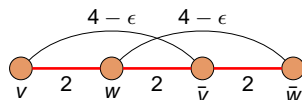
- ▶ An example with 2 terminal pairs $R = \{(v, \bar{v}), (w, \bar{w})\}$.
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Concept: Cost-Sharing



$$\xi(v, \bar{v}) = 1$$

$$\xi(w, \bar{w}) = 4$$

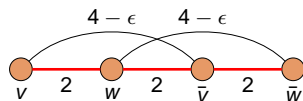
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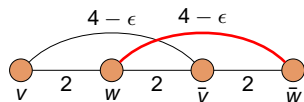
$$\xi(v, \bar{v}) = 3$$

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Notation:

- ▶ $R_{-u\bar{u}}$: all pairs except (u, \bar{u})
- ▶ $F_{-u\bar{u}}$: AKR forest for $R_{-u\bar{u}}$.
Ex: $F_{-v\bar{v}}$.
- ▶ $c_{G|F_{-u\bar{u}}}(z, \bar{z})$: min-cost z, \bar{z} -path in G when edges in $F_{-u\bar{u}}$ are contracted.
Ex: $c_{G|F_{-v,\bar{v}}}(v, \bar{v}) = 4 - \epsilon$

Concept: Strictness



$$\xi(v, \bar{v}) = 3$$

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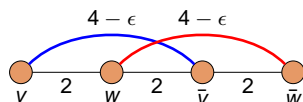
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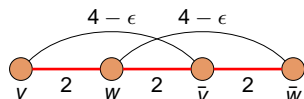
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Definition: Cost-shares ξ are β -strict if

$$c_{G|F_{-u\bar{u}}}(u, \bar{u}) \leq \beta \cdot \xi_{u, \bar{u}}$$

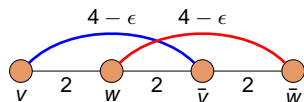
for all $(u, \bar{u}) \in R$.

$$c_{G|F_{-w\bar{w}}}(v, \bar{v}) = 4 - \epsilon \leq \frac{4}{3} \xi(v, \bar{v})$$

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Cost-shares in this example are $\frac{4}{3}$ -strict.

Concept: Strictness



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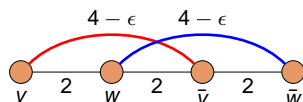
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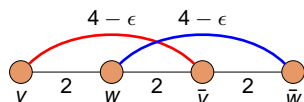
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Concept: Strictness

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A Steiner forest algorithm AKR is β -strict if it returns a cost-share ξ_{st} for all $(s, t) \in R$ such that

1. $\sum_{(s,t) \in R} \xi_{st} \leq c(F^*)$
2. For any $(s, t) \in R$, $c_{G|F_{-st}}(s, t) \leq \beta \cdot \xi_{st}$

Notation:

- ▶ F^* = min-cost Steiner forest for R
- ▶ F_{-st} = apx Steiner forest for $R_{-st} = R \setminus \{(s, t)\}$ computed by AKR
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- ▶ Run AKR on R to compute cost-shares ξ_{st} for all $(s, t) \in R$
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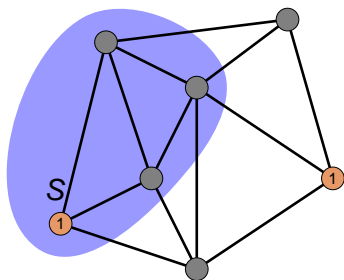
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Remainder of this Lecture

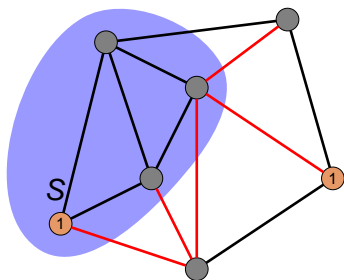
Show that the standard primal-dual algorithm for Steiner forests due to [Agrawal, Klein and Ravi](#) is 2-approximate and 4-strict.

Steiner Cuts



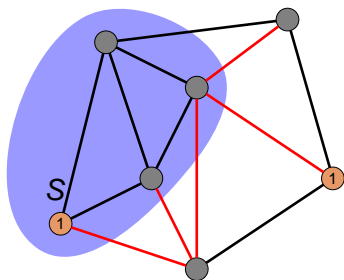
- ▶ A subset $S \subseteq V$ is called a **Steiner cut** if S separates at least one terminal pair
- ▶ Every feasible Steiner forest needs to have one edge crossing every Steiner cut
- ▶ Use \mathcal{U} for the set of all Steiner cuts

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Undirected Cut Relaxation

Primal LP Relaxation:

$$\begin{aligned}
 \min \quad & \sum_{e \in E} c_e \cdot x_e \\
 \text{s.t.} \quad & \sum_{e \in \delta(U)} x_e \geq 1 \quad \forall U \in \mathcal{U} \\
 & x_e \geq 0 \quad \forall e \in E
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$(\delta(U))$: Edges in the cut defined by U

