

Approximation algorithms for prize-collecting forest problems with submodular penalty functions

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Chaitanya Swamy (University of Waterloo)
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Overview of the talk

- 1 Getting familiar with the problem
- 2 A primal-dual approach
- 3 An LP-rounding approach

1 Getting familiar with the problem

2 A primal-dual approach

3 An LP-rounding approach

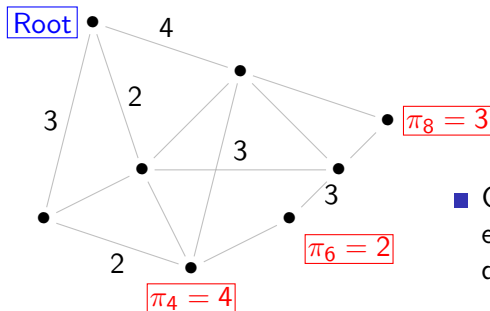
Problems in context: PCST

[Bienstock, Goemans, Simchi-Levi, and Williamson]

Prize-collecting
Steiner tree

Prize-collecting Steiner tree problem (PCST)

[Bienstock, Goemans, Simchi-Levi, and Williamson]



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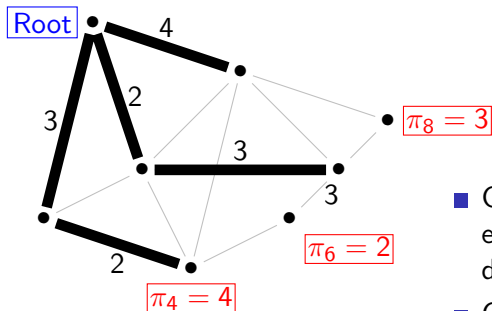
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- Goal: Minimize the cost of edges + sum of penalties of disconnected vertices.

(Not all penalties are shown)

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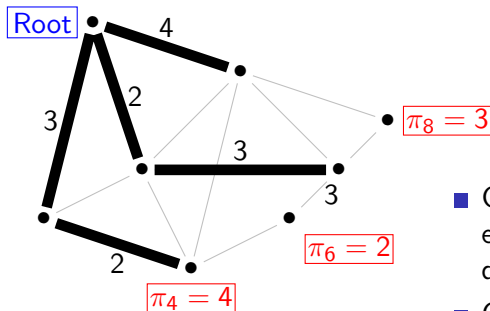
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- Cost of the solution is $(4+2+3+3+2)+(2+3) = 19$.

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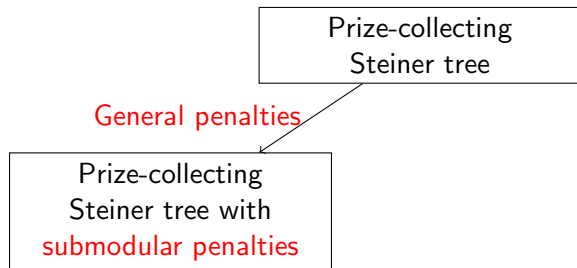
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Known results

A primal-dual 2-approximation by [Goemans and Williamson].

Problems in context: PCST with submodular penalties

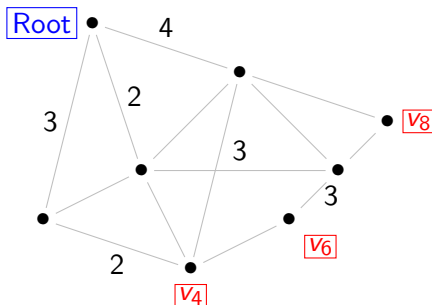
[Hayrapetyan, Swamy, Tardos]



