

# Minimum-Cost Bounded-Degree Spanning Tree

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# The Problem: Finding Optimal Spanning Trees

## Objective

Given a weighted undirected connected graph, find a minimum-cost spanning tree where each node's degree is within a specified bound.

## Research Goals

Remove reliance on time-consuming Linear Programming and adapt combinatorial local search techniques.

# Local Search for Unweighted Graphs

## Minimum-Degree Spanning Tree Problem

Find a spanning tree  $T$  that minimizes the maximum degree of any vertex  $v$  in  $T$ .

## Computational Complexity

- **NP-hard:** The problem is computationally intensive.
- **Hamiltonian Path:** A spanning tree with max degree 2 is a Hamiltonian path, which is NP-complete to decide.

Exact solutions are infeasible, driving the need for approximation algorithms and heuristics.

# Designing the Local Search Algorithm

A polynomial-time local search algorithm finds a tree  $T$  with a maximum degree at most  $2 \cdot \text{OPT} + \lceil \log_2 n \rceil$ .

## Local Move Definition

Identify edges not in  $T$  that create a cycle. Select an edge  $(v, w)$  where  $\max(d_T(v), d_T(w)) \leq d_T(u) - 2$ .

## Move Properties

After a move,  $d_{T'}(u) = d_T(u) - 1$ . The algorithm targets high-degree nodes and terminates when no further moves are possible.

# OPT + 1 Local Search Framework

This framework operates in phases and subphases to reduce node degrees.

## Phase-Based Execution

Algorithm runs in phases, each removing all degree- $k$  nodes. Each phase has subphases targeting single degree- $k$  nodes.

## Subphase Actions

Identify reducible degree- $(k-1)$  nodes, mark them, and update graph components. This enables local moves to reduce node degrees.

# Linear Programming Formulation

We use an LP formulation for a graph  $G = (V, E)$  with edge costs  $c_e \geq 0$  and degree bounds  $b_v \geq 1$  for constrained vertices  $W \subseteq V$ .

$$\text{Minimize: } \sum_{e \in E} c_e x_e$$

Subject to:

- $\sum_{e \in E} x_e = |V| - 1$  (spanning tree)
- $\sum_{e \in E(S)} x_e \leq |S| - 1, \forall S \subseteq V, |S| \geq 2$  (acyclicity)
- $\sum_{e \in \delta(v)} x_e \leq b_v, \forall v \in W$  (degree bounds)
- $x_e \geq 0, \forall e \in E$  (non-negativity)

# LP-Based Deterministic Rounding

An iterative process to derive a spanning tree from the LP relaxation.

## Solve LP Relaxation

Obtain a basic optimal solution  $x$  and identify its support  $E(x) = \{e : x_e > 0\}$ .

## Iterative Reduction

Repeatedly remove zero-edges, add single incident edges to solution, and remove vertices from  $W$  under specific conditions.

This returns a spanning tree  $F$  with cost  $\leq$  LP optimal value, and  $\deg_F(v) \leq b_v + 2$  for each  $v \in W$ .

# Refinement to +1 Violation Bound

An improved algorithm guarantees a tighter bound on degree violations.

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## Improved Algorithm

Repeatedly solve LP relaxation and remove vertices  $v$  from  $W$  that are guaranteed to have  $\leq b_v + 1$  support edges.

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## Resulting Tree

When  $W$  is empty, compute an ordinary minimum-cost spanning tree. This yields  $\deg_F(v) \leq b_v + 1$ .

# Flow-Based Constraint Reduction

## Motivation

Replace the exponential number of cycle constraints with a polynomial-sized flow formulation.

### Extended Variable Set

- Original edge variables:  $x_e$
- Flow variables:  $f_{uv}^a \geq 0$  for each arc  $(u, v)$  and commodity  $a \in V \setminus \{r\}$ .

### Flow Constraints

- Capacity constraints:
- $$\forall \{u, v\} \in E, \forall a \in V \setminus \{r\} : f_{uv}^a \leq x_{uv}, f_{vu}^a \leq x_{uv}.$$
- Per-commodity supply/demand ensures flow from source  $a$  to sink  $r$ .

# Feedback Vertex Set (FVS)

- A set  $F \subseteq V$  is an FVS if removing  $F$  makes the graph acyclic (every cycle intersects  $F$ ).
- Goal: find  $F$  of minimum total cost  $c(F) = \sum_{v \in F} c(v)$ .
- Problem is NP-hard, so we use approximation algorithms.

# Local Ratio for FVS: How It Works

- **Compute  $\alpha$ :**  $\alpha = \min_{v \in V} c(v)/(d(v) - 1)$ , where  $d(v)$  is the degree of  $v$  (assume  $d(v) \geq 2$ ).
- **Reduce costs:** For every vertex set:  $c'(v) = c(v) - \alpha(d(v) - 1)$ . (By choice of  $\alpha$ , at least one  $v$  has  $c'(v) = 0$ .)
- **Select & remove:** Let  $F_0 = \{v \in V : c'(v) = 0\}$ . Add  $F_0$  to the solution and remove them from the graph:  $G' \leftarrow G \setminus F_0$ .
- **Recurse:** Run the same procedure on  $(G', c'|_{V(G')})$  to get  $F'$ .
- **Combine:** The final FVS is  $F = F_0 \cup F'$ .

# Bounded-Degree Spanning Tree

- Each vertex  $v$  has a degree bound  $b_v$ .
- In any spanning tree  $T$ , define the exceedance:  $e_T(v) = d_T(v) - b_v$ . (How much the degree of  $v$  exceeds its allowed limit.)
- The quality of a tree is measured by its maximum exceedance:  $\max_{v \in V} e_T(v)$ .
- Goal: Find a spanning tree  $T$  that minimizes this maximum exceedance:  $\min_T \max_v e_T(v)$ .

# Local Search Framework

- For the current tree  $T$ , compute the current worst exceedance:  $k = \max_v e_T(v)$ .
- Define the critical vertex sets:  $D_k = \{v : e_T(v) = k\}$ ,  $D_{k-1} = \{v : e_T(v) = k - 1\}$ . These are the vertices closest to violating the bounds the most.
- The algorithm searches for a local move (edge swap) that:
  - reduces the degree of at least one vertex in  $D_k$ ,
  - while maintaining a valid spanning tree.

# Lower Bound on the Optimal Exceedance

- Suppose at level  $k$  no local improving move exists.
- Then every spanning tree  $T'$  must have:  $\max_v e_{T'}(v) \geq k - 1$ .
- Meaning: Even the optimal spanning tree cannot reduce the worst exceedance below  $k - 1$ .
- Therefore:  $k \leq \text{OPT}_{\text{exc}} + 1$ .
- This gives the algorithm a +1 approximation guarantee.