Matroidal Degree-Bounded Minimum Spanning Trees

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Abstract

We consider the minimum spanning tree (MST) problem under the restriction that for every vertex \( v \), the edges of the tree that are adjacent to \( v \) satisfy a given family of constraints. A famous example thereof is the classical degree-bounded MST problem, where for every vertex \( v \), a simple upper bound on the degree is imposed. Iterative rounding/relaxation algorithms became the tool of choice for degree-constrained network design problems. A cornerstone for this development was the work of Singh and Lau [18], who showed that for the degree-bounded MST problem, one can find a spanning tree violating each degree bound by at most one unit and with cost at most the cost of an optimal solution that respects the degree bounds.

However, current iterative rounding approaches face several limits when dealing with more general degree constraints. In particular, when several constraints are imposed on the edges adjacent to a vertex \( v \), as for example when a partition of the edges adjacent to \( v \) is given and only a fixed number of elements can be chosen out of each set of the partition, current approaches might violate each of the constraints by a constant, instead of violating the whole family of constraints by at most a constant number of edges. Furthermore, it is also not clear how previous iterative rounding approaches can be used for degree constraints where some edges are in a super-constant number of constraints.

We extend iterative rounding/relaxation approaches both on a conceptual level as well as aspects involving their analysis to address these limitations. Based on these extensions, we present an algorithm for the degree-constrained MST problem where for every vertex \( v \), the edges adjacent to \( v \) have to be independent in a given matroid. The algorithm returns a spanning tree of cost at most \( \text{OPT} \) such that for every vertex \( v \), it suffices to remove at most 8 edges from the spanning tree to satisfy the matroidal degree constraint at \( v \).

1 Introduction

Recently, much effort has been put on designing approximation algorithms for degree-constrained network design problems. This development was motivated by various applications as for example VLSI design, vehicle routing, and applications in communication networks [7, 3, 17]. One of the most prominent and elementary problems here, which attracted lots of attention in recent years, are degree-constrained (MST) problems.

In the most classical setting, known as the degree-bounded MST problem, the problem is to find a spanning tree \( T \subseteq E \) of minimum cost in a graph \( G = (V, E) \) under the restriction that the degree of each vertex \( v \) with respect to \( T \) is at most some given value \( B_v \). Since checking feasibility of a degree-bounded MST problem is already NP-hard, interest arose in finding low-cost spanning trees that violate the given degree constraints slightly. A long chain of papers (see [7, 12, 13, 4, 5] and references therein) led
to algorithms with various trade-offs between cost of the spanning tree and violation of the degree bounds. In recent years, important progress was achieved for the degree-bounded MST problem, which also led to a variety of new techniques. Goemans [8] showed how to find a spanning tree violating each degree constraint by at most two units, and whose cost is bounded by the cost OPT of an optimum spanning tree that satisfies the degree constraints. Enhancing the iterative rounding framework introduced by Jain [9] with a relaxation step, Singh and Lau [18] obtained a stronger version of the above result, which is essentially best possible, where degree constraints are only violated by at most one unit. They work with an LP relaxation of the problem, and iteratively drop degree constraints from the LP that cannot be violated by more than one unit in later iterations. The adapted LP is then solved again to obtain a possibly sparser basic solution that allows for further degree relaxations. Edges not used in the current optimal solution to the LP are removed from the graph, and edges that have a weight of one are fixed, while updating degree bounds accordingly. A degree bound at a vertex \( v \) is removed whenever it is at most one unit lower than the current number of edges adjacent to \( v \).

We are interested in obtaining results of similar strength for more general degree bounds. Consider for example the following type of degree constraints: for every vertex \( v \), a partition \( E_1^v, \ldots, E_w^v \) of the set \( \delta(v) \) of edges adjacent to \( v \) is given, and within each set \( E_i^v \) of the partition, only a given number of edges can be chosen. The algorithm of Singh and Lau [18] as well as the one of Goemans [8] can easily be adapted to this setting. (In particular, the algorithm of Singh and Lau was even presented in this precise setting.) However, with both of these approaches, the constraint imposed by each set \( E_i^v \) can be violated by a constant. We are interested in having at most a constant violation over all degree constraints at \( v \), i.e., for every vertex \( v \in V \), at most a constant number of edges have to be removed from the spanning tree to satisfy all constraints at \( v \). Another more general example that will be useful to illustrate limits of current methods is obtained when imposing constraints for each vertex \( v \) on a laminar family on the edges adjacent to \( v \), instead of only considering a partition.

Adapting Goemans’ algorithm to these stricter bounds on the degree violation seems to be difficult, since a crucial step of this algorithm is to cover the support \( E^* \) of a basic solution to the natural LP relaxation by a constant number of spanning trees (for the degree-bounded MST problem, Goemans showed [8] that two spanning trees suffice). This result allows for orienting the edges in \( E^* \) such that every vertex has at most a constant number of incoming arcs, at most one in each spanning tree. Dropping for every vertex all incoming arcs from its degree constraint then leads to a matroid intersection problem, whose solution violates each degree constraint by at most a constant. To be able to decompose \( E^* \) into a constant number of spanning trees, one needs to show that for any subset of the vertices \( S \subseteq V \), only a linear number (in \( |S| \)) of edges have both endpoints in \( S \). In the classical degree-bounded MST problem, this sparseness property follows from the fact that when considering only edges with both endpoints in \( S \), there are at most a linear number (in \( |S| \)) of linearly independent and tight spanning tree constraints due to combinatorial uncrossing, and only a linear number of degree constraints within \( S \). However, in more general settings as highlighted above, the number of degree constraints within \( S \) can be super-linear.

Iterative relaxation looks more promising for a possible extension to generalized degree bounds. However, current iterative rounding approaches face several limits when trying to adapt them. In particular, when dealing with the partition bounds as explained above, a simple adaptation of the relaxation rule, where for a vertex \( v \) all constraints at \( v \) would be dropped as soon as it is safe to do so due to a small support \( E^* \), risks to get stuck because there might be no vertex whose degree constraint can be relaxed. Furthermore, previous approaches (as used in [18, 2]) to show that the support is sparse fail in our setting because of a possible super-linearity of the total number of degree constraints. Additionally, previous iterative relaxation approaches crucially rely on the property that any edge is in at most a constant number of degree constraints to obtain violations that are bounded by a constant. However, this does not hold when dealing for example with degree constraints given by a laminar family.

