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Abstract

We consider the following network design problem, that we call the **Generalized Terminal Backup Problem**: given a graph (or a hypergraph) $G_0 = (V, E_0)$, a set of (at least 2) terminals $T \subseteq V$ and a requirement $r(t)$ for every $t \in T$, find a multigraph $G = (V, E)$ such that $\lambda_{G_0+G}(t, T-t) \geq r(t)$ for any $t \in T$. In the **minimum cost** version the objective is to find G minimizing the total cost $c(E) = \sum_{uv \in E} c(uv)$, given also costs $c(uv) \geq 0$ for every pair $u, v \in V$. In the **degree-specified version** the question is to decide whether such a G exists, satisfying that the number of edges is a prescribed value $m(v)$ at each node $v \in V$. The **Terminal Backup Problem** solved in [1] is the special case where G_0 is the empty graph and $r(t) = 1$ for every terminal $t \in T$. We solve the Generalized Terminal Backup Problem in the following two cases.

In the first case we solve the degree-specified version by a splitting-off theorem. This splitting-off theorem in turn provides the solution for the minimum cost version in the case when c is **node-induced**, that is $c(uv) = w(u) + w(v)$ for some node weights $w : V \rightarrow \mathbb{R}_+$.

In the second solved case we turn to the general minimum cost version, and we are able to solve it when G_0 is the empty graph. This includes the **Terminal Backup Problem** [1] ($r \equiv 1$) and the **Maximum-Weight b -matching Problem** ($T = V$). The solution depends on an interesting new variant of a theorem of Lovász and Cherkassky, and on the solution of the so-called **Simplex Matching** problem [1].

Our algorithms run in strongly polynomial time for both problems.

1 Introduction

Edge-connectivity augmentation problems usually mean the following: find a graph satisfying certain edge-connectivity requirement, and any number of parallel

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edges is allowed between any pair of the nodes. The objective function is usually to minimize the number of edges in the graph found, while the edge-connectivity requirements can vary from problem to problem. The classical result of edge-connectivity augmentation is the theorem of Watanabe and Nakamura [25], who determined the minimum number of edges of a graph $G = (V, E)$ which gives a k -edge-connected graph when added to the input graph $G_0 = (V, E_0)$. This was generalized by Bang-Jensen and Jackson [3] who solved the same problem in the case when G_0 can even be a hypergraph. Another generalization is the **local edge-connectivity augmentation problem** solved by Frank [8], which is the following. Given a graph $G_0 = (V, E_0)$ and requirement $r(u, v) \in \mathbb{Z}_+$ for every pair of nodes $u, v \in V$, find the minimum number of edges of a multigraph G satisfying $\lambda_{G_0+G}(u, v) \geq r(u, v)$ for every pair $u, v \in V$. Here, the edge-connectivity between u and v is denoted by $\lambda(u, v)$ (see Section 2.1 for definition). Note that the same problem becomes NP-complete, if G_0 can be a hypergraph [15]. Ishii and Hagiwara [12] solved the so-called **node-to-area edge-connectivity augmentation problem** which is the following. Given a graph $G_0 = (V, E_0)$, a collection of subsets \mathcal{W} of V (called **areas**) and a function $r : \mathcal{W} \rightarrow \mathbb{Z}_+$, find a graph $G = (V, E)$ with smallest possible number of edges such that $\lambda_{G_0+G}(x, W) \geq r(W)$ for any $W \in \mathcal{W}$ and $x \in V$. It is shown in [19] that this problem is NP-complete, however, the authors of [12] have given a polynomial algorithm solving it if $r(W) \geq 2$ for every $W \in \mathcal{W}$ (see also [11]). More generalizations, abstract versions and related results were given by [4, 5, 14, 23], good surveys can be found in [10, 24].