PCST with submodular penalties

[Hayrapetyan, Swamy, Tardos]

$$\pi(A) + \pi(B) \geq \pi(A \cup B) + \pi(A \cap B)$$



$$\pi(\{v_6, v_8\}) = 5; \quad \pi(\{v_4, v_6\}) = 7$$

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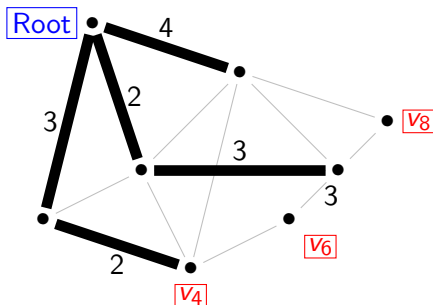
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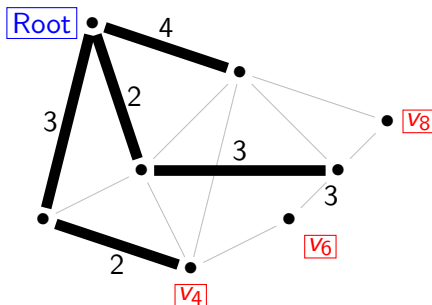
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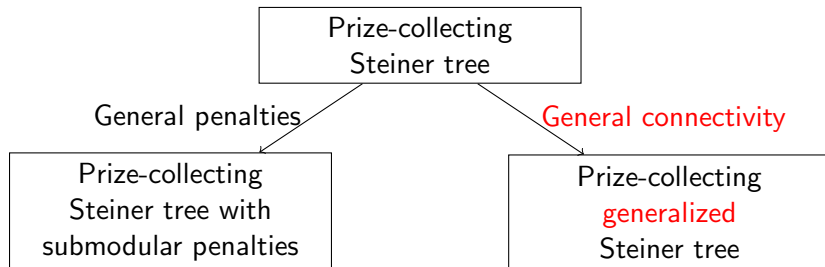
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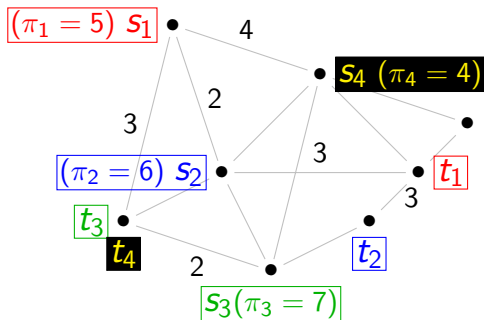
Problems in context: PCGST

[Hajiaghayi and Jain]



Prize-collecting generalized Steiner tree problem (PCGST)

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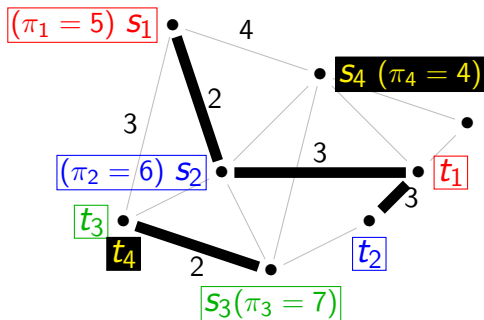
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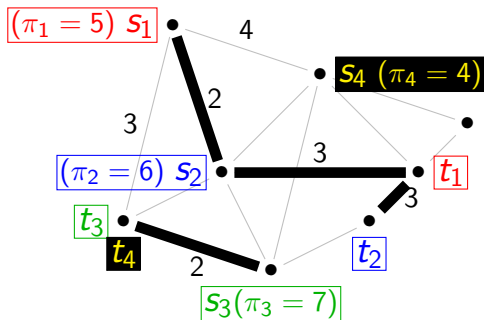
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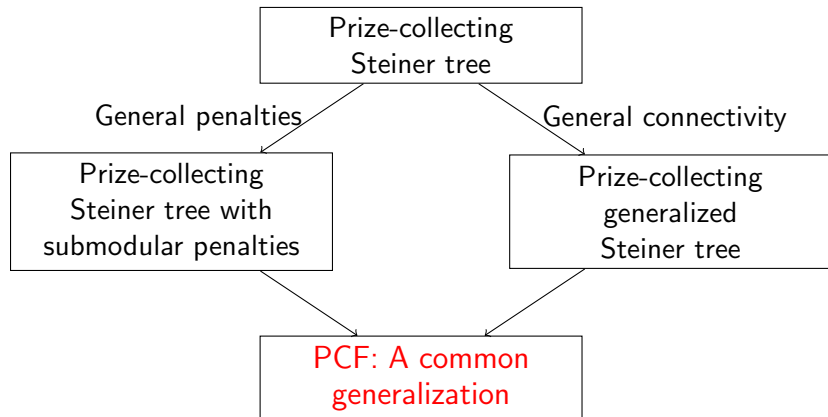
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Known results