In this paper we show how to extend iterative relaxation approaches, both from a conceptual point of
view as well as aspects involving their analysis, to tackle a wide class of MST problems with generalized
degree bounds, namely when the degree bounds for every vertex are given by a matroid. In particular, this
includes the partition bounds and the more general laminar bounds mentioned above.

Our results. We present an iterative rounding/relaxation algorithm for finding a matroidal degree-bounded
MST. The degree bounds are given as follows: for every vertex \( v \), a matroid \( M_v = (\delta(v), I_v) \) over
the ground set \( \delta(v) \) is given with independent sets denoted by \( I_v \subseteq 2^{\delta(v)} \). The problem (without relaxed degree
constraints) is to find a spanning tree \( T \) in \( G \) satisfying \( T \cap \delta(v) \in I_v \) \( \forall v \in V \), and minimizing a linear
cost function \( c : E \rightarrow \mathbb{R}_+ \). We say that a given spanning tree \( T \) violates a degree constraint \( M_v \) by at most
\( k \in \mathbb{N} \) units, if it suffices to remove at most \( k \) edges \( R \subseteq \delta(v) \cap T \) from \( T \) to satisfy the constraint \( M_v \),
i.e., \( (T \setminus R) \cap \delta(v) \in I_v \). Hence, the partition and laminar bounds mentioned above correspond to the case
where all matroids \( M_v \) are partition or laminar matroids, respectively. We show the following.

Theorem 1. There is an efficient algorithm for the matroidal degree-bounded MST problem that returns a
spanning tree of cost at most the cost of an optimal solution, and violates each degree bound by at most 8
units.

To overcome problems faced by previous iterative relaxation approaches, we enhance the iterative relax-
ation step, and exploit polyhedral structures to prove stronger sparseness results. The polytope used as a
relaxation of the matroidal degree-bounded MST asks to find a point \( x \in \mathbb{R}^E \) in the spanning tree polytope
such that for every vertex \( v \in V \), the restriction of \( x \) to \( \delta(v) \) is in the matroid polytope \( P_{M_v} \) of \( M_v \).

To be able to always find possible relaxation steps, our iterative rounding procedure tries to achieve
a somewhat weaker goal than previous approaches. The algorithm of Singh and Lau [18] relaxes degree
constraints with the goal to approach the spanning tree polytope, which is integral. In our approach, the goal
we pursue is to remove every edge \( \{u, v\} \) from at least one of the two degree constraints at \( u \) or \( v \). As soon
as no edge is part of both degree constraints at its endpoints, the problem is a matroid intersection problem,
since all degree constraints together can be described by a single matroid over the support of the current
LP solution. Thus, once we are in this situation, the current LP will be integral and no further rounding
steps are needed. Hence, in our relaxation step, we try to find a vertex \( v \) such that we can remove all edges
adjacent to \( v \) that are still in both degree constraints from the degree constraint at \( v \). Edges adjacent to \( v \) that
are only contained in the degree constraint at \( v \) will not be removed from the constraint \( M_v \). Our approach
has thus some similarities with Goemans’ method, but instead of removing right at the start every edge from
one degree constraint, we do this iteratively and hereby profit from additional sparseness that is obtained
by solving the LP relaxation after each degree adaptation step. As we will see in Section 2 the way how
we remove edges from a constraint is strictly speaking not a relaxation, and we therefore prefer to use the
term degree adaptation instead of degree relaxation. The above degree adaptation step alone shows not to
be sufficient for our approach, since one might still end up in a situation were no further degree adaptation
can be performed because the graph is too dense. To obtain greater sparsity, we use a second type of degree
adaptation, where for some vertex \( v \) we remove (almost) the full degree constraint at \( v \) if this cannot lead to
a large violation of the degree constraint at \( v \).

The main step in the analysis is to prove that it is always possible to apply at least one of two suggested
degree adaptations. A first step in this proof is to show that the support of a basic solution to the LP relaxation
is sparse. We obtain sparsity by showing that if there are \( k \in \mathbb{N} \) linearly independent and tight constraints
(with respect to the current LP solution \( x \)) of the polytope \( P_{M_v} \), then \( x(\delta(v)) \geq k \). Since summing \( x(\delta(v)) \)
over all vertices is equal to \( 2(|V| - 1) \), because \( x(E) = |V| - 1 \) as \( x \) is in the spanning tree polytope, there
are at most \( 2(|V| - 1) \) linearly independent and tight degree constraints.

The crucial part in the analysis is to show that vertices to which no further degree adaptation can be
performed do not have very low degrees in average, implying that some of the other vertices are likely to
have low degrees and therefore admit a degree constraint adaptation. To prove this property, we exploit the interplay between degree bounds and spanning tree constraints to show that any degree two node can either be treated separately and allows for reducing the problem, or implies a reduction in the maximum number of linearly independent and tight spanning tree constraints.

Related work. The study of spanning trees with degree constraints can be traced back to Fürer and Raghavachari [7], who presented an approximation algorithm for the degree-bounded Steiner Tree problem which violates each degree bound by at most one, but does not consider costs. This result generated much interest in the study of degree-bounded network design problems, leading to numerous results and new techniques in recent years for a variety of problems, including degree-bounded arborescence problems, degree-bounded $k$-edge-connected subgraphs, degree-bounded submodular flows, degree-bounded bases in matroids (see [16, 17, 11, 14, 15, 10, 2, 6, 1] and references therein).

