Delegate and Conquer: An LP-Based Approximation Algorithm for Minimum Degree MSTs*

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Abstract. In this paper, we study the minimum degree minimum spanning tree problem: Given a graph G = (V, E) and a non-negative cost function c on the edges, the objective is to find a minimum cost spanning tree T under the cost function c such that the maximum degree of any node in T is minimized.

We obtain an algorithm which returns an MST of maximum degree at most $\Delta^* + k$ where Δ^* is the minimum maximum degree of any MST and k is the distinct number of costs in any MST of G. We use a lower bound given by a linear programming relaxation to the problem and strengthen known graph-theoretic results on minimum degree subgraphs [3,5] to prove our result. Previous results for the problem [1,4] used a combinatorial lower bound which is weaker than the LP bound we use.

1 Introduction

The minimum spanning tree problem is a fundamental problem in combinatorial optimization. It also has various applications, especially in network design. A favorable property of a connecting network is not only to have the lowest possible cost but also to have small load on all nodes. A natural way to formulate this problem is via the minimum degree minimum spanning tree (MDMST) problem. In an instance of the MDMST problem, we are given a graph G=(V,E) and a non-negative cost function c on the edges, and the objective is to find a minimum cost spanning tree T under the cost function c such that the maximum degree of T is minimized. Here, the maximum degree of T is the maximum degree among all vertices in T.

The MDMST problem is closely related to the Hamiltonian path problem. If the maximum degree of an MST in an unweighted graph is at most 2, we get a Hamiltonian path. Since we do not assume that the costs are metric, no approximation is possible unless we relax the degree constraints [6]. Hence, for the MDMST problem, the natural criterion for approximation is the maximum degree of the minimum spanning tree.

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1.1 Previous Work

For the MDMST problem, Fischer [4] gave a polynomial time algorithm which returns a minimum spanning tree with maximum degree $b\Delta^* + \log_b n$ for any b > 1 where Δ^* is the maximum degree of the optimal MST based on the techniques on Furer and Raghavachari [5]. A generalization of the MDMST problem is the bounded degree minimum spanning tree problem (BDMST) in which one is given degree bounds (B_v for vertex v) in an undirected graph with edge costs c and we demand a minimum cost tree satisfying the degree bounds. The BDMST problem is closely related to the well-studied Travelling Salesman Problem [8]. In particular, if we set $B_v = 2$ for each vertex v, the BDMST problem reduces to the Travelling Salesman Path Problem which has been studied by Lam and Newman [11].

For the BDMST problem, Konemann and Ravi [9,10] gave bi-criteria approximation algorithms which return a spanning tree with $O(B_v + \log n)$ bound on the degree of vertex v and cost $O(c_{opt})$. Here n is the number of vertices in the input graph and c_{opt} is the minimum cost of a spanning tree obeying the degree bounds. Chaudhuri et al [1,2] gave a quasi-polynomial time algorithm for the MDMST problem which returns a tree of maximum degree $O(\Delta^* + \frac{\log n}{\log \log n})$ and a polynomial time algorithm that returns a tree of maximum degree $O(\Delta^*)$. They also generalize both their algorithm for the BDMST problem giving algorithms with similar bounds on the degree as in the MDMST problem and cost $O(c_{OPT})$. All these results [9,1,2] for the BDMST problem are derived from results for the MDMST problem [4,1,2], thus motivating us to concentrate on the latter. Subsequent to our work, Goemans [7] has shown an algorithm for the BDMST problem which returns a tree of optimal cost and degree of vertex v at most $B_v + 2$ for each $v \in V$.

An interesting restriction of the MDMST problem arises when all costs are in $\{1,\infty\}$. Then, as every spanning tree of cost 1 edges is an MST, the MDMST problem reduces to finding a spanning tree in the undirected graph induced by the cost one edges with minimum maximum degree. Fürer and Raghavachari [5] gave an algorithm which returns a tree with maximum degree within $\Delta^* + 1$, where Δ^* is the degree of the optimal tree.