A primal-dual 3-approximation, and an LP-rounding 2.54-approximation by [Hajiaghayi and Jain].

Problems in context: Prize-collecting forest problem



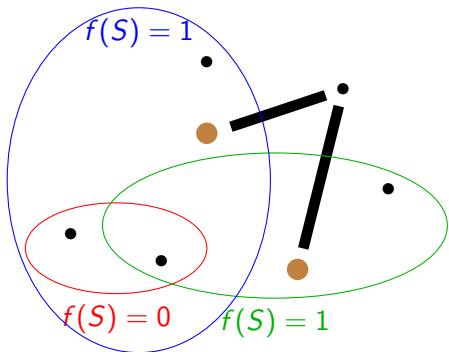
Prize collecting forest problem: Generalization in two dimensions I

Prize-collecting forest problem has two components.

1. Connectivity requirement

■ PCST $\xrightarrow{\text{Generalize}}$ PCGST $\xrightarrow{\text{Generalize}}$ PCF.

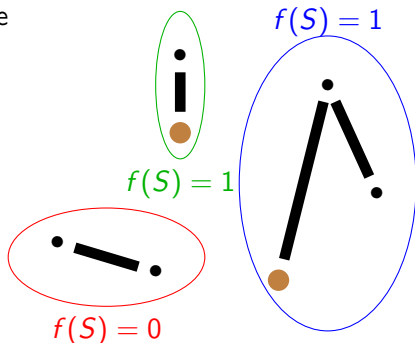
- Generalization to **arbitrary** functions: $f : 2^V \rightarrow \{0, 1\}$.
 - $f(S) = 1$ means at least one edge is required "leaving" S .



Prize collecting forest problem: Generalization in two dimensions II

2. Penalty

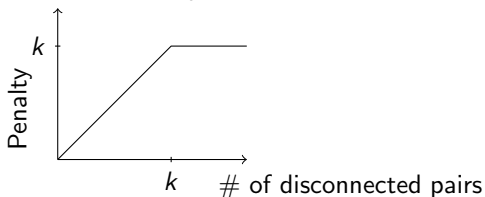
- PCST $\xrightarrow{\text{Generalize}}$ PCST with submodular penalties $\xrightarrow{\text{Generalize}}$ PCF.
- Generalize to wider class of penalty: $\pi : 2^{2^V} \rightarrow \mathbb{N}_+$.
- Given a solution $F \subseteq E$, let \mathcal{F} be the collection of all **violated** subsets.
 - Violated subset: S such that $f(S) = 1$ but there is no edge leaving them ($\delta(S) \cap F = \emptyset$).
- Then, penalty of the solution is $\pi(\mathcal{F})$.
 - $\pi(\cdot)$ should be submodular **[on families of subsets]**.
 - Need other *natural* requirements.



$$\pi(\{\text{Green, Blue}\}) = 8 \text{ (example)}$$

How general is this generalization?

- It captures
 - Prize-collecting Steiner tree with submodular penalties.
 - Prize-collecting generalized Steiner tree.
- Penalty of a pair could depend on already disconnected pairs.
 - Submodular penalty function on set of pairs.
 - Penalty function with a cap.