Spanning tree problems with a somewhat different notion of generalized degree bounds have been considered in [2] and [1]. In these papers, the term “generalized degree bounds” is used as follows: given is a family of sets $E_1, \ldots, E_k \subseteq E$, and the number of edges that can be chosen out of each set $E_i$ is bounded by some given value $B_i \in \mathbb{N}$. In [2], using an iterative relaxation algorithm, whose analysis is based on a fractional token counting argument, the authors show how to efficiently obtain a spanning tree of cost at most $OPT$ and violating each degree bound by at most $\max_{e \in E} \{i \in [k] \mid e \in E_i\}$, the maximum coverage of any edge by the sets $E_i$. In [1], a new iterative rounding approach was presented for the problem when the sets $E_1, \ldots, E_k$ correspond to the edges $E_i = \delta(C_i)$ of a family of cuts $C_i \subseteq V$ for $i \in [k]$ that is laminar. Contrary to previous settings where iterative rounding approaches were applied, here, it is possible that an edge lies in a super-constant number of degree constraints. At each iteration, the algorithm reduces the number of degree constraints by a constant factor, replacing some constraints with new ones if necessary. This is done in such a way that degree constraints are violated by at most a constant in every iteration, leading to a spanning tree of cost at most $OPT$, that violates each degree constraint by at most $O(\log(|V|))$.

Organization. In Section 2 we present our algorithm for the matroidal degree-bounded MST problem. The analysis of the algorithm is presented in Section 3.

2 The algorithm

Since during the execution of our algorithm the underlying graph will be modified, we denote by $H = (W, F)$ the current state of the graph, whereas $G = (V, E)$ always denotes the original graph. For brevity, terminology and notation is with respect to the current graph $H$ when not specified further. To distinguish between initial degree constraints and current degree constraints, we denote by $N_w$ the current constraints for $w \in W$—which will as well be of matroidal type—whereas $M_v$ denotes the initial degree constraints at $v \in V$. The vertices of $H$ are called nodes since they might contain several vertices of $G$ due to edge contractions.

The algorithm starts with $H = G$ and $N_v = M_v$ for $v \in V$, and the LP relaxation we use is the following,

$$(LP1) \begin{array}{ll}
\min & c^T x \\
\text{s.t.} & x \in P_{st} \\
& x|_{\delta(w)} \in P_{N_w} \quad \forall w \in W
\end{array}$$

where $P_{st}$ denotes the spanning tree polytope of $H$, $P_{N_w}$ denotes the matroid polytope that corresponds to $N_w$, and $x|_{\delta(w)}$ denotes the vector obtained from $x \in \mathbb{R}^E$ by considering only the components that correspond to $\delta(w)$. 


There is a set of nodes \( Q = Q(H, x) \subseteq W \) that has a special role in our algorithm due to its relation with tight spanning tree constraints. The node set \( Q \) is defined and used after having contracted edges of weight one. Hence, assume that \( H \) does not contain any edge \( f \in F \) with \( x(f) = 1 \). Then \( Q \) is defined as follows: we start with \( Q = \emptyset \) and as long as there is a node \( w \in W \) such that \( x(\delta(w) \cap F[W \setminus Q]) = 1 \), where \( F[W \setminus Q] \) is the set of all edges with both endpoints in \( W \setminus Q \), we add \( w \) to \( Q \). One can easily observe that \( Q \) does not dependent on the order in which nodes are added to it. As we will see later, edges adjacent to these nodes can often be ignored from degree constraints due to strong restrictions that are imposed by the spanning tree constraints.

The box on top of the page gives a description of our algorithm, omitting details of how to deal with the matroidal degree bounds when removing or contracting edges. We discuss these missing points in the following.

Notice, that a basic solution to \( (LP1) \) can be determined in polynomial time by the ellipsoid method, even if the involved matroids are only accessible through an independence oracle. Depending on the matroidal degree bounds involved, \( (LP1) \) can be solved more efficiently by using a polynomially-sized extended formulation.

A tight spanning tree constraint, as considered in step \([50] \), corresponds to a set \( L \subseteq W, L \neq \emptyset \) such that \( x(F[L]) = |L| - 1 \). Fixing a tight spanning tree constraint means that this constraint has to be fulfilled with equality in all linear programs of type \( (LP1) \) solved in future iterations. It is well-known that if \( supp(x) = F \), then any maximal family of linearly independent and tight spanning tree constraints with respect to \( x \) defines the minimal face of the spanning tree polytope on which \( x \) lies (see e.g. [3]). Hence, due to step \([50] \), we have that if the LP solution at some iteration of the algorithm is on a given face of the spanning tree polytope, then all future solutions to \( (LP1) \) will be as well on this face.

Fixing tight spanning tree constraints shows to be useful since they often imply strong conditions on the edges, which can be exploited when having to make sure that degree constraints are not violated too

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The fact that \( H \) does not contain 1-edges is needed here to make sure that the order is unimportant in the definition of \( Q \). With 1-edges it might be that during the iterative construction of \( Q \), one ends up with two nodes connected by a single edge of weight one, in which case any one of the two remaining nodes can be included in \( Q \), but not both. This is actually the only bad constellation that leads to a dependency on the order in the definition of \( Q \).
much. In particular, consider a node $w \in Q$ which, in the iterative construction of $Q$, could have been added as the first node, i.e., $x(\delta(w)) = 1$. When fixing tight spanning tree constraints, one can observe that any spanning tree satisfying those tight constraints with equality contains precisely one edge adjacent to $w$. Furthermore, the fixing of tight spanning tree constraints guarantees that a node $w \in Q$ will stay in $Q$ in later iterations until an edge adjacent to $w$ is contracted. Hence, all edges being in some iteration adjacent to a node $w \in Q$, will be adjacent to a node in $Q$ in all later iterations until they are either deleted or contracted. This property is important in our approach since a type B degree adaptation ignores edges adjacent to $w$ and we want to make sure that an edge which is once ignored will never be considered during a later type B degree adaptation.

**Contracting and removing edges.** To fill in the remaining details of our algorithm, it remains to discuss how edges are contracted and removed. Throughout the algorithm, any degree constraint $N_w$ of a node $w$ containing the vertices $v_1, \ldots, v_k \in V$ can always be written as a disjoint union of matroidal constraints $N_{v_1}, \ldots, N_{v_k}$, where $N_{v_i}$ corresponds to the “remaining” degree bound at $v_i$ and is a matroid over the edges $\delta(w)$ that are adjacent to $v_i$. Whenever an edge $f = \{w_1, w_2\}$ of weight one is contracted in step 2b of the algorithm to obtain a new node $w$, the new degree constraint $N_w$ at $w$ is obtained by taking a disjoint union of the matroids $N_{w_1}/f$ and $N_{w_2}/f$, where $N_{w_1}/f$ and $N_{w_2}/f$ correspond to the matroids obtained from $N_{w_1}$ and $N_{w_2}$, respectively, by contracting $f$. This operation simply translates the degree constraints from $w_1$ and $w_2$ to the merged node $w$. The property that a degree bound on $w$ is a disjoint union of degree bounds of the vertices represented by $w$, is clearly maintained by this contraction.