Weighted versions of edge-connectivity augmentation problems are often called **survivable network design problems**. Here we want to find a minimum-cost subgraph of a given supply graph so that the edge-connectivity requirements are satisfied. Parallel copies of the edges might or might not be allowed. These problems are usually NP-hard already in very simple cases, as an example consider the minimum-cost 2-edge-connected subgraph problem. In the Steiner Tree Problem we want to find a minimum cost set of edges that connects every pair of a set of terminals (clearly, the optimum solution can be chosen to be a tree). In its generalization, the Generalized Steiner Network Problem we have a requirement $r(u, v)$ for every pair of nodes $u, v \in V$ and the question is to find a minimum cost graph G so that $\lambda_G(u, v) \geq r(u, v)$ for every pair $u, v \in V$. Jain [13] has given a framework of 2-approximation algorithms that includes many different survivable network design problems (for example, the Generalized Steiner Network Problem). A polynomially solvable survivable network design problem is the **Terminal Backup Problem**, defined as follows. Given a set of terminals $T \subseteq V$ and costs $c(uv) \geq 0$ for every pair $u, v \in V$, find a minimum cost set of edges in which every terminal is connected to *some other terminal*. Clearly, the optimum solution of this problem can always be chosen to be a forest. The Terminal Backup Problem was introduced and solved in [1]. Note the similarity of this problem with the Steiner Tree Problem: here we want that every terminal is connected to *some other* terminal, while the Steiner Tree Problem requires that every terminal is connected to *all other* terminals.

In this paper we consider the following uncapacitated network design problem, which generalizes the Terminal Backup Problem.

Problem 1 (Generalized Terminal Backup Problem, **Problem GTBP**). *Given a graph (or a hypergraph) $G_0 = (V, E_0)$, a set of (at least 2) terminals $T \subseteq V$, and a requirement $r(t)$ for every $t \in T$, find a multigraph $G = (V, E)$ such that $\lambda_{G_0+G}(t, T-t) \geq r(t)$ for any $t \in T$.*

Note that G_0 can be a hypergraph, but G has to be a graph here, in which we can include any number of parallel edges between any pair of nodes.

In the **minimum cost version** of Problem 1 (Problem **MC-GTBP**) we want to minimize the total cost $c(E) = \sum_{uv \in E} c(uv)$ of the solution found, given also costs $c(uv) \geq 0$ for every pair $u, v \in V$. In the **degree-specified version** of Problem 1 (Problem **DS-GTBP**) we want to decide whether such a graph G exists, satisfying that the number of edges is a prescribed value $m(v)$ at each node $v \in V$.

In this paper we solve the following special cases of Problem GTBP.

1. An **edge-connectivity augmentation type problem**: we start with the minimum cost version for $c \equiv 1$ and solve the degree-specified version by a splitting-off theorem. This splitting-off theorem in turn provides the solution for the minimum cost version in the case when c is **node-induced**. Here, the cost function c is said to be **node-induced** if there exists a weight function $w : V \rightarrow \mathbb{R}_+$ such that $c(uv) = w(u) + w(v)$ for every pair $u, v \in V$.
2. A **survivable network design problem**: we turn to the general minimum cost version, and we are able to solve it when G_0 is the empty graph. The solution depends on Lemma 24, a variant of Theorem 2, which is of independent interest. The second ingredient of the solution is the algorithm given by Anshelevich and Karagiozova [1] for the problem called **Simplex Matching Problem**.

Problem GTBP is a new network design problem. It includes the Terminal Backup Problem [1] (by letting G_0 to be an empty graph and $r \equiv 1$) and the Maximum-Weight b -matching Problem ($T = V$), but it seems that this particular problem was not considered before, we have not found this type of question in the literature. A special case of this problem (the degree-specified version) was raised by András Frank (private communication). The following, somewhat related theorem of Lovász and Cherkassky can be considered as a motivation for our problem.

Theorem 2 (Lovász [16] and Cherkassky [7]). *Let $G = (V, E)$ be an undirected graph and $T \subseteq V$ a set of terminals so that the degree of v is even for every $v \in V - T$. Then there is a set F of edge-disjoint paths such that each path has its endnodes in T and for each element $t \in T$, the paths in F ending at t form a maximum set of edge-disjoint $(t, T-t)$ -paths.*

We give an interesting variant of this theorem (see Lemma 24). Theorem 2 was generalized in many directions, for example Mader [17] determined the maximum number of edge-disjoint T -paths in a graph G in which the degree of v is not necessarily even for every $v \in V - T$ (where a path is called a **T -path** if both its endnodes are in T , see also [22, Corollary 73.2b]). We could not see our Lemma 24 as an easy corollary of these results.