1.2 Our Work and Contributions

All previous algorithms for the MDMST problem worked with a combinatorial lower bound given by a *witness set*. The major contribution in this paper is working with a *stronger* lower bound given by a natural linear programming relaxation of the problem. Also, we strengthen the existing results of Fürer and Raghavachari [5] and Ellingham and Zha [3]. This helps us prove our main theorem below. Here, the maximum degree restriction can be generalized to specify separate bounds on individual nodes.

Theorem 1. Given an instance of the minimum degree minimum spanning tree problem on a graph G = (V, E) with a cost function c on the edges and a degree bound B_v on vertex v for each $v \in V$, there exists a polynomial time algorithm

which shows either that the degree upper bounds are infeasible for any minimum spanning tree of G or returns an MST in which the degree of each vertex v is at most $B_v + k$ where k is the number of distinct costs in any MST.

Note that our Theorem 1 strictly generalizes the result of [5] since k = 1 in an unweighted graph. We introduce the following new ideas to prove Theorem 1:

- We use linear programming relaxation as a check for infeasibility instead of the witness set that has been used previously [4,1]. If the degree bounds are feasible for a *fractional MST*, we use the optimal LP solution to divide the total degree bound of a vertex v into k parts, each assigned to a set of incident edges of a particular cost.

Our strategy is to deal with edges of distinct costs separately. We use the known results for the unweighted case of the problem given by Fürer and Raghavachari [5] and its generalizations by Ellingham and Zha [3] to prove a weaker version of theorem 1 with degree guarantees of $B_v + 2k - 1$ instead of $B_v + k$ as claimed in the theorem. This we prove in Section 3.

- We strengthen the existing results of [5] in Theorem 4, by showing that when we do not find a witness for infeasibility we can obtain a solution where the degree bound is strictly satisfied for one chosen vertex while still ensuring that the violation of this bound is at most one for any other vertex. Similarly, we strengthen the results of Ellingham and Zha [3] in Theorem 5 (Section 4). We believe that these improvements are interesting in their own right.
- We use the strengthened guarantees in Theorem 4 and Theorem 5 to prove Theorem 1. We do this by applying the methods of Theorem 5 on different unweighted subgraphs, each naturally defined by edges of a particular cost that are used in an MST. This application proceeds in the top-down order by considering subgraph of progressively decreasing costs. At each step, we assign vertices to cost classes in which they can exceed their degree bound by at most one. We then inductively ensure that any such vertex does not exceed its bound in any other cost class. The resulting delegate-and-conquer algorithm is presented in Section 5, along with a proof of our main result.

2 Structure of MSTs

In this section, we prove some properties of MSTs. We then show the implication of these properties on the structure of the optimal solution to the linear programming relaxation to the MDMST problem.

2.1 Forest over Forest Problem

We define a new problem which will be used later for the MDMST problem.

Given a forest F of a graph G, we call H a F-tree of G if H does not contain any edge $e = \{u, v\}$ such that both u and v are in the same component of F and $F \cup H$ is a spanning tree over each connected component of G. Note that for any F-tree H of G, |H| = (number of connected components in F) – (number of connected components of G).

In an instance of the forest over forest problem, we are given an unweighted graph G = (V, E) a forest F with connected components $C(F) = \{C_1, \ldots, C_k\}$ and a degree bound B_v for each vertex $v \in V$. The problem is to find a F-tree H of G such that $deg_H(v) \leq B_v$.

We also define a notion of witness set which forms the basis of the algorithms of Ellingham and Zha [3] and Fürer and Raghavachari [5]. Given a set $W \subset V$ and partition \mathcal{P} of connected components of F, we say (W, \mathcal{P}) is a witness if each edge e with endpoints in different sets of \mathcal{P} must have at least one endpoint in W. The following lemma is straightforward and proved in [3].

Lemma 1. [3] If (W, \mathcal{P}) is a witness, then $\sum_{w \in W} deg_H(w) \geq |\mathcal{P}| - \kappa(G)$ for any F-tree H of G, where $\kappa(G)$ is the number of connected components of G.

The following theorem was proved by Ellingham and Zha [3] for the forest over forest problem.