As highlighted in the box, we adapt constraints by removing for some node $w \in W$ a set of edges $U \subseteq \delta(w)$ from the constraint $N_w$. When removing $U$ from $N_w$, we construct a new degree constraint given by a matroid $N_w'$ over the elements $\delta(w)$ such that the following properties hold.

**Property 2.**

i) $N_w'$ is a disjoint union of matroidal constraints $N'_{v_1}, \ldots, N'_{v_k}$ corresponding to vertices contained in $w$.

ii) edges of $U$ are free elements of $N_w'$, i.e., if $I$ is independent in $N_w'$ then $I \cup U$ is independent in $N_w'$.

iii) any independent set of $N_w'$ can be transformed into one of $N_{v_i}$ by removing at most $|U| - x(U)$ edges.

iv) the previous LP solution $x$ is still feasible with respect to $N_w'$, i.e., $x_{|\delta(w)} \in P_{N_w'}$.

Any removal operation satisfying the above properties can be used in our algorithm. Before presenting such a removal operation, we first mention a few important points. To avoid confusion, we want to highlight that removing $U$ from $N_w$ does not simply correspond to deleting the elements $U$ from the matroid $N_w$. For any edge $f \in \delta(w)$ that is free in $N_w$, we say that $f$ is not contained in the degree constraint $N_w$, and it is contained otherwise. When all edges adjacent to a given node $w$ are not contained in its degree constraint, which corresponds to $N_w$ being a free matroid, we say that the node $w$ has no degree constraint.

We now discuss how to remove a set of edges $U \subseteq \delta(w)$ from $N_w$ to obtain an adapted degree bound $N_w'$ satisfying Property 2. Let $v_1, \ldots, v_p$ be all vertices contained in the node $w$, and we consider the decomposition of $N_w$ into a disjoint union of matroids $N_{v_1}, \ldots, N_{v_k}$, where $N_{v_i}$ for $i \in [k]$ corresponds to the “remaining” degree bound at $v_i$. To remove $U$ from $N_w$, we adapt each matroid $N_{v_i}$ as follows to obtain a new matroid $N'_{v_i}$. Let $S_i$ be the ground set of $N_{v_i}$, i.e., all edges in $\delta(w)$ being adjacent to $v_i$. Let $M_1 = (S_i, I_1)$ be the matroid with independent sets

$$I_1 = \{I \subseteq S_i \cap U \mid |I| \leq |S_i \cap U| - |x(S_i \cap U)|\}.$$

Hence, $M_1$ is a special case of a partition matroid. Let $M_2 = M_1 \lor N_{v_i}$ be the union of the matroids $M_1$ and $N_{v_i}$, and let $M_3 = M_2/(S_i \cap U)$ be the matroid obtained from $M_2$ by contracting $S_i \cap U$. The degree bound $N_{v_i}'$ is obtained by a disjoint union of $M_3$ and a free matroid over the elements in $S_i \cap U$. The new degree constraint $N_w'$, that results by removing $U$ from $N_w$, is given by the disjoint union of the matroids $N_{v_1}'$, $N_{v_2}'$. 


Lemma 3. The above procedure to remove elements from a degree constraint satisfies Property 2.

Proof. By construction, when removing a set $U \subseteq \delta(w)$ from a degree bound $N_w$, which can be written as a disjoint union of $N_{v_1}, \ldots, N_{v_k}$, a matroidal bound $N'_w$ is determined which is a disjoint union of $N'_{v_1}, \ldots, N'_{v_k}$. Hence point (i) of Property 2 holds.

Let $S_i$ be the ground set of the matroids $N'_{v_i}, N_{v_i}$ for $i \in [k]$. Since $N'_w$ is a disjoint union of $N'_{v_1}, \ldots, N'_{v_k}$, it suffices for point (ii) to prove that if $I'$ is independent in $N'_{v_i}$ then $I' \cup (S_i \cap U)$ is independent in $N'_{v_i}$. This follows since $N'_{v_i}$ was obtained by a disjoint union of the matroid $M_3$, as defined above, and a free matroid over $S_i \cap U$.

For point (iii), consider an independent set $I'$ in $N'_{v_i}$. Since all edges in $U \cap S_i$ are free in $N'_{v_i}$, we can assume $(U \cap S_i) \subseteq I'$. Consider how the matroid $N'_{v_i}$ was constructed by the use of the matroids $M_1, M_2, M_3$. We start by observing that $U \cap S_i$ is independent in $M_2 = M_1 \lor N_{v_i}$. Let $r_1$ be the rank function of $N_{v_i}$, and $r_2$ be the rank function of $M_2$. Since $x \in P_{N_{v_i}}$, we have that $r_1(S_i \cap U) \geq x(S_i \cap U)$.

Furthermore, since $M_2 = M_1 \lor N_{v_i}$, it suffices for point (ii) to prove that if $(x \cap U) \subseteq I'$, then $I' \cap (S_i \cap U)$ is independent in $M_1$, we have

$$r_2(S_i \cap U) = \min \{ |S_i \cap U|, r_1(S_i \cap U) + |S_i \cap U| - |x(S_i \cap U)| \} = |S_i \cap U|,$$

showing independence of $S_i \cap U$ in $M_2$. Because $N'_w$ was obtained by a disjoint union of the matroid $M_3$ and a free matroid over the elements $S_i \cap U$, we can write $I' = I_3 \cup (S_i \cap U)$ with $I_3$ independent in $M_3$. Furthermore, as $M_3 = M_{2}/(S_i \cap U)$ and $S_i \cap U$ is independent in $M_2$, the set $I'$ is independent in $M_2$.