The paper is organized as follows. In Section 2 we give the necessary definitions and results. In Section 3 we solve the edge-connectivity augmentation problem by first solving the minimum cardinality case in subsection 3.1, and then proving the splitting-off theorem and exploring its consequences in subsection 3.3. In Section 4 we solve the survivable network design problem: in subsection 4.1 we prove the main ingredient of our solution, Lemma 24, and reduce the problem to a generalization of the simplex matching problem, in subsection 4.2 we give a pseudo-polynomial time algorithm, and in subsection 4.3 we improve the running time to strongly polynomial. We close the paper with some concluding remarks in Section 5.

2 Preliminaries

2.1 Hypergraphs and edge-connectivity

For general graph theoretic notations we will follow [9]. For subsets X, Y of a ground set V let $X - Y = \{v \in X : v \notin Y\}$; sometimes we will also use $X + Y$ to mean $X \cup Y$. A **hypergraph** (or sometimes **multihypergraph**) is a pair $H = (V, \mathcal{E})$ where V is some finite set of nodes and \mathcal{E} is a multiset of subsets of V . The members of \mathcal{E} are called **hyperedges** and the multiplicity of a hyperedge is represented as a binary number. A hyperedge of size at most 2 is called a **graph edge** (or simply **edge**), and a hyperedge of size 1 is called a **loop**. A graph is a special hypergraph containing only edges (the term **multigraph** is used as a synonym of the term graph). A **simple hypergraph** is a hypergraph in which every hyperedge has multiplicity 1. If H and G are hypergraphs on the same node set V then $H + G$ is the hypergraph on node set V in which the multiplicity of a hyperedge is the sum of its multiplicities in H and in G . For a hypergraph $H = (V, \mathcal{E})$ and a set $X \subseteq V$ we say that a hyperedge $e \in \mathcal{E}$ enters X if neither $e \cap X$ nor $e \cap (V - X)$ is empty, and we define $d_H(X) = |\{e \in \mathcal{E} : e \text{ enters } X\}|$. If a set contains only one element v then we will write v instead of $\{v\}$; thus $d_H(v)$ means $d_H(\{v\})$, etc.

A **path** between nodes s and t of a hypergraph H is an alternating sequence of distinct nodes and hyperedges $(s = v_0, e_1, v_1, e_2, \dots, e_k, v_k = t)$, such that $v_{i-1}, v_i \in e_i$ for all i between 1 and k . For sets $S, T \subseteq V$ of nodes in a hypergraph $H = (V, \mathcal{E})$, the **edge-connectivity** $\lambda_H(S, T)$ between S and T in H is defined as the maximum number of pairwise hyperedge-disjoint paths, where each path has one endnode in S , and the other in T (where we understand $\lambda_H(S, T) = \infty$ if $S \cap T \neq \emptyset$). The following theorem of Menger shows that this value coincides with the size of a minimum S - T cut.

Theorem 3 (Menger's Theorem for hypergraphs [18]). *Let $H = (V, \mathcal{E})$ be a hypergraph, and $S, T \subseteq V$. Then*

$$\lambda_H(S, T) = \min\{d_H(X) : T \subseteq X \subseteq V - S\}.$$

2.2 Skew-supermodular functions

A function $p : 2^V \rightarrow \mathbb{R} \cup \{-\infty\}$ is called a **set function**. We say that a graph G **covers** a set function p if $d_G(X) \geq p(X)$ holds for every $X \subseteq V$. Problem GTBP can be formulated as the problem of covering a skew-supermodular set function by a graph, as will be shown in Section 3. In this subsection, we describe some notations and properties of skew-supermodular functions.

A set function $p : 2^V \rightarrow \mathbb{Z} \cup \{-\infty\}$ is called **skew-supermodular** if at least one of the following two inequalities holds for every $X, Y \subseteq V$:

$$\begin{aligned} p(X) + p(Y) &\leq p(X \cap Y) + p(X \cup Y), & (\cap\cup) \\ p(X) + p(Y) &\leq p(X - Y) + p(Y - X). & (-) \end{aligned}$$

A set function is **symmetric** if $p(X) = p(V - X)$ for every $X \subseteq V$. For a hypergraph H , we can easily see that $p = -d_H$ is symmetric and satisfies both $(\cap\cup)$ and $(-)$ for any $X, Y \subseteq V$. Let the **symmetrized** p^s of a set function p be defined with the formula $p^s(X) = \max(p(X), p(V - X))$ for every $X \subseteq V$. We can see that a graph G covers p if and only if it covers p^s . We can also see the following claim.