Theorem 2. [3] There exists a polynomial time algorithm which given an instance of the forest over forest problem over a graph G = (V, E) and a forest F with degree bound B_v for each vertex $v \in V$, returns a F-tree H and a witness (W, P) such that:

- 1. If $W \neq \phi$, then the witness (W, \mathcal{P}) shows that $\sum_{w \in W} deg_{H'}(w) \geq (\sum_{w \in W} B_w) + 1$ for each F-tree H' of G, i.e., the degree bounds are infeasible for any F-tree of G.
- 2. If $W = \phi$, then $deg_H(v) \leq B_v + 1$ for each $v \in V$.

The MDST problem (unweighted MDMST problem) is a special case when $F = \phi$. Then the problem reduces to finding a spanning tree of G and the guarantees of the above theorem are exactly the same as those of Fürer and Raghavachari [5].

2.2 Laminar Structure of an MST

Given a graph G = (V, E) with cost function c on the edges, let the cost function c take at most k different values on the edges of the MST. Without loss of generality, we can delete all edges of G of other costs since they do not occur in any MST. We also assume, without loss of generality, that the range of c is $\{1, \ldots, k\}$ as the particular values do not change the structure of any MST.

Let $G^{\leq i}$ denote the graph over V(G) with only those edges of E(G) that cost at most i. We let $G^{\leq 0}$ denote the graph over vertex set V(G) with no edges. Let G^i denote the graph with vertex set V(G) and edges in E(G) which cost exactly i. The following lemma is a standard result about minimum spanning trees.

Lemma 2. T is a minimum spanning tree of a graph G iff T^i is a $G^{\leq i-1}$ -tree of $G^{\leq i}$ for each i.

Hence, we also delete all edges e of cost i or higher which have both endpoint of $G^{\leq i-1}$ without affecting any MST T of G. More importantly, Lemma 2 implies that we can independently select the edges of each cost class one at a time and solve the appropriate unweighted forest over forest problem to form an MST. The main issue is to manage the degree of any vertex across different cost classes.

2.3 LP Relaxation

We formulate the following integer program MST_{IP} for the problem considered in Theorem 4 which is a generalization of the MDMST problem.

$$opt_B = min \quad \sum_{e \in E} c_e x_e \tag{1}$$

$$s.t. \ x(\delta(v)) \le B_v \ \forall v \in V, \tag{2}$$

$$x \in SP_G,$$
 (3)

$$x integer.$$
 (4)

Here SP_G is the spanning tree polyhedron, i.e., a linear description of the convex hull of all spanning trees of G. It is well known that optimization over SP_G can be achieved in polynomial time [12]. We then relax the integrality conditions to obtain MST_{LP} . If the optimum value of MST_{LP} is more than the cost of an MST of G, then clearly the problem is infeasible.

Let x^* denote an optimal basic feasible solution to MST_{LP} . The following lemma follows directly from LP duality and is implicit in [9].

Lemma 3. [9] The optimal basic feasible solution x^* to MST_{LP} can be written as a convex combination of spanning trees, i.e, there exists spanning trees T_0, \ldots, T_n and constants $\lambda_0, \ldots, \lambda_n$ such that $x^* = \sum_{i=0}^n \lambda_i T_i$, $\sum_{i=0}^n \lambda_i = 1$ and $\lambda_i > 0$ for each $0 \le i \le n$. Here n is the number of vertices in the graph G.

The following corollary to Lemma 3 is straightforward.

Corollary 1. If $c(x^*) = c_{MST}$ then each of the spanning trees T_0, \ldots, T_n obtained from Lemma 3 are minimum spanning trees.

Proof. As $x^* = \sum_{i=0}^n \lambda_i T_i$, we have $c_{MST} = c(x^*) = \sum_{i=0}^n \lambda_i c(T_i) \le \sum_{i=0}^n \lambda_i c_{MST} = c_{MST} \sum_{i=0}^n \lambda_i = c_{MST}$. Hence, each of the inequalities $c_{MST} \le c(T_i)$ must hold at equality.