As $M_2 = M_1 \lor N_{v_i}$, we have $I' = I_1 \cup I$, with $I_1$ independent in $M_1$ and $I$ independent in $N_{v_i}$. Since $M_1$ is a matroid of rank $|S_i \cap U| - |x(S_i \cap U)|$, we have that $I$ is obtained from $I'$ by removing at most $|S_i \cap U| - |x(S_i \cap U)|$ elements as desired.

Let $x_i = x|_{S_i}$ for $i \in [k]$. To show point (iv), it suffices to prove that $x_i \in P_{N_{v_i}} \forall i \in [k]$, since $N'_w$ is a disjoint union of $N'_{v_1}, \ldots, N'_{v_k}$. Let $z_i \in [0, 1]^{S_i}$ be given by

$$z_i(f) = \begin{cases} 1 & \text{if } f \in S_i \cap U, \\ x_i(f) & \text{if } f \in S_i \setminus U. \end{cases}$$

Observe that $z_i - x_i \in P_{M_1}$ because the support of $z_i - x_i$ is a subset of $S_i \cap U$, $||z_i - x_i||_1 = |S_i \cap U| - x(S_i \cap U)$ and any $|S_i \cap U| - |x(S_i \cap U)|$ elements of $S_i \cap U$ are independent in $M_1$. Hence $z_i \in P_{M_2}$, since $M_2 = M_1 \lor N_{v_i}, x_i \in P_{N_{v_i}}$, and $z_i - x_i \in P_{M_1}$. As $M_3 = M_2/(S_i \cap U)$, we have that the restriction of $z_i$ on $S_i \setminus U$, which is equal to $x_i|_{S_i \setminus U}$, is in $P_{M_3}$. Since $N'_{v_i}$ is the union of $M_3$ and a free matroid over $S_i \cap U$, this finally implies that $x_i \in P_{N'_{v_i}}$.

\qed

3 Analysis of the algorithm

Lemma 4. During the execution of the algorithm, for every vertex $v \in V$, at most one constraint adaptation of type A and one of type B is performed that removes edges of $\delta(v)$ from degree constraints containing $v$.

Proof. When a type A degree adaptation is applied to a node $w \in W$ that contains $v$, no further type A degree adaptation can remove any edges in $\delta(v) \cap F$ from the constraint containing $v$, since those edges are not anymore contained in both degree constraints at their endpoints.

Similarly, when a type B degree adaptation is applied to a node $w$ that contains $v$, all edges in $\delta(w) \cap F[W \setminus Q]$ are removed from the degree constraint at $w$ and thus cannot be removed again at a later type B degree adaptation. Hence, the only possibility to remove further edges adjacent to $v$ in a later type B
degree adaptation is that some edge \( f \in F \) which was—at some iteration of the algorithm—not considered for a possible removal by a type B adaptation because of being adjacent to a node in \( Q \), can be removed by a type B adaptation at a later stage. However as already discussed, since we fix all tight spanning tree constraints, an edge that is adjacent to a node \( w \in Q \) in some iteration, will remain so until it is either deleted or contracted in step 2a or 2b of the algorithm. Hence, this “bad constellation” can never occur.

We exploit that our removal operation satisfies point (iii) of Property 2 to bound the maximum possible degree violation. In particular, for each vertex \( v \in V \), every time edges \( U \) with \( U \cap \delta(v) \neq \emptyset \) are removed from the current degree constraint \( N_w \) at the node \( w \) that contains \( v \), the degree constraint at \( v \) can be violated at most by an additional \(|U| - x(U)|\) units. Since we only perform degree adaptations for sets \( U \) with \(|U| - x(U)| \leq 4 \), and Lemma 4 guarantees that at most two adaptations are performed that involve the degree constraint at \( v \), we obtain the following result.

Corollary 5. If the algorithm terminates, then the returned tree violates each degree constraint by at most 8 units.

A main step for proving that we can always apply one of the two suggested degree adaptations, is to prove that a basic solution to \((LP1)\) is sufficiently sparse. A first important building block for proving sparsity is the following result.

Lemma 6. Let \( x \) be any solution to \((LP1)\) whose support equals \( F \). Then for every node \( w \in W \), the maximum number of linearly independent constraints of the matroid polytope \( P_{N_w} \) that are tight with respect to \( x \), is bounded by \( x(\delta(w)) \).

Proof. Let \( C \subseteq 2^{\delta(w)} \) be a family with a maximum number of sets that correspond to linearly independent constraints of the matroid polytope \( P_{N_w} \) that are tight with respect to \( x \). By standard uncrossing arguments, \( C \) can be chosen to be a chain, i.e. \( C = \{C_1, \ldots, C_p\} \) with \( C_1 \subseteq C_2 \subseteq \cdots \subseteq C_p \) (see [9] for more details). We have to show that \( p \leq x(\delta(w)) \). Let \( r \) be the rank function of \( N_w \). Define \( C_0 = \emptyset \) and for \( i \in [p] \) let \( R_i = C_i \setminus C_{i-1} \). Since \( C \) is a family of tight constraints, we have

\[
x(R_i) = r(C_i) - r(C_{i-1}) \quad \forall i \in [p] .
\]

Because \( R_i \subseteq supp(x) \), the left-hand side of (1) is strictly larger than zero. Furthermore, the right-hand side is integral and must therefore be at least one. Hence \( x(R_i) \geq 1 \) for \( i \in [p] \), which implies \( x(\delta(v)) \geq \sum_{i=1}^{p} x(R_i) \geq p \). \( \square \)

Notice, that the above lemma implies that a basic solution \( x \) to \((LP1)\) has a support of size at most \( 3(|W| - 1) \), because of the following. We can assume that all edges that are not in the support of \( x \) are deleted from the graph. Due to Lemma 6 at most \( \sum_{w \in W} x(\delta(w)) \) linearly independent constraints of the polytopes \( \{P_{N_w} \mid w \in W\} \) can be tight with respect to \( x \), and since \( x \) is in the spanning tree polytope of \( H \), this bound equals \( \sum_{w \in W} x(\delta(w)) = 2(|W| - 1) \). Furthermore, at most \( |W| - 1 \) linearly independent constraints of \( P_{st} \) are tight with respect to \( x \) due to uncrossing. This shows in particular that in the first iteration of the algorithm, we can find a node \( w \in W \) to which a type A degree constraint adaptation can be applied, because

\[
\sum_{w \in W} (|\delta(w)| - x(\delta(w))) = 2|F| - 2(|W| - 1) \leq 4(|W| - 1),
\]

and hence there must be a node \( w \in W \) with \(|\delta(w)| - x(\delta(w))| \leq 4 \).