Claim 4 ([5]). *The symmetrized of a skew-supermodular function is (symmetric and) skew supermodular.*

For a function $m : V \rightarrow \mathbb{R}$ (or a vector $m \in \mathbb{R}^V$), we denote $m(X) = \sum_{v \in X} m(v)$ for $X \subseteq V$. For a set function $p : 2^V \rightarrow \mathbb{Z} \cup \{-\infty\}$ we introduce the polyhedron

$$C(p) = \{x \in \mathbb{R}^V : x(Z) \geq p(Z) \text{ for every } Z \subseteq V, x \geq 0\}.$$

This polyhedron will be used to characterize the feasibility of the degree-specified version of Problem GTBP (see Section 3.3). An important property of $C(p)$ is the following.

Theorem 5 ([2]). *If $p : 2^V \rightarrow \mathbb{Z} \cup \{-\infty\}$ is a skew supermodular function with $p(\emptyset) \leq 0$ then $C(p)$ is an integer polyhedron (namely an **integer contrapolymatroid**).*

A **subpartition of V** is a family of disjoint subsets of V . We say that an $x \in C(p)$ is **minimal** if we cannot decrease $x(v)$ at any v without violating some condition in the definition of $C(p)$. The properties of contrapolymatroids relevant for us are formulated in the following corollary of Theorem 5. See details about contrapolymatroids in [22].

Corollary 6. *If p is as in Theorem 5 then we have the following.*

- $\max\{\sum_{X \in \mathcal{X}} p(X) : \mathcal{X} \text{ is a subpartition of } V\} = \min\{1 \cdot x : x \in C(p)\}.$
- *Any minimal $m \in C(p)$ achieves $m(V) = \min\{1 \cdot x : x \in C(p)\}.$*
- *Given any $w : V \rightarrow \mathbb{R}_+$, an (integer) optimal solution of $\min\{w \cdot x : x \in C(p)\}$ can be found in polynomial time (with a simple greedy algorithm), assuming that we can test membership in $C(p)$.*

2.3 The splitting-off operation

Let $p : 2^V \rightarrow \mathbb{Z} \cup \{-\infty\}$ be a symmetric, skew-supermodular function that satisfies $p(\emptyset) \leq 0$ and let $m : V \rightarrow \mathbb{Z}$ be a nonnegative function satisfying $m(X) \geq p(X)$ for any $X \subseteq V$ (i.e. an integer element of $C(p)$). We would like to decide whether there is a graph G covering p that satisfies $d_G(v) = m(v)$ for every $v \in V$. Let $u, v \in V$ be two nodes with $m(u), m(v) > 0$. The operation **splitting-off (at u and v)** is the following: we substitute m and p with m' and p' where $m'(x) = m(x)$ if $x \in V - \{u, v\}$ and $m'(x) = m(x) - 1$ if $x \in \{u, v\}$ and $p' = p - d_{(V, \{(uv)\})}$ (where $(V, \{(uv)\})$ is a graph having only one edge: note that p' is symmetric and skew-supermodular). One can observe that this is indeed the usual notion of splitting-off: if we introduce a graph $H = (V + s, E)$ with a new node s , every edge of E incident to s and $m(v)$ parallel edges between s and v for any $v \in V$, then we are back at the well-known (undirected) splitting-off operation (as introduced in Section 8.1 of [9]). If $m'(X) \geq p'(X)$ holds for any $X \subseteq V$, then we say that the splitting off is (p, m) -**admissible**. A set X is called (p, m) -**tight**, if $m(X) = p(X)$, and it is called (p, m) -**dangerous** if $m(X) - p(X) \leq 1$. We will only say admissible, tight and dangerous, if p and m are clear from the context. The following claim is well-known.

Claim 7 (see e.g. [5]). *The splitting off at u and v is admissible if and only if there is no dangerous set containing both u and v .*

Contraction of tight sets is a standard technique in splitting-off proofs (see for example [5], where contraction is explained in detail).

Lemma 8 (see e.g. [5]). *Let $u, v \in V$ with $m(u), m(v) > 0$. If we contract a tight set $X \subseteq V$, then the splitting at u' and v' is admissible in the contracted instance if and only if the splitting at u and v is admissible in the original instance (where u' (v') is the contracted image of u (v , respectively)).*

We will also use the following lemma.