2.4 LP Relaxation for Forest over Forest Problem

Given a forest over forest problem of constructing a F-forest of graph G with degree bound B_v for each vertex $v \in V$, we formulate the following natural IP formulation for the forest over problem which we call the $IP_{FOR}(F,G)$. Observe that this is a feasibility problem as a forest over forest problem is over an unweighted graph.

$$opt = min 0 (5)$$

$$s.t. \ x(\delta(v)) \le B_v \ \forall v \in V, \tag{6}$$

$$x \in SF(G/F),$$
 (7)

$$x \in \{0, 1\},$$
 (8)

Here G/F denotes the graph formed when we shrink components of each component of F in to a single vertex in G and SF(G) denote the natural linear formulation for the incidence vectors of all maximal spanning forests of G generalized from [12,9] (This is the same as the formulation for the incidence vectors of the bases of the graphic matroid of G). If we relax the integrality constraints, we get a LP relaxation which we denote by $LP_{FOR}(F,G)$. Later in Lemma 5, we show that if there exists a witness showing infeasibility of the degree bounds then the above LP relaxation is also infeasible.

2.5 Decomposing MSTs into Forests over Forests

To obtain any minimum spanning tree on an edge-weighted graph G, we need to solve the $LP_{FOR}(G^{\leq i}, G^{\leq i+1})$ for each i = 0, ..., k-1 where k is the number of distinct edge-costs in any minimum spanning tree of G. Now, we show that MST_{LP} actually solves each of these forest over forest problems with appropriate degree bounds.

Let $B_v^i = \sum_{e \in \delta(v), c(e)=i} x_e^*$. Observe that $\sum_{i=1}^k B_v^i \leq B_v$. Let $y_e^i = x_e^*$ if c(e) = i else $y_e^i = 0$ for each $i = 1, \ldots, k$ and $e \in E$. Then we have the following lemma.

Lemma 4. For each $1 \le i \le k$, y^i is a feasible solution to the linear programming relaxation of the forest over forest problem of finding a $G^{\le i-1}$ -tree of $G^{\le i}$ with degree bound B^i_v for each vertex $v \in V$.

Proof. Let $x^* = \sum_{j=0}^n \lambda_j T_j$ as in Lemma 3. Let H^i_j be the forest formed by cost-i edges in tree T_j . Clearly, each of the forests H^i_j is a valid $G^{\leq i-1}$ -tree of $G^{\leq i}$ but may violate the degree bounds. By definition, $y^i = \sum_{j=0}^n \lambda_j H^i_j$ and hence is a valid fractional solution to $LP_{FOR}(G^{\leq i-1}, G^{\leq i})$. Also, it satisfies the degree constraints by definition as $\sum_{e \in \delta(v)} y^i(e) = \sum_{e \in \delta(v), c(e)=i} x_e = B^i_v$. \square

The following lemma shows that the LP gives a stronger notion of infeasibility than any witness.

Lemma 5. If there exists a witness (W, \mathcal{P}) showing that the degree bounds are infeasible then the $LP_{FOR}(F, G)$ is infeasible.

Proof. If (W, \mathcal{P}) is a witness showing that the degree bounds are infeasible then for any F-tree H, $\sum_{v \in W} deg_H(v) \ge \sum_{v \in W} B_v + 1$. Hence, the above holds for F-trees H_0, \ldots, H_n . For any convex combination, we get

$$\sum_{v \in W} \sum_{i=0}^{n} \alpha_i deg_{H^i}(v) \ge \sum_{i=0}^{n} \alpha_i (\sum_{w \in W} B_w + 1) \ge \sum_{v \in W} (\sum_{i=0}^{n} \alpha_i B_v) + 1 = \sum_{v \in W} B_v + 1$$

since $\sum_{i=0}^{n} \alpha_i = 1$. Since any feasible solution to $LP_{FOR}(F, G)$ dominates a convex combination of F-trees (variant of Lemma 3), the degree constraint for at least one node in W must be violated.

3 Weaker Algorithm

As a warm-up, we describe an algorithm which uses the linear programming relaxation for the MDMST problem and the algorithm of Ellingham and Zha [3] as stated in Theorem 2 to obtain a weaker guarantee than claimed in Theorem 1.