However, in later iterations, the above reasoning alone is not anymore sufficient because many vertices do not have degree constraints anymore. Still, by assuming that no type B constraint adaptation is possible,
and using several ideas to obtain stronger sparsity, we show that the above approach of finding a good vertex for a type A degree adaptation by an averaging argument can be extended to a general iteration.

For the rest of this section, we consider an iteration of the algorithm at step 2d with a current basic solution $x$ to $(LP1)$, and assume that $|W| > 1$, and that no type B degree adaptation can be applied. We then show that there is a type A constraint adaptation that can be performed under these assumptions. This implies that our algorithm never gets stuck, and hence proves its correctness.

Since we often deal with the spare $1 - x(f)$ of an edge $f \in F$, we use the notation $z = 1 - x$. Furthermore, we partition $F$ into the sets $F_2, F_1$ and $F_0$ of edges that are contained in 2, 1 and 0 degree constraints, respectively. Hence, at the first iteration we have $F_2 = F$. Our goal is to show that $\sum_{w \in W} z(\delta(w) \cap F_2) = 2z(F_2) \leq 4|Y|$, where $Y \subseteq W$ is the set of all nodes $w$ with $\delta(w) \cap F_2 \neq \emptyset$. By an averaging argument this then implies that there is at least one node $w \in Y$ to which a type A constraint adaptation can be applied. Notice that the set $F_2$ cannot be empty (and hence also $Y \neq \emptyset$); if $F_2 = \emptyset$, then the current $LP1$ corresponds to a matroid intersection problem since every edge is contained in at most one degree constraints, and hence all degree constraints form together a single matroid over $F$; in this case $LP1$ is integral and a full spanning tree would have been contracted after step 2b, which leads to $|W| = 1$ and contradicts our assumption $|W| > 1$.

**Lemma 7.** Let $\mathcal{L}$ be a maximum family of linearly independent spanning tree constraints that are tight with respect to $x$. Then

$$2z(F_2) \leq 2|\mathcal{L}| + 2(|W| - 1) - 2(|F_0| + |F_1|) - 2x(F_0).$$

**Proof.** We can rewrite $2z(F_2)$ as follows by using the fact that $x(F) = |W| - 1$ (because $x$ is in the spanning tree polytope of $H$).

$$2z(F_2) = 2z(F) - 2z(F_0) - 2z(F_1) = 2(|F| - x(F)) - 2z(F_0) - 2z(F_1) = 2|F| - 2(|W| - 1) - 2z(F_0) - 2z(F_1) \tag{2}$$

Using classical arguments we can bound the size of the support of $x$, which is by assumption equal to $|F|$, by the number of linearly independent tight constraints from the spanning tree polytope and the degree polytopes $P_{N_w}$ for $w \in W$. In particular $x$ is uniquely defined by the tight spanning tree constraints $\mathcal{L}$ completed with some set $\mathcal{D}$ of linearly independent degree constraints, and we have $|F| = |\mathcal{L}| + |\mathcal{D}|$. The degree constraints $\mathcal{D}$ can be partitioned into $\mathcal{D}_w$ for $w \in W$, where $\mathcal{D}_w$ are linearly independent constraints of the matroid polytope $P_{N_w}$. By Lemma 6 $|\mathcal{D}_w|$ is bounded by the sum of $x$ over all edges in $\delta(w)$ that are contained in the degree constraint at $w$. When summing these bounds up over all $w \in W$, each edge in $F_2$ is counted exactly twice, and each edge in $F_1$ exactly once. Hence,

$$|\mathcal{D}| \leq 2x(F_2) + x(F_1) = 2x(F) - x(F_1) - 2x(F_0) = 2(|W| - 1) - x(F_1) - 2x(F_0).$$

Using $|F| = |\mathcal{L}| + |\mathcal{D}|$ and the above bound, we obtain from (2)

$$2z(F_2) \leq 2|\mathcal{L}| + 2(|W| - 1) - 2(z(F_0) + z(F_1) + 2x(F_0) + x(F_1)) = 2|\mathcal{L}| + 2(|W| - 1) - 2(|F_0| + |F_1|) - 2x(F_0),$$

where the last inequality follows from $z(U) + x(U) = |U|$ for any $U \subseteq F$. 

The size of a family $\mathcal{L}$ of linearly independent tight spanning tree constraints can easily be bounded by $|W| - 1$ using the fact that one can assume $\mathcal{L}$ to be laminar by standard uncrossing arguments (and $\mathcal{L}$ contains

\footnote{Notice that the assumption $|W| > 1$ is not redundant. Whereas we know that at the beginning of the iteration $|W| > 1$ did hold, this could have changed after contracting edges in step 2b.}
no singleton sets). However, this result shows not to be strong enough for our purposes. To strengthen this bound we exploit the fact that if \( L \) contains close to \( |W| - 1 \) sets, then there are many nodes \( w \in W \) that are “sandwiched” between two sets of \( L \), i.e., there are two sets \( L_1, L_2 \in L \) with \( L_2 = L_1 \cup \{w\} \), which in turn implies \( x(\delta(w) \cap E[L_2]) = 1 \). Notice that for any degree two node \( v \) which is not in \( Q \), we have \( x(U) \neq 1 \) for all \( U \subseteq \delta(w) \). Hence, such a node cannot be “sandwiched” between two tight spanning tree constraints, and we expect that the more such nodes we have, the smaller is \( |L| \). The following result quantifies this observation. It is stated in the general context of a spanning tree polytope of a general connected graph (not being linked to our degree-constrained problem).