Dual LP:

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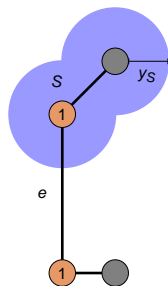
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Visualizing the Dual

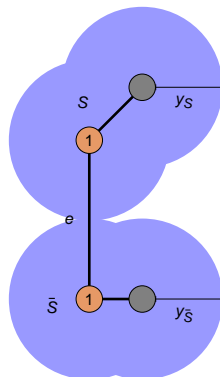


- ▶ The dual y_S of Steiner-cut S is visualized as **moat** around S of radius y_S
- ▶ The dual constraint for edge e is **tight** if

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Here: $y_S + y_{\bar{S}} = c_e$

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Primal-Dual Algorithm

- ▶ Algorithm starts with an empty (infeasible) primal solution F and dual (feasible) solution $y_U = 0$ for all Steiner cuts $U \in \mathcal{U}$
- ▶ Goal: Compute feasible primal/dual pair (F, y) such that cost of F is bounded within dual objective function value, e.g.,

$$c(F) \leq \alpha \cdot \sum_{U \in \mathcal{U}} y_U$$

- ▶ By weak duality, computed solution F is α -approximate Steiner forest

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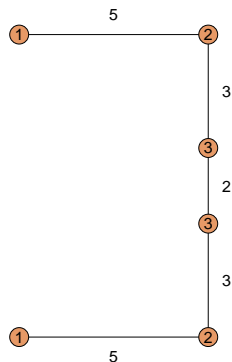
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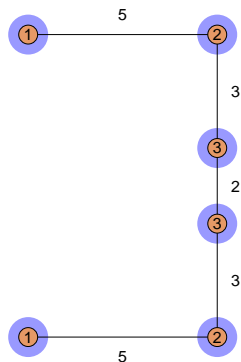
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Primal-Dual Algorithm



- ▶ Initially: Raise duals for all singleton Steiner cuts simultaneously... until some edge/path becomes tight
- ▶ Add tight segment to F
- ▶ Terminal is **active** if it is separated from its mate
- ▶ Raise the duals of active connected components

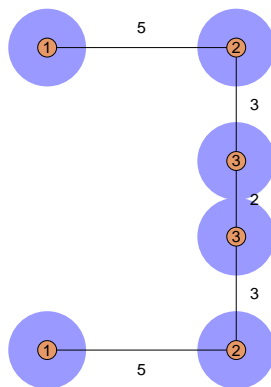
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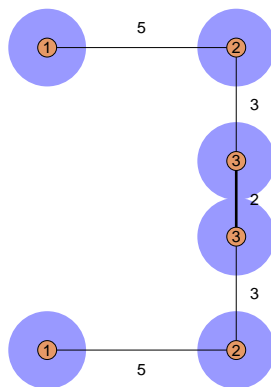
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Time: 1

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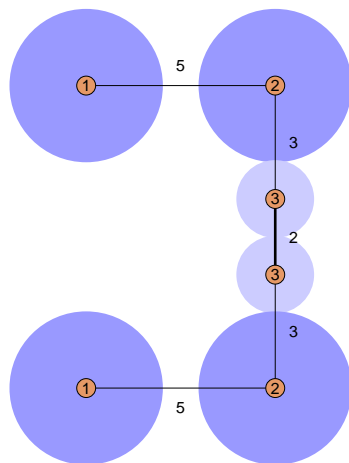
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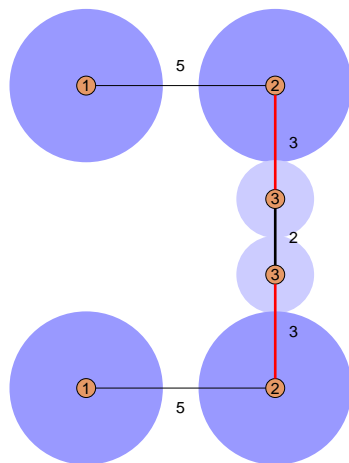
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Time: 2

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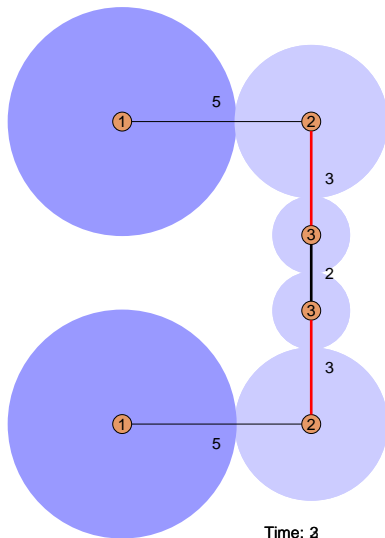
Primal-Dual Algorithm