Given an instance of MDMST problem over G = (V, E), cost function c and degree bound B_v on vertex v, the algorithm Alg-Weak is as follows:

- 1. Find x^* the optimum solution to the linear programming relaxation to the problem. If $c(x^*) > c_{MST}$, declare the problem infeasible.
- 2. Define $B_v^i = \sum_{e \in \delta(v), c(e)=i} x_e^*$. For each i, we construct $G^{\leq i-1}$ -tree H^i of $G^{\leq i}$ with degree bounds $\lceil B_v^i \rceil$ for each vertex $v \in V$ using the algorithm described in Theorem 2.
- 3. Return the MST $T = \bigcup_i H^i$.

Theorem 3. Algorithm Alg-Weak for the MDMST problem on graph G with a degree bound B_v on vertex v for each $v \in V$ and a cost function c returns an MST such that the degree of any vertex v is at most $B_v + 2k - 1$ or shows that the degree bounds are infeasible for any MST of G. Here, k is the number of distinct edge costs in any MST.

Proof. The algorithm declares the degree bounds infeasible only if $c(x^*) > c_{MST}$. Clearly, then the degree bounds are infeasible for any MST. We only need to argue that if the $c(x^*) = c_{MST}$, then the tree returned satisfies the claimed degree bounds and is an MST.

First, observe the tree T returned is an MST as it is a union of $G^{\leq i-1}$ -tree H^i of $G^{\leq i}$ for each i (see Lemma 2).

We only need to show that the algorithm in Theorem 2 returns a F-tree and not a witness set showing infeasibility of the degree bounds. However, this directly follows from Lemma 5.

Observe that the degree of any vertex v in tree T is exactly $deg_T(v) = \sum_{i=1}^k deg_{H^i}(v) \leq \sum_{i=1}^k (\lceil B_v^i \rceil + 1) < \sum_{i=1}^k (B_v^i + 2) \leq B_v + 2k$. Here the last inequality follows from the fact that $\sum_{i=1}^k B_v^i \leq B_v$. which proves the degree bound as claimed. This proves Theorem 3.

4 A Refined Characterization of Witnesses

In this section, we strengthen Theorem 2 which will help us obtain improved guarantees for the MDMST problem. However, to illustrate the strengthening without getting mired in notation, we first state and prove the strengthening of the result of Fürer and Raghavachari [5] for spanning trees rather than for forests over forests. Recall however that Theorem 2 is a generalization of the result of Fürer and Raghavachari [5], so the ideas in this strengthening generalize with some extra work allowing us to prove Theorem 5 in the the spirit of Theorem 4 (that is described in extended version of this paper [13]).

4.1 Improving Unweighted Minimum Degree Spanning Trees

Fürer and Raghavachari [5] present an algorithm which returns a tree of maximum degree $\Delta^* + 1$ where Δ^* is the minimum maximum degree of any tree.

We prove a stronger version of their theorem which is useful for the weighted version of the problem. The algorithm is similar to the algorithm of Fürer and Raghavachari [5] but our stopping criterion is more stringent.

Given a tree T and an edge $f \notin T$, let Cycle(T, f) denote the set of vertices on the unique cycle in $T \cup f$.

Theorem 4. Given a connected graph G = (V, E), degree bound B_v for each vertex $v \in V$, there is a polynomial time algorithm which returns a spanning tree T and witness set $W \subset V$ (possibly empty) such that

- 1. Infeasibility: If $W \neq \phi$, then for any tree T', $\sum_{w \in W} \deg_{T'}(w) \geq \sum_{w \in W} B_w + 1$, i.e., the degree bounds are infeasible for any spanning tree of G.
- 2. Solution: If $W = \phi$, then for each node $v \in V$, $deg_T(v) \leq B_v + 1$.
- 3. Strong Solution: If $W = \phi$, then for each node in $v \in V$, there exists a tree T_v such that $deg_{T_v}(v) \leq B_v$ and for each $u \in V \setminus \{v\}$, $deg_{T_v}(u) \leq B_u + 1$.

