

A Matter of Degree:

Improved Approximation Algorithms for Degree-Bounded Minimum Spanning Trees

J. Könemann

(joint work with R. Ravi)



Talk Outline

- The Problem
- Minimum-Degree MST's
- A Lagrangean Formulation
- Our Algorithm
- The Analysis



Degree-Bounded MST's

Given an undirected graph $G=(V,E)$, a non-negative cost function

$$c : E \rightarrow \mathcal{R}^+$$

and a parameter B .

Let $d_T(v)$ be the degree of node v and

$\Delta(T)$ be the maximum node degree in T

(BMST)

Find spanning tree T of G

with $\Delta(T) \leq B$

such that $c(T)$ is minimized



Our Result

Theorem: Given $G=(V,E)$ and positive parameter B , we compute T with

1. $c(T) \leq (1 + \frac{1}{\omega}) \cdot \text{opt}$
2. $\Delta(T) \leq r(1 + \omega)B + \log_r n$

where $r > 1$, $\omega > 0$ and opt is the cost of the optimum degree- B -bounded MST.

e.g.: $\omega=1$ and $r=2$ yields T with

$$c(T) \leq 2 \cdot \text{opt} \quad \text{and} \quad \Delta(T) \leq 4B + \log_2 n$$



Previous Work

[Ravi et. al., 93] show how to compute spanning tree T with

1. $c(T) \leq O(\log n) \cdot opt$
2. $\Delta(T) \leq O(\log n) \cdot B$

The authors **extend their results** to

- Steiner trees and generalized Steiner forests
- Node-weighted case



Talk Outline

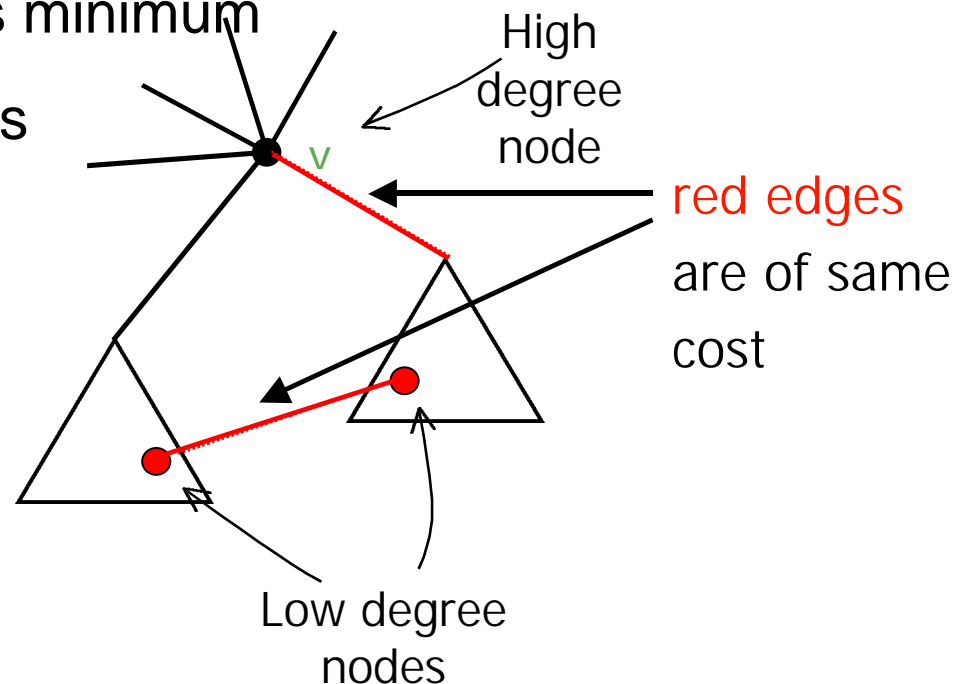
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Minimum-Degree MST's

Problem: Given an undirected graph $G=(V,E)$ and a non-negative **cost function** c on the edges.

Find a **min-cost spanning tree** T such that its **maximum degree** $\Delta(T)$ is minimum

Idea: locally optimal trees



Call this an **Improvement** for v



Minimum-Degree MST's

Definition[Fürer, Raghavachari]

A tree T is called **pseudo optimal** if no improvement is applicable to any node v with

$$d_T(v) \geq \Delta(T) - \log |V|$$

Denote **application** of this procedure to tree T by $P_{\text{local}}(T)$



Fractional Trees

Let T^a be a **convex combination** of minimum-cost spanning trees for cost function c

$$T^a = \sum_{\substack{\text{min-cost} \\ \text{spanning tree } T}} \mathbf{a}_T \cdot T \quad \leftarrow \begin{array}{l} \text{incidence vector} \\ \text{of tree } T \end{array}$$

Let the **fractional degree** of T^a at v be

$$\mathbf{d}_c^a(v) = \sum_{\substack{\text{min-cost} \\ \text{spanning tree } T}} \mathbf{a}_T \mathbf{d}_T(v)$$

Define **minimum maximum** fractional degree as

$$\Delta_c^* = \min_{\text{convex comb. } \mathbf{a}} \max_v \mathbf{d}_c^a(v)$$



A Key Lemma

Modification of theorem by [Fischer, Fürer, Raghavachari] due to Éva Tardos

Key Degree Lemma A **pseudo-optimal** min-cost spanning tree T can be computed in **poly-time** and

$$\Delta(T) \leq r\Delta_c^* + \lceil \log_r n \rceil$$

where $r > 1$.