Time: 2

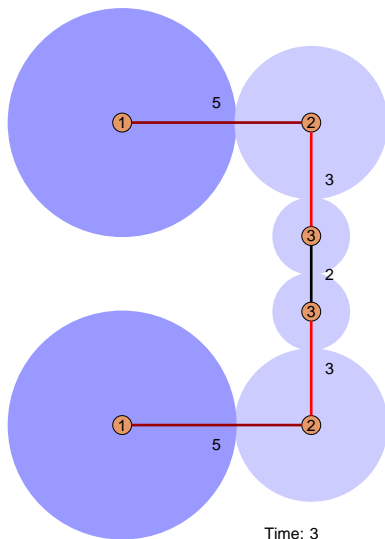
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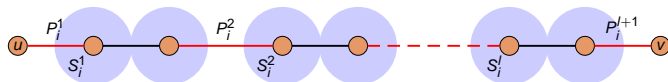
Approximation Guarantee

Theorem (Agrawal, Klein, Ravi '95)

The cost of the computed forest F is

$$c(F) \leq 2 \cdot \sum_{U \in \mathcal{U}} y_U \leq 2 \cdot \text{opt}$$

Primal-Dual Algorithm: Different View

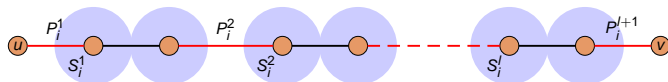


- ▶ Can view execution of algorithm AKR as picking paths

$$P_1, \dots, P_q$$

- ▶ When path P_j becomes tight
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 - ▶ segments P_i^1, \dots, P_i^{l+1} are added to existing forest

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Adding Strictness

1. Need to compute cost-shares ξ_{st} for all $(s, t) \in R$ such that

$$\sum_{(s,t) \in R} \xi_{st} \leq \text{opt}$$

- ▶ Final forest $F = P_1 \cup \dots \cup P_q$ has cost at most $2 \cdot \text{opt}$
- ▶ Whenever a path P_i becomes tight, can distribute half of the cost of the added segments as cost-share
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Total cost-share distributed is $\frac{1}{2}c(F) \leq \text{opt}$.

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- ▶ This implies:
Total cost-share distributed is $\frac{1}{2}c(F) \leq \text{opt}$.

Adding Strictness

1. Need to compute cost-shares ξ_{st} for all $(s, t) \in R$ such that

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2. Need to augment forest F_{-st} at cost

$$c_{G|F_{-st}}(\mathbf{s}, \mathbf{t}) \leq \beta \cdot \xi_{st}$$

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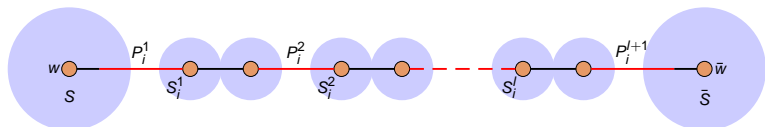
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Crucial Notion: Witnesses



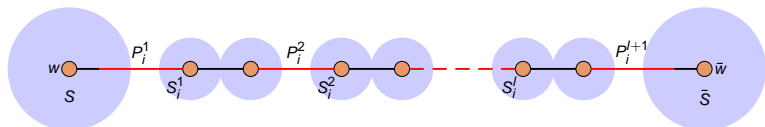
- ▶ Suppose, AKR adds path P_i to connect S and \bar{S} at time τ_i .
- ▶ S and \bar{S} are active \implies both contain active terminals.

Witnesses:

Carefully chosen active terminals w and \bar{w} in S and \bar{S} that are **closest** to P_i .

For all $e \in P_i$, let $\mathcal{W}_e = \{w, \bar{w}\}$ be the set of its witnesses.