While the algorithm of Fürer and Raghavachari [5] results in a tree satisfying conditions 1 and 2 only, we continue to improve the solution until we satisfy condition 3 or find a new witness for infeasibility.

Algorithm Alg-Unweighted

- 1. Find any spanning tree T.
- 2. Initialize $Ugly(T) = \{v | deg_T(v) \geq B_v + 2\}$, $Bad(T) = \{v | deg_T(v) = B_v + 1\}$, $Good(T) = \{v | deg_T(v) \leq B_v\}$, MakeGood(u) = (u) for each $u \in Good(T)$. Return (T, ϕ) if $Bad(T) \cup Ugly(T) = \phi$.
- 3. If there exists edges $e = (u_1, u_2) \in T$ and $f = (v_1, v_2) \in E \setminus T$, such that e and f are swappable (i.e., e lies in the cycle closed by f in T), $v_1, v_2 \in Good(T)$ and either u_1 or $u_2 \notin Good(T)$, then do for each $w \in Cycle(T, f) \cap (Ugly(T) \cup Bad(T))$:
 - (a) $Good(T) \leftarrow Good(T) \cup \{w\}$
 - (b) $Ugly(T) \leftarrow Ugly(T) \setminus \{w\}, Bad(T) \leftarrow Bad(T) \setminus \{w\}.$
 - (c) $Makegood(w) \leftarrow (v_1, v_2)$.
- 4. If any w is shifted from Ugly(T) to Good(T) in Step 3, then $T \leftarrow Improve(w, T)$ and Return to Step 2.
- 5. Return $(T, W = Ugly(T) \cup Bad(T))$.

The procedure Improve(w,T) is implemented as follows:

- 1. If MakeGood(w) = w, then return T.
- 2. If MakeGood(w) = (u, v), let T_u and T_v be the subtree containing u and v in $T \setminus W$ where $W = Bad(T) \cup Ugly(T)$. Here, Bad(T) and Ugly(T) are as defined by the algorithm before w is shifted to Good(T). Let $T'_u = Improve(u, T_u)$ and $T'_v = Improve(v, T_v)$. Return $T' = T \cup T'_u \cup T'_v \cup \{u, v\} \setminus (T_u \cup T_v \cup e)$ where $e \in Cycle(\{u, v\}, T)$ and is incident at w.

The procedure Improve(w) ensures that the degree of one ugly vertex reduces by at least 1 while no new ugly vertices are introduced in the resulting swaps. This is ensured by the following Lemma from [5]. A vertex v is called *non-blocking* in T if $deg_T(v) \leq B_v$.

Lemma 6. [5] Suppose that $w \in Bad$ is marked Good in iteration i, when edge (u, v) is scanned in Step 3c of the algorithm. Then w can be made non-blocking by applying improvements to the components of F_i containing u and v where F_i is the subgraph of T generated by nodes marked good in iteration i.

Now, we prove Theorem 4.

Proof. Suppose $W \neq \phi$. Let C_1, \ldots, C_r be the components formed after removing W from T. Clearly, there does not exist any edge from C_i to C_j for any i,j else we would have found it in Step 3. Also number of components is at least $r \geq \sum_{w \in W} deg_T(w) - 2(|W|-1) \geq \sum_{w \in W} (B_w+1) - 2(|W|-1)$, since $deg_T(w) \geq B_w+1$ for each $w \in Bad(T)$ and $deg_T(w) > B_w+1$ for each $w \in Ugly(T)$. Let $W = \{w_1, w_2, \ldots, w_p\}$. Then, $(W, \mathcal{P} = \{C_1, \ldots, C_r\{w_1\}, \ldots, \{w_p\}\})$ is a witness as there is no swap edge wrt to W. Hence by Lemma 1 since G is connected, there must be at least r + |W| - 1 edges incident at vertices in W in any tree T', i.e., $\sum_{w \in W} deg_{T'}(w) \geq \sum_{w \in W} B_w + |W| - 2(|W|-1) + |W|-1 = \sum_{w \in W} B_w + 1$. This proves (1) in the theorem.