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A Lagrangean Formulation

IP formulation of the our problem

$$\begin{aligned} \text{opt} = \min & \quad c(T) \\ \text{s.t.} & \quad \text{spanning tree } T \end{aligned}$$

dualize \longrightarrow $\mathbf{d}_T(v) \leq B \quad \forall v \in V$

Lagrangean relaxation

$$(LR(B)) \quad z_{LR} = \max_{I \geq 0} \min_{\text{spanning tree } T} c(T) + \sum_{v \in V} I_v (d_T(v) - B)$$

Fact: (weak duality) $z_{LR} \leq \text{opt}$




A few Observations

We can rewrite objective function of Lagrangean

$$\sum_{uv \in T} c_{uv} + \sum_{v \in V} \mathbf{l}_v (\mathbf{d}_T(v) - B)$$

Think of \mathbf{l}_v is being added to each edge incident to v

$$\sum_{uv \in T} (c_{uv} + \mathbf{l}_u + \mathbf{l}_v) - B \sum_{v \in V} \mathbf{l}_v$$


c_{uv}^l



More Observations

$$(LR(B)) \quad \max_{I \geq 0} \quad \min_{\text{spanning tree } T} \quad c^I(T) - B \sum_v I_v$$

Constant

Inner minimum is just an **MST** computation! We denote this by $MST(c^I)$.

Theorem: Solution to $(LR(B))$ can be computed in **poly-time**.




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Our Algorithm

Given: graph $G=(V,E)$, $c : E \rightarrow \mathcal{R}^+$
and $B > 0$

1. $\lambda = \text{Solve}(\text{LR}((1+\omega)B))$
 2. $T^I = \text{MST}(c^I)$
 3. $T = \text{Plocal}(T^I)$
 4. Output T
- notice **weakened**
degree constraints
- 



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Lagrangean Properties

Recall

$$(LR((1+w)B)) \quad \max_{I \geq 0} \quad \min_{\text{spanning tree } T} \quad c^I(T) - (1+w)B \sum_v I_v$$

We can formulate this as an LP !

$$(P) \quad \begin{array}{ll} \max & \mathbf{h} \\ \text{s.t.} & \mathbf{h} \leq c^I(T) - (1+w)B \sum_{v \in V} I_v \quad \forall \text{sp-trees } T \\ & I \geq 0 \end{array}$$

\mathbf{h}



LP formulation

$$\begin{aligned} & \max \quad \mathbf{h} \\ \text{(P)} \quad & \text{s.t.} \quad \mathbf{h} \leq c^l(T) - (1+w)B \sum_{v \in V} l_v \quad \forall \text{sp-trees } T \\ & \quad \mathbf{l} \geq 0 \end{aligned}$$

$$\begin{aligned} & \min \quad c(T^{\mathbf{a}}) \\ \text{(D)} \quad & \text{s.t.} \quad \mathbf{d}^{\mathbf{a}}(v) \leq (1+w)B \quad \forall v \in V \\ & \quad \mathbf{a} \text{ convex comb.} \end{aligned}$$



Lagrangian Properties

Proposition: Let λ be an optimum solution to

$(LR((1+w)B))$ then there is a convex combination

$$T^a = \sum_{\substack{\text{min-cost spanning} \\ \text{tree } T \text{ for } c^l}} a_T T$$

such that 1. $\forall v \in V : d_{c^l}^a(v) \leq (1+w)B$

2. $I_v > 0 \Rightarrow d_{c^l}^a(v) = (1+w)B$

Proof: complementary slackness.

Corollary [Tardos] $\Delta_{c^l}^* \leq (1+w)B$



Degree of output tree

Step 3 of our algorithm applies **Plocal** to tree $T^I = \text{MST}(c^I)$

Final tree T has **degree** at most

$$r\Delta_{c^I}^* + \lceil \log_r n \rceil$$

by **key degree lemma**

...and

$$\Delta_{c^I}^* \leq (1 + \mathbf{w})B$$

by the last **Corollary**.



Proving low Cost

min. cost of any degree- B -bounded spanning tree

Theorem: $c(T) \leq (1 + 1/w) \cdot \text{opt}$

Proof-Sketch:

$$\begin{aligned} c(T) &= c(T) + \sum_v \mathbf{1}_v(d_T(v) - B) + \sum_v \mathbf{1}_v(B - d_T(v)) \\ &\leq \text{opt} + \sum_v \mathbf{1}_v(B - d_T(v)) \end{aligned}$$

Now bound $B \sum_v \mathbf{1}_v$ by $\frac{\text{opt}}{w}$.

Uses the fact that λ is optimum for $(1 + \omega)B$ instead of just B critically.



Open Questions and Conclusion

Can the presented framework be **generalized**?

Can the result be extended to **Steiner networks**?

What about **individual node degrees**?