Lemma 9 ([5, 20]). *Let $p : 2^V \rightarrow \mathbb{Z} \cup \{-\infty\}$ be a symmetric skew-supermodular function and $m \in C(p) \cap \mathbb{Z}^V$. If $\max\{p(X) : X \subseteq V\} > 1$, then there is an admissible splitting-off.*

3 Solution of the edge-connectivity augmentation problem

In this section we solve the following variants of Problem GTBP. We start with the **minimum cardinality version**, in which the number of edges $|E|$ of G is to be minimized (that is, the minimum cost version with cost function $c \equiv 1$). Then we prove a splitting-off theorem that solves the **degree-specified version**. Unlike other edge-connectivity augmentation problems, here the minimum cardinality version of the problem is easier than the degree-specified version. The splitting-off theorem gives rise to the solution of the minimum cost version for **node-induced cost function** (that is, we find a graph G minimizing $\sum_{v \in V} w(v)d_G(v)$), given some node-weights $w(v) \geq 0$ for every $v \in V$).

3.1 Notation and the minimum cardinality version

Consider Problem GTBP above. We introduce the following notation. For any terminal $t \in T$ let $d_t = \min\{d_{G_0}(X) : X \cap T = \{t\}\}$ and we say that $X \subseteq V$ is a **t -mincut** (in G_0) if $d_{G_0}(X) = d_t$ and $X \cap T = \{t\}$. We can easily see the following.

Claim 10. *The intersection and the union of two t -mincuts are again t -mincuts.*

For a terminal $t \in T$ let X_t (Y_t) denote the inclusionwise minimal (maximal, respectively) t -mincut. By Claim 10, the sets X_t and Y_t are well defined.

Lemma 11. *For two different terminals $t, t' \in T$ we have $X_t \cap Y_{t'} = \emptyset$, where $Y_{t'}$ is an arbitrary t' -mincut (and X_t is defined above). Consequently, $\{X_t : t \in T\}$ is a subpartition of V .*

Proof. Assume $X_t \cap Y_{t'} \neq \emptyset$. Since

$$d_t + d_{t'} = d_{G_0}(X_t) + d_{G_0}(Y_{t'}) \geq d_{G_0}(X_t - Y_{t'}) + d_{G_0}(Y_{t'} - X_t) \geq d_t + d_{t'},$$

we have $d_{G_0}(X_t - Y_{t'}) = d_t$, which contradicts the minimality of X_t . \square

Let us define a set function $R : 2^V \rightarrow \mathbb{Z} \cup \{-\infty\}$ by

$$R(X) = \begin{cases} r(t) & \text{if } X \cap T = \{t\}, \\ -\infty & \text{otherwise.} \end{cases}$$

It is clear that a graph G is feasible for Problem GTBP if and only if $d_G(X) \geq R(X) - d_{G_0}(X)$ holds for every subset $X \subseteq V$ (i.e., G covers $R - d_{G_0}$).

Claim 12. *The function R is skew-supermodular (and then so is the function $R - d_{G_0}$).*

Proof. Let $X, Y \subseteq V$. We can assume that $R(X)$ and $R(Y)$ are both finite, otherwise there is nothing to prove. If $X \cap T = Y \cap T$ then $(\cap \cup)$ holds for R (with equality), otherwise $(-)$ holds for R (again, with equality). The skew-supermodularity of R implies the skew-supermodularity of $R - d_{G_0}$. \square

Let $R^s(X) = \max\{R(X), R(V - X)\}$ for any $X \subseteq V$ (the **symmetrized of R**): it is a symmetric and skew supermodular function by Claim 4. Observe that $R(X) = R^s(X)$, unless $|T - X| = 1$. Let finally $p(X) = R^s(X) - d_{G_0}(X)$ for any $X \subseteq V$ (called the **deficiency function** for this instance of Problem GTBP), which is symmetric and skew-supermodular. Note that G covers $R - d_{G_0}$ if and only if G covers p . Notice that $p(X) = r(t) - d_t$ for any t -mincut X if $|T| \geq 3$.

By using these notations, we can solve the minimum cardinality version.