Suppose now that $W = \phi$. Algorithm Alg - Unweighted returns a spanning tree T and set $W = Ugly(T) \cup Bad(T) = \phi$. Hence every vertex has been marked Good implying that for any vertex $v \in V$, $deg_T(v) \leq B_v + 1$, proving (2).

Now, we prove (3). Assume that $W = \phi$. Take any $v \in V$. Hence, $v \in Good(T)$ where T is the final tree returned by the algorithm. Either $deg_T(v) \leq B_v$ in which case $T_v = T$ suffices. Else $deg_T(v) = B_v + 1$ and v was shifted from Bad(T) to Good(T) in step 3b. Then by Lemma 6, there exist a series of swaps which do not increase the degree of any vertex $u \in V$ above $B_u + 1$ and make v non-blocking. The tree T_v obtained after performing these swaps by invoking $Improve(v, T_v)$ suffices for proving (3).

4.2 Forests over Forests Revisited

In this section, we obtain strengthening of the results of Ellingham and Zha [3] on the lines of the results in Section 4.1. We are given an instance of forest over forest problem to construct a F-tree of G satisfying the degree bounds B_v for each vertex $v \in V$. We first begin with a few definitions.

Let C(F) be set of connected components of F. We will refer these connected components as *supernodes*. For any vertex v, we will denote F_v to be the supernode containing v.

We present the following theorem in the spirit of Theorem 4.

Theorem 5. Given a graph G = (V, E), a forest F, a degree bound B_v for each vertex $v \in V$, there exists a polynomial time algorithm StrongForest which returns a F-tree H of G and witness set (W, C_W) where $W \subset V$ and C_W is a partition of C(F) such that

- 1. Infeasibility: If $W \neq \phi$, then $\sum_{w \in W} deg_{H'}(w) \geq \sum_{w \in W} B_w + 1 \ \forall \ F$ -trees H', i.e, the degree bounds are infeasible for any F-tree of G.
- 2. Solution: If $W = \phi$, then for each $v \in V$, $deg_H(v) \leq B_v + 1$ and in each supernode $F_i \in C(F)$ there is at most one vertex for which the above inequality is satisfied at equality.
- 3. Strong Solution: If $W = \phi$ then for each supernode $F_i \in C(F)$ there exists a F-tree H_i which satisfies the condition (2) above. Moreover, for each vertex $u \in F_i$ we have $\deg_{H_i}(u) \leq B_u$.

Ellingham and Zha [3] prove the above theorem with conditions 1 and a weaker version of condition 2 and we strengthen it proving condition 3. Due to space considerations we omit the algorithm and the proof of the theorem since they are very similar to that for trees, only more notationally tedious. They are included in the technical report [13]. Note that the above theorem strictly generalizes Theorem 4, by setting $F = \emptyset$.

Another point to note here is that the strong solution guarantee can be applied to each connected component of G, i.e., we can choose one supernode from each connected component of G when obtaining the strong solution. This follows from the fact the F-tree problem over each connected component of G is independent and can be treated separately. We use this fact critically later in the algorithm for the MDMST problem.

5 Delegating Vertices Using Refined Witnesses

We now describe an algorithm, which given an MDMST problem on graph G = (V, E) with cost function c and degree bounds B_v gives a better guarantee than one in Section 2. We use the algorithm StrongForest of Theorem 5 instead of the algorithm forest over forest to obtain a improved guarantee for the MDMST problem.

Algorithm Delegate-and-Conquer

1. Step 1: Initialization

Solve the LP relaxation for the MDMST problem to obtain the optimal solution x^* and if $c(x^*) > c_{MST}$, we declare the instance infeasible. Else, let $B_v^i = \sum_{e \in \delta(v), c(e) = i} x_e^*$ for each $v \in V$ and for each $i = 1, \ldots, k$. Observe that $\sum_{i=1}^k B_v^i \leq B_v$ for each $v \in V$. Observe that by Lemma 5, invoking the algorithm StrongForest of Theorem 5 to construct a $G^{\leq i-1}$ -tree of $G^{\leq i}$ with degree bounds $[B_v^i]$ will always result in an empty witness set.