Our results

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 - A primal-dual 3-approximation.
 - A $\frac{1}{1-e^{-1/2}} \approx 2.54$ approximation via LP-rounding.
- Ideas from [Goemans-Williamson] and [Hajiaghayi-Jain].
 - Contribution: A tractable generalization of two earlier problems.
 - Main issue: Handle the doubly exponential-size LP.
 - Primal-dual algorithm implementation
 - Submodular function minimization and
 - rational approximation.
 - LP-rounding approach
 - There is a short solution.
 - Solve the LP by a novel convex programming formulation.

A primal-dual algorithm

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An LP formulation (Primal program)

[Hajiaghayi and Jain]

- Natural LP.
- Variables for edges: x_e .
- Variables for families: $z_{\mathcal{F}}$.

$$\text{Min} \quad \sum_e c_e x_e \quad + \quad \sum_{\mathcal{F}} \pi(\mathcal{F}) z_{\mathcal{F}} \quad \text{(PCF-LP)}$$

$$\text{s.t.} \quad \underbrace{\sum_{e \in \delta(S)} x_e}_{\text{Edge contribution}} \quad + \quad \underbrace{\sum_{\mathcal{F}: S \in \mathcal{F}} z_{\mathcal{F}}}_{\text{Family contribution}} \quad \geq f(S) \quad \text{for all } S \subseteq V,$$

$$x_e, z_{\mathcal{F}} \geq 0 \quad \text{for all } e, \mathcal{F}.$$

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An LP formulation (Dual program)

- Variables for subsets: y_S .

$$\text{Max} \quad \sum_{S \subseteq V} f(S) \cdot y_S \quad (\text{PCF-D})$$

$$\text{s.t.} \quad \sum_{S: e \in \delta(S)} y_S \leq c_e \quad \text{for all } e \in E \quad (\text{type-(edge)})$$

$$\sum_{S: S \in \mathcal{F}} y_S \leq \pi(\mathcal{F}) \quad \text{for all } \mathcal{F} \subseteq 2^V \quad (\text{type-(family)})$$

$$y_S \geq 0 \quad \text{for all } S \subseteq V.$$

The primal dual algorithm

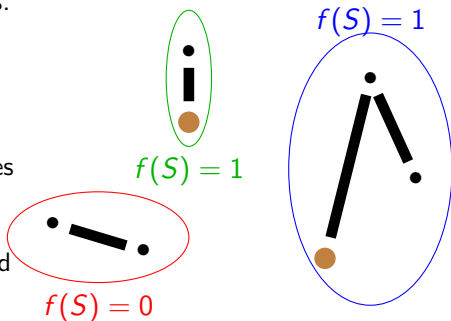
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- n components for n vertices.

- 1 **Active component** (No edge leaving it, but need one).
- 2 **Inactive component** (Does not need an edge leaving it).
- 3 **Marked component** (Need an edge leaving it, but instead pay the penalty).



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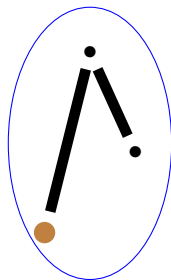
- Increase dual for active components (uniformly), until one of the following events happens.
- An edge goes tight.
 - Merge the two components.

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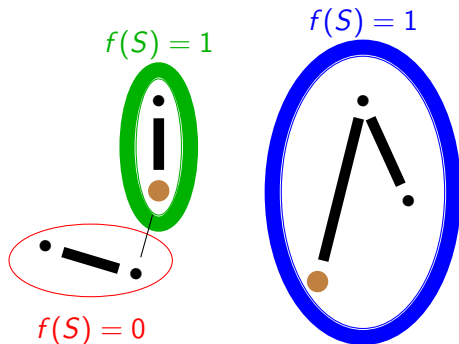