Theorem 13. *Suppose that $p(X_{t_1}) = \max_{t \in T} p(X_t)$ for some $t_1 \in T$. The minimum number of edges of a graph G that satisfies the requirements of Problem GTBP is equal to $\gamma = \max\{p(X_{t_1}), \lceil \frac{1}{2} \sum \{p(X_t) : t \in T, p(X_t) > 0\} \rceil\}$.*

Proof. It is clear from Lemma 11 that γ is a lower bound. On the other hand, let us find an arbitrary loopless graph G on nodeset T such that $d_G(t) \geq p(X_t)$ for every $t \in T$ and $|E(G)| = \gamma$. Such a graph exists and satisfies our requirements, since $\lambda_G(t, T - t) \geq p(X_t)$ for every $t \in T$. \square

3.2 Properties of the contrapolymatroid

In this subsection, we show some properties of the contrapolymatroid $C(p)$, where $p = R^s - d_{G_0}$ is defined as in the previous subsection.

Lemma 14. *Suppose that $p(X_{t_1}) = \max_{t \in T} p(X_t)$ for some $t_1 \in T$. Then, we have $\min\{1 \cdot x : x \in C(p)\} = \max\{p(X_{t_1}) + p(V - X_{t_1}), \sum\{p(X_t) : t \in T, p(X_t) > 0\}\}$.*

Proof. Clearly, $\max\{p(X) : X \cap T = \{t\}\} = p(X_t)$ and $\max\{p(X) : |X \cap T| = 1\} = p(X_{t_1})$. Let Z_1, Z_2, \dots, Z_k be a subpartition attaining $\sum_{i=1}^k p(Z_i) = \max\{\sum_{X \in \mathcal{X}} p(X) : \mathcal{X} \text{ is a subpartition of } V\} = \min\{1 \cdot x : x \in C(p)\}$. By the definition of the function p , for every $i = 1, 2, \dots, k$ either $|T \cap Z_i| = 1$, or $|T - Z_i| = 1$. Assume first that there exists an i such that $|T - Z_i| = 1$: say this holds for $i = 1$. In this case $k \leq 2$ and by the symmetry of p we have $p(Z_1) = p(Z_2)$, and the best value we can get for $p(Z_2)$ is $p(X_{t_1})$, that is $\sum_{i=1}^k p(Z_i) = p(X_{t_1}) + p(V - X_{t_1})$ in this case. The other case is when $|T \cap Z_i| = 1$ for every $i = 1, 2, \dots, k$. In this case $\sum_{i=1}^k p(Z_i) = \sum\{p(X_t) : t \in T, p(X_t) > 0\}$, using that $\{X_t : t \in T\}$ is a subpartition. \square

By observing that $p(X_{t_1}) = p(V - X_{t_1})$, we have the following as a corollary.

Corollary 15. *If $\min\{1 \cdot x : x \in C(p)\}$ is odd, then $p(X_{t'}) < \sum\{p(X_t) : t \in T - t', p(X_t) > 0\}$ for any $t' \in T$ and $\min\{1 \cdot x : x \in C(p)\} = \sum\{p(X_t) : t \in T, p(X_t) > 0\}$.*

Membership oracle for $C(p)$. In order to turn our proofs into polynomial algorithms, we describe a membership oracle for $C(p)$, where $p = R^s - d_{G_0}$. This oracle is needed in Corollary 6, and in our Splitting-off Theorem (see Section 3.3); note that this implies a membership oracle for $C(p - d_G)$ for any graph G , since we can add G to G_0 . Given some $x : V \rightarrow \mathbb{Z}_+$, we want to decide whether $x \in C(p)$ or not. This is done as follows. Add a new node s to G_0 and an edge with multiplicity $x(v)$ between s and every $v \in V$. Denote the resulting hypergraph by H . We claim that $x \in C(p)$ if and only if $\lambda_H(t, T - t) \geq r(t)$ holds for every $t \in T$, which can be checked with maximum flow computations. We prove this claim. If $x \notin C(p)$ then $x(Z) < R^s(Z) - d_{G_0}(Z)$ for some $Z \subseteq V$. By the definition of the function R , there exists some $t \in T$ so that $Z \cap T = \{t\}$ or $Z \cap T = T - \{t\}$: for this t we have $\lambda_H(t, T - t) < r(t)$. On the other hand, if $\lambda_H(t, T - t) < r(t)$ for some $t \in T$ then $d_H(Z) < r(t)$ for some set $Z \subseteq V + s$ separating t and $T - t$. We can assume that $s \notin Z$ and then for this set we have $x(Z) < p(Z)$.

3.3 The splitting-off theorem and its consequences

In this section we solve the degree-specified version of Problem GTBP. For an instance of this problem, recall that $p : 2^V \rightarrow \mathbb{Z} \cup \{-\infty\}$ is defined by $p(X) = R^s(X) - d_{G_0}(X)$ for $X \subseteq V$. We start with the following splitting-off result.

Theorem 16. *If $m \in C(p) \cap \mathbb{Z}^V$ is minimal and $0 < m(V) \neq 3$ then there exists an admissible splitting-off.*

Proof. If $m(V) = 1$ then m cannot be minimal, if $m(V) = 2$ then clearly there exists an admissible splitting-off, so we can assume that $m(V) \geq 4$.

If $\max\{p(X) : X \subseteq V\} > 1$ then there exists an admissible splitting-off by Lemma 9. So we can assume that $p \leq 1$. By Corollary 6, there exists a subpartition \mathcal{X} of V such that X is tight and $p(X) > 0$ for each $X \in \mathcal{X}$. Since $p \leq 1$ and $m(V) \geq 4$, we can assume that there exist tight sets $X_1, X_2, X_3, X_4 \in \mathcal{X}$ with $p(X_i) = m(X_i) = 1$ ($i = 1, 2, 3, 4$). Choose an arbitrary $x \in X_1$ and $y \in X_2$ with $m(x) > 0, m(y) > 0$, and assume that the splitting-off at x and y is not admissible. This means that there exists a set X containing x, y with $m(X) \leq p(X) + 1$. By Lemma 8, we can assume that $X_i = \{t_i\}$ for $i = 1, 2, 3, 4$, implying that $t_1, t_2 \in X$. This implies (by the definition of the function p) that $|T - X| = 1$, so we can assume that $t_3 \in X$ holds, too. But then $m(X) \geq 3$, so X cannot be dangerous, since $p(X) \leq 1$, a contradiction. \square

Corollary 17. *If $m \in \mathbb{Z}^V$ is a minimal member of $C(p)$, and $m(V)$ is even then there exists a graph G with $d_G(v) = m(v)$ at every $v \in V$ satisfying the requirements of Problem GTBP.*

Proof. Clearly follows from Theorem 16 by induction. \square

Now we are ready to give a solution of Problem DS-GTBP. If a specified degree of some vertex is too large compared to other degrees (i.e., $m(v) > m(V - v)$ for some $v \in V$), then we need to care about loops. For a node $v \in V$ in a graph $G = (V, E)$ let $d_G^+(v)$ be $d_G(v)$ plus 2 times the number of loops at v , which is a standard definition of the degree of v in a graph with loops. Recall that, for a hypergraph $G_0 = (V, \mathcal{E}_0)$ and a set $T \subseteq V$ with $|T| \geq 2$, $X \subseteq V$ is a t -mincut (in G_0) if $d_{G_0}(X) = d_t := \min\{d_{G_0}(X) : X \cap T = \{t\}\}$ and $X \cap T = \{t\}$. The following theorem gives a solution of Problem DS-GTBP.¹

Theorem 18. *Assume we are given an instance of Problem DS-GTBP (that is, a hypergraph $G_0 = (V, \mathcal{E}_0)$, a set of at least two terminals $T \subseteq V$, requirements $r : T \rightarrow \mathbb{Z}_+$, and degree-specifications $m : V \rightarrow \mathbb{Z}_+$), there exists a solution of this problem (that is a multigraph $G = (V, E)$ with $d_G^+(v) = m(v)$ at every $v \in V$ and $\lambda_{G_0+G}(t, T - t) \geq r(t)$ for every $t \in T$) if and only if*

1. $m \in C(p) \cap \mathbb{Z}_+^V$, $m(V)$ is even, and
2. at least one of the following holds:
 - (a) $\min\{1 \cdot x : x \in C(p)\}$ is even, or
 - (b) there exists a $t_0 \in T$ such that $m(X_{t_0}) > \max\{p(X_{t_0}), 0\}$, or
 - (c) there exist a $y \in V - \bigcup_{t \in T} X_t$ and a $t_0 \in T$ such that $m(y) > 0, p(X_{t_0}) > 0$, and any t_0 -mincut X containing y satisfies $m(X) > p(X) + 2$.