2. Step 2: Using StrongForest

Find a $G^{\leq k-1}$ -tree H^k of $G^{\leq k}$ with degree bounds $\lceil B_v^k \rceil$ for each vertex v using the algorithm described in Theorem 5. Let $S^k = \{v | deg_{H^k}(v) = \lceil B_v^k \rceil + 1\}$. Observe that at most one vertex of any connected component of $G^{\leq k-1}$ lies in S^k . This follows from condition (2) of Theorem 5 as each connected component of $G^{\leq k-1}$ is a supernode in the forest-over-forest problem solved. Also, let $M^k = H^k$.

3. Step 3: Delegating the vertices to cost classes

For i = k - 1 down to 1, repeat

- (a) From each connected component of $G^{\leq i}$ there is at most one vertex in S^{i+1} (proved in Lemma 7). Apply algorithm StrongForest of Theorem 5 to each component G_j^i of graph $G^{\leq i}$. Apply condition (3) of Theorem 5 by selecting from each component G_j^i the supernode containing v where $v \in S^{i+1}$ to obtain $G^{\leq i-1}$ -tree M^i of $G^{\leq i}$.
- (b) Define $S^i=\{v|deg_{\cup_{r=i}^kM^r}(v)=(\sum_{j=i}^k\lceil B_v^j\rceil)+1\}$

Return $T = \bigcup_{i=1}^k M^i$.

Theorem 6. Given an instance of the MDMST problem over a graph G = (V, E), cost function c and degree bound B_v for vertex $v \in V$, Algorithm Delegate-and-Conquer returns either an MST T such that $deg_T(v) \leq B_v + k$ or shows that the degree bounds are infeasible for any MST. Here k is the number of different costs in any MST.

Proof. We declare the problem infeasible when $c(x^*) > c_{MST}$ in which case the problem is clearly infeasible. The Step 2 of the algorithm returns $G^{\leq k-1}$ -tree H^k of graph $G^{\leq k}$ satisfying the conditions of Theorem 5. First we prove the following claim.

Lemma 7. There is at most one vertex of S^i in each connected component of $G^{\leq i-1}$ for each $1 \leq i \leq k$.

Proof. The proof of the claim is by induction for i=k down to 1. Clearly, this is true for S^k from condition (2) of Theorem 5. Suppose it is true for S^{i+1} such that $2 \le i+1 \le k$. We claim that it is true for S^i . Observe that the candidate vertices for S^i are vertices in S^{i+1} or vertices which exceed their corresponding degree bound $\lceil B_v^i \rceil$ in M^i .

Take any connected component G_j^{i-1} of $G^{\leq i-1}$. If there is some vertex v in G_j^{i-1} that is in S^{i+1} , then G_j^{i-1} is chosen in Step 3(a) of the algorithm as the selected supernode. Hence, no vertex in this connected component exceeds the degree bound $\lceil B_v^i \rceil$ in M^i by condition (3) of Theorem 5. Hence, v remains the only vertex that might exceed its total degree bound in $\bigcup_{r=i}^k M^i$. Else, if the connected component of $G^{\leq i-1}$ is such that there is no vertex of S^{i+1} in it, then we introduce at most one vertex which exceeds the degree bound in M^i by condition (2) of Theorem 5. In either case there is at most one vertex of S^i in each component of $G^{\leq i-1}$. Hence, the property holds for each $1 \leq i \leq k$.

Hence, we obtain
$$deg_T(v) = \sum_{i=1}^k deg_{M^i}(v) \le (\sum_{i=1}^k \lceil B_v^i \rceil) + 1 < \sum_{i=1}^k (B_v^i + 1) + 1 = B_v + k + 1$$
. This implies that $deg_T(v) \le B_v + k$ for each $v \in V$.

Observe that in the above proof, if each B_v^i were integral, then we would have obtained that $deg_T(v) \leq B_v + 1$ as we would "save" k-1 in rounding of fractional values.

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