**Lemma 8.** Let \( y \) be a point in the spanning tree polytope for a given graph \( G = (V, E) \), and let

\[
S(G, y) = \{ v \in V \mid |\delta(v)| = 2, \ y(U) \neq 1 \ \forall U \subseteq \delta(v) \}.
\]

Let \( L \subseteq 2^V \) be any linearly independent family of spanning tree constraints that are tight with respect to \( y \). Then

\[
|L| \leq |V| - 1 - \left\lfloor \frac{1}{2} S(G, y) \right\rfloor.
\]

**Proof.** To simplify notation let \( S = S(G, y) \). By standard uncrossing arguments (see for example [8]), we can assume that \( L \) is laminar. We first consider the case that there is a set \( L \in L \) with \( L \subseteq S \). Let \( L \) be a minimal set in \( L \) with this property. Since \( L \) is a tight spanning tree constraint, we have that \( y|_{E[L]} \) is in the spanning tree polytope of \( G[L] \), and hence \( y(\delta(v) \cap E[L]) \geq 1 \) for \( v \in L \). As \( L \subseteq S \), we have \( |\delta(v)| = 2 \) and \( y(e) < 1 \) for \( v \in L \) and \( e \in \delta(v) \). This implies that every vertex in \( L \) must have both of its neighbors in \( L \) to satisfy \( y(\delta(v) \cap E[L]) \geq 1 \). Since \( G \) is connected, as we assumed that there is a point in the spanning tree polytope of \( G \), we must have \( L = V = S \). Furthermore \( |V| \geq 3 \), because vertices in \( L \) have degree two. Hence the claim trivially follows since \( |L| = 1 \).

Now assume that there is no set \( L \in L \) with \( L \subseteq S \). We show that there exists a set \( R \subseteq S \) of size at least \( |R| \geq \frac{1}{2}|S| \), such that the laminar family \( L_R = \{ L \setminus R \mid L \in L \} \) over the elements \( V \setminus R \) satisfies the following:

i) \( L_R \) has no singleton sets,

ii) \( |L_R| = |L| \), i.e., any two sets \( L_1, L_2 \in L \) with \( L_1 \subseteq L_2 \), satisfy \( L_2 \setminus L_1 \subseteq R \).

Notice that this will imply the claim since \( |L_R| \leq |V \setminus R| - 1 \), because \( L_R \) is laminar without singleton sets, and hence \( |L| = |L_R| \leq |V \setminus R| - 1 \leq |V| - 1 - \frac{1}{2}|S| \). It remains to define the set \( R \) with the desired properties. For \( L \in L \), let \( V_L \subseteq L \) be all vertices in \( L \) that are not contained in any set \( P \in L \) with \( P \subseteq L \). For each set \( L \in L \), include an arbitrary set of \( \left\lfloor \frac{1}{2} |S \cap V_L| \right\rfloor \) elements of \( S \cap V_L \) in \( R \). Since the sets \( V_L \) for \( L \in L \) are a partition of all vertices \( V \), we clearly have \( |R| \geq \frac{1}{2}|S| \). Furthermore \( R \) satisfies the desired properties as we show below.

i) Assume by sake of contradiction that \( L_R \) contains a singleton set, i.e., there is a set \( L \in L \) with \( |L \setminus R| = 1 \). We can assume that \( L \) is a minimal set in \( L \). By assumption we have \( L \not\subseteq S \), and since \( R \subseteq S \), the element in \( L \setminus R \) is not in \( S \). Hence, \( R \) contains all elements \( L \cap S \), which is only possible if \( |L \cap S| = 1 \) and therefore \( |L| = 2 \). However, this implies that there must be an edge of weight one between the two vertices in \( L \), which contradicts the fact that one of those vertices is in \( S \).

ii) Assume by contradiction that there are two sets \( L_1, L_2 \in L \) with \( L_1 \not\subseteq L_2 \) that satisfy \( L_2 \setminus L_1 \subseteq R \). We can choose \( L_1 \) and \( L_2 \) such that there is no set \( L \in L \) with \( L_1 \subseteq L \not\subseteq L_2 \). By choice of \( R \), this can only happen if \( L_2 \setminus L_1 \) contains exactly one vertex \( v \in S \). This implies \( y(\delta(v) \cap E[L_2]) = 1 \), which contradicts the fact that \( v \in S \).

Lemma 8 can easily be generalized to the subgraph of a given graph \( G \) obtained by deleting the vertices \( Q(G, y) \). This form of the lemma is more useful for our analysis because of our special treatment of vertices in \( Q \).
Lemma 9. Let \( y \) be a point in the spanning tree polytope of a given graph \( G = (V, E) \) with \( y(e) \neq 1 \ \forall e \in E \), let \( G' = G[V \setminus Q(G, y)] \), and let \( y' \) be the projection of \( y \) to the edges in \( G' \). Let \( \mathcal{L} \) be any linearly independent family of spanning tree constraints of \( G \) that are tight with respect to \( y \). Then

\[
|\mathcal{L}| \leq |V| - 1 - \left\lfloor \frac{1}{2}|S(G', y')| \right\rfloor.
\]

Proof. By standard uncrossing arguments, we can assume that \( \mathcal{L} \) is a maximal laminar family of tight spanning tree constraints. We prove the result by induction on the number of elements in \( Q = Q(G, y) \). If \( Q = \emptyset \), then the result follows from Lemma 8. Let \( q \in Q \) be a possible first element added to \( Q \) during the iterative construction of \( Q \), i.e., \( y(\delta(q)) = 1 \). This implies that \( V \setminus \{q\} \) is a tight spanning tree constraint.

Let \( H = G[V \setminus \{q\}], y_H = y|_{E[V \setminus \{q\}]} \) and \( Q_H = Q(H, y_H) \). Since \( Q_H = Q \setminus \{q\} \), we can apply the induction hypothesis to the graph \( H \) to obtain that any maximal family \( \mathcal{L}_H \) of linearly independent tight spanning tree constraints in \( H \) with respect to \( y_H \) satisfies

\[
|\mathcal{L}_H| \leq |V \setminus \{q\}| - 1 - \left\lfloor \frac{1}{2}|S(G', y')| \right\rfloor.
\]

The claim follows by observing that \( \mathcal{L} = \mathcal{L}_H \cup \{V\} \) is a maximal family of tight spanning tree constraints in \( G \), and hence

\[
|\mathcal{L}| = |\mathcal{L}_H| + 1 \leq |V| - 1 - \left\lfloor \frac{1}{2}|S(G', y')| \right\rfloor.
\]

\[\square\]

Combining Lemma 8 with Lemma 7 we obtain the following bound, where we use \( S = S(H[W \setminus Q], x|_{F[W \setminus Q]}) \) to simplify the notation. To get rid of the rounding on \( \frac{1}{2}|S| \) we use \( 2\lfloor \frac{1}{2}|S| \rfloor \geq |S| - 1 \).