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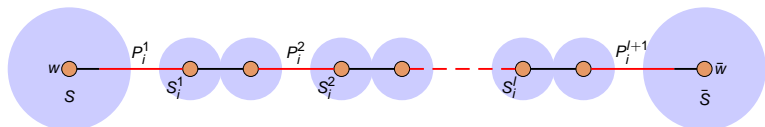
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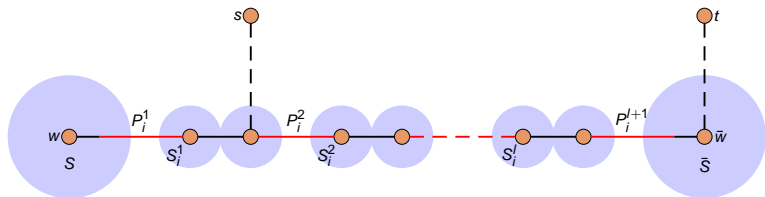
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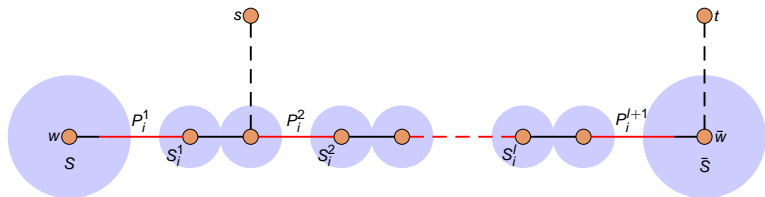
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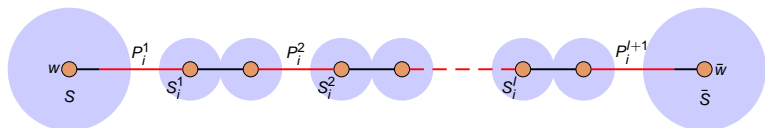
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Symmetric Cost-Sharing

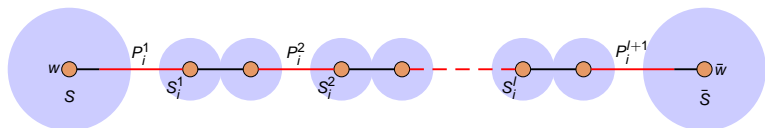


- ▶ w, \bar{w} : witnesses for the edges in e in P_i
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$$\xi_v(e) := \frac{1}{4}c_e$$

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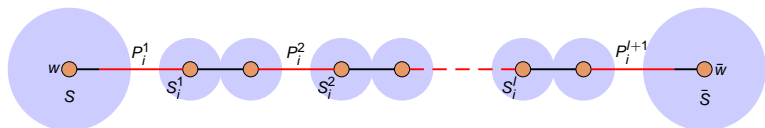


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Theorem: AKR is 2-approximate and 4-strict.

Proof.

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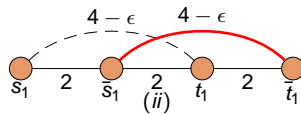
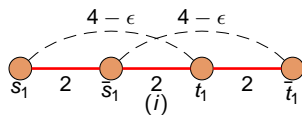
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Bad Examples and Insights



- ▶ Analysis is tight:

Cost-share of (s_1, t_1) for path $\langle s_1, \bar{s}_1, t_1 \rangle$ is 1.

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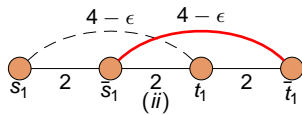
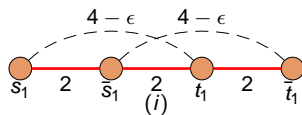
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Total cost-share of (s_1, t_1) in our algorithm is $\frac{3}{2}$.

We could have shown $\frac{4}{3/2} = \frac{8}{3}$ -strictness!

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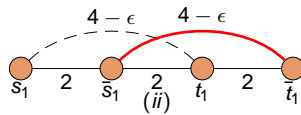
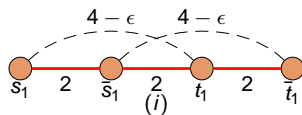
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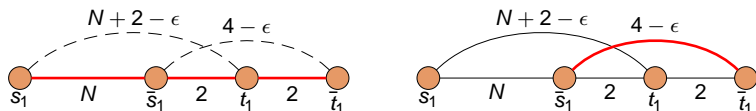
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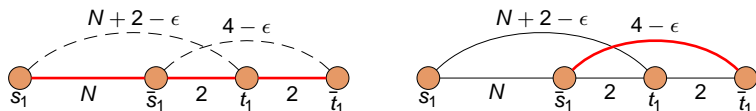


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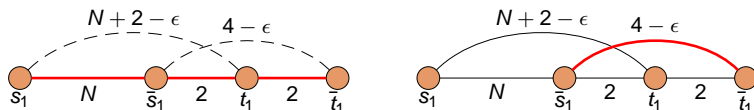


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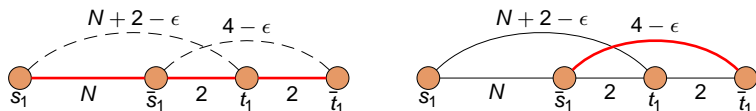


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Conclusion and Open Issues

- ▶ This lecture: AKR is 4-strict \implies 6-apx for MROB
- ▶ The analysis is tight but can be strengthened: replacing **symmetric** by **asymmetric** cost-sharing rule leads to 3-strictness
- ▶ Can also show:
Current Steiner forest algorithms are no better than $\frac{8}{3}$ strict.
Conjecture: AKR is $\frac{8}{3}$ -strict.