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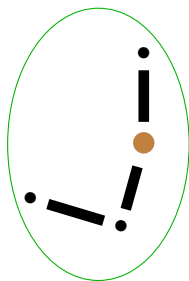
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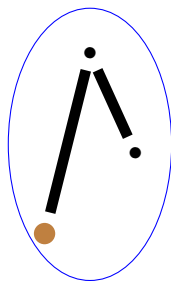
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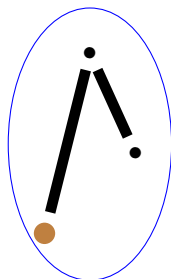
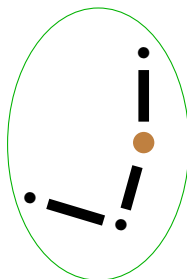
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- A family goes tight.
 - Stop increasing the dual for all participating sets, and **mark** all of them.
 - The value of dual is equal to the penalty of this family (tightness condition).

$$\pi(\{\text{Green}, \text{Blue}\}) = 8$$

$$y_S = 3$$

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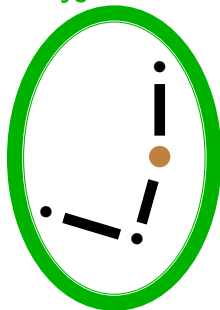
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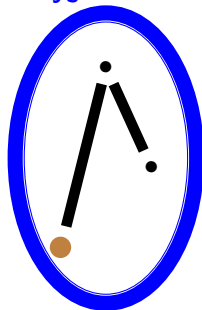
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- Stop when no active component remains.
- Penalty for all marked subsets has been paid for.
- Reverse delete step
 - 1 Consider edge in the reverse order of what they were added to the solution.
 - 2 Consider deleting an edge. Has the penalty of two resulting components already been paid for? If yes, delete the edge, else retain it.

The idea behind the proof

- The penalty paid by the algorithm is at most OPT.
 - The family of all marked sets is tight (Union of two tight families is also tight).
 - Penalty of the solution is at most

$$\pi(\text{Marked Sets}) = \sum_{S \in \text{Marked}} y_S \leq \text{OPT}.$$

- The cost of edges is bounded by twice the dual objective.
 - The idea here is to use the fact that the **average degree** of a tree is at most 2.
 - This follows the proof in [Goemans and Williamson '95], with some care because of the more general situation.

Implementation: Finding a tight family

- A family is a collection of subsets of V .
- There are doubly exponentially many families.
- Which family goes tight next (in the dual increase phase)?
 - Reduce the number of candidate families to $2^{O(n)}$ (from 2^{2^n}) and use submodular function minimization.

Claim

If a set S has $\hat{y}_S = 0$ and is not active, then it will **not** be in the next tight family.

Finding a tight family

- There are only $3n$ sets which
 - have positive dual or
 - are active.
- We only need to consider families which contain just these sets.

Maximize ε

subject to

$$\sum_{S \in \mathcal{F}} \hat{y}_S + \varepsilon \cdot (\text{Number of active sets in } \mathcal{F}) \leq \pi(\mathcal{F}); \quad \text{for all } \mathcal{F}.$$

\mathcal{F} here is a subfamily of subsets which either have positive dual or are active.

- This is still an exponential-size LP.

How to solve the program?

Maximize ε

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$$\sum_{S \in \mathcal{F}} \hat{y}_S + \varepsilon \cdot (\text{Number of active sets in } \mathcal{F}) \leq \pi(\mathcal{F}); \quad \text{for all } \mathcal{F}.$$

- Use binary search to find the correct value of ε .
- The interval can be halved using submodular function minimization.
 - Ground elements: active subsets and subsets having positive dual.
 - Once the interval becomes sufficiently small, use rational approximation to find the optimal value.

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- Challenges.
 - What do we mean by solving the linear program? There are 2^n constraints and 2^{2^n} variables.
 - [Hajiaghayi and Jain] had an underlying compact formulation.

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 - Given the edge values, the optimum values corresponding to families can be computed by a convex program.
 - Compact, convex programming formulation of the problem (in x_e variables).

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- Details in the paper. . .

Future directions

- Are there other problems that fit our framework?
- Is there a 2-approximation for PCF?
 - [Hayrapetyan, Swamy, Tardos] had a 2-approximation for a special case.
- Is there a general enough subclass of PCF which admits a 2-approximation?

Thank You for your time and attention

Comments/Questions?