¹ We have to mention that in the SODA version of this paper there was an error: unfortunately Theorem 3.2 in [6] is not true. The correct splitting-off statement is given in Theorem 18.

Proof. Notice that the intersection of two t_0 -mincuts is again a t_0 -mincut, so condition (2c) can be reformulated as follows: either there is no t_0 -mincut containing y , or $m(X_0) > p(X_0) + 2$ for the inclusionwise minimal t_0 -mincut X_0 containing y .

Necessity: Assume that the required solution G exists but the conditions are not met. Clearly, the existence of G implies that $m \in C(p)$ and that $m(V)$ is even, therefore $\min\{1 \cdot x : x \in C(p)\}$ is odd, $m(X_t) = \max\{p(X_t), 0\}$ for every $t \in T$, and for every $y \in V - \bigcup_{t \in T} X_t$ and $t \in T$ such that $m(y) > 0, p(X_t) > 0$, there exists a t -mincut X containing y such that $m(X) \leq p(X) + 2$. We get a contradiction by induction on the number of edges in G . The base case $E(G) = \emptyset$ is obvious, so assume that $E(G) \neq \emptyset$. Since $\min\{1 \cdot x : x \in C(p)\}$ is odd and $m(X_t) = \max\{p(X_t), 0\}$ for every $t \in T$, there exist a $t' \in T$ and a $y \in V - \bigcup_{t \in T} X_t$ such that there exists an edge xy in G for some $x \in X_{t'}$. Let X_0 be the inclusionwise minimal t' -mincut in G_0 containing y : by our conditions, $m(X_0) = p(X_0) + 2$ must hold (we use that the splitting-off at x and y must be admissible by the existence of G , therefore $m(X_0) \leq p(X_0) + 1$ cannot be the case).

Consider the following modified instance of Problem DS-GTBP: let $G'_0 = G_0 + xy$, $m' = m - \chi_{\{x,y\}}$ and every other parameter is as in the original instance (that is, we start with a splitting-off at x and y), and let $G' = G - xy$. Let p' be the deficiency function for this modified instance (that is, $p' = R^s - d_{G'_0}$). We show that the conditions fail for this instance, and G' is a valid solution for this instance, leading to a contradiction by induction. Notice that the existence of G' implies that $m' \in C(p') \cap \mathbb{Z}_+^V$ (and clearly, $m'(V)$ is even). The inclusionwise minimal t' -mincut in G'_0 is X_0 , and $p'(X_0) = m'(X_0)$. Furthermore, X_t is the inclusionwise minimal t -mincut in G' for every $t \in T - t'$, and $p'(X_t) = p(X_t)$ and $m(X_t) = m'(X_t)$ for every $t \in T - t'$. This shows that condition (2b) fails also in the obtained instance. In what follows, we show that conditions (2a) and (2c) fail, respectively.

- **Condition (2a).** Since $p \geq p'$, $C(p) \subseteq C(p')$, therefore $\min\{1 \cdot x : x \in C(p)\} \geq \min\{1 \cdot x : x \in C(p')\}$. On the other hand, by Lemma 14, $\min\{1 \cdot x : x \in C(p')\} \geq \sum\{p'(X_t) : t \in T - t_0, p'(X_t) > 0\} + p'(X_0) = \sum\{p(X_t) : t \in T, p(X_t) > 0\} = \min\{1 \cdot x : x \in C(p)\}$, therefore equality has to hold here, so $\min\{1 \cdot x : x \in C(p')\}$ is odd.
- **Condition (2c).** Let $z \in V - (X_0 \cup \bigcup_{t \in T - t_0} X_t)$ with $m'(z) > 0$.

If Z_0 is the smallest t_0 -mincut containing z in G_0 , then $m(Z_0) \leq p(Z_0) + 2$, since the original instance does not satisfy condition (2c). Furthermore, we have $m(X_0) = p(X_0) + 2$, $m(X_0 \cap Z_0) \geq m(X_t) = p(X_t)$, and $p(Z_0) = p(X_0) = p(X_t)$. By combining these inequalities, we have $m(X_0 \cup Z_0) \leq p(X_t) + 4$. Since $X_0 \cup Z_0$ is a t_0 -mincut containing z in G'_0 and $\{x, y\} \subseteq X_0 