Corollary 10.

\[
2z(F_2) \leq 4(|W| - 1) - 2(|F_0| + |F_1|) - 2x(F_0) - |S| + 1.
\]

The following lemma implies the correctness of our algorithm. We recall that \( Y \subseteq W \) is the set of all nodes \( w \in W \) such that \( \delta(w) \cap F_2 \neq \emptyset \).

Lemma 11. There is a node \( w \in Y \) such that a type A constraint adaptation can be applied to \( w \).

Proof. Let \( \overline{Y} = W \setminus Y \). We will prove that

\[
4|\overline{Y}| \leq 2(|F_0| + |F_1|) + 2x(F_0) + |S|.
\]

Together with Corollary 10 this then implies \( 2z(F_2) \leq 4|Y| - 3 \), which in turn implies by an averaging argument that there is at least one node in \( Y \) to which a type A constraint adaptation can be applied. To prove Lemma 11 we apply a fractional token counting argument: we show that if we interpret the right-hand side of (3) as a (fractional) amount of tokens, then we can assign those tokens to the vertices in \( \overline{Y} \) such that each vertex in \( \overline{Y} \) gets at least 4 tokens.

We think of the tokens corresponding to \( S = S(H[W \setminus Q], x|_{F[W \setminus Q]}) \) as residing at the endpoints of the edges in \( F_0 \cup F_1 \). Each edge \( f \in F_0 \) gets \( 2 + 2x(f) \) tokens, \( 1 + x(f) \) at each endpoint. Each edge \( f \in F_1 \) gets \( 1 + x(f) \) tokens at the endpoint which does not contain \( f \) in its degree constraint, and \( 1 - x(f) \) tokens at the other endpoint. The tokens assigned to the endpoints of the edges thus sum up to \( 2(|F_0| + |F_1|) + 2x(F_0) \).

We start by assigning tokens to vertices in \( \overline{Y} \). By definition of the vertices in \( \overline{Y} \), we can order the elements in \( Q = \{q_1, \ldots, q_p\} \) such that for \( i \in [p] \), we have \( x(F_{q_i}) = 1 \) where \( F_{q_i} = \{\{q_i, v\} \in F \mid v \in W \setminus \{q_1, \ldots, q_{i-1}\}\} \). Since \( x(F_{q_i}) = 1 \) and no edge \( f \in F \) satisfies \( x(f) = 1 \) (such an edge would have been contracted), we have \( |F_{q_i}| \geq 2 \). Each vertex \( q_i \in Q \) gets all the tokens at both endpoints of the edges in \( F_{q_i} \). Since \( |F_{q_i}| \geq 2 \), \( q_i \) receives indeed at least four tokens.
Let $H' = (W', F') = H[W \setminus Q]$ be the induced subgraph over the vertices $W \setminus Q$, and let $x' = x|_{F'}$. Notice that $x'$ is in the spanning tree polytope of $H'$ since the set of edges $U \subseteq F$ that have at least one endpoint in $Q$ satisfy $x(U) = |U|$, and hence $x'(F') = |F'| - 1$.

The remaining tokens are allocated as follows. Each node $w \in \overline{Y} \cap W'$ gets for every edge $f \in \delta_{H'}(w)$, the tokens of $f$ at the endpoint at $w$. Furthermore, every node in $S$ gets an additional token from the term $|S|$. The attributed tokens clearly do not exceed the right-hand side of (3). It remains to show that each node $w \in \overline{Y} \cap W'$ gets at least 4 tokens. We distinguish the following three cases: (i) $w \in S$, (ii) $w \not\in S$ and none of the edges $\delta_{H'}(w)$ is contained in the degree constraint at $w$, and (iii) $w \not\in S$ and at least one edge of $\delta_{H'}(w)$ is contained in the degree constraint at $w$. Notice that the vertices considered in case (i) are precisely all vertices in $H'$ of degree two, because if there was a degree two vertex $w \in W' \setminus S$, then $w$ would have been included in $Q$. Hence, all vertices considered in case (ii) or case (iii) have degree at least 3 in $H'$.

Case (i): $w \in S$. Because $|\delta_{H'}(w)| = 2$, we have that both edges in $\delta_{H'}(w)$ are not contained in the degree constraint at $w$, since otherwise a type B degree adaptation could have been performed at $w$. Hence, $w$ receives $2 + x(f_1) + x(f_2)$ tokens from those two edges plus one token from $|S|$, resulting in $3 + x(f_1) + x(f_2)$ tokens. Since $x'$ is in the spanning tree polytope of $H'$, we have $x(f_1) + x(f_2) = x(\delta_{H'}(w)) \geq 1$, and thus $w$ receives at least 4 tokens.

Case (ii): $w \not\in S$ and none of the edges $\delta_{H'}(w)$ is contained in the degree constraint at $w$. The total number of tokens received by $w$ thus equals $|\delta_{H'}(w)| + x(\delta_{H'}(w)) \geq 3 + x(\delta_{H'}(w))$, since $|\delta_{H'}(w)| \geq 3$. The claim follows again by observing that $x'$ is in the spanning tree polytope of $H'$, which implies $x(\delta_{H'}(w)) \geq 1$.

Case (iii): $w \not\in S$ and at least one edge of $\delta_{H'}(w)$ is contained in the degree constraint at $w$. Let $U$ be the set of all edges in $\delta_{H'}(w)$ that are contained in the degree constraint at $w$. Since no type B degree adaptation can be performed at $w$, we have $|U| - x(U) > 4$. However, $|U| - x(U)$ is exactly the number of tokens that $w$ receives from the edges in $U$. Hence, at least 4 tokens are assigned to $w$. \hfill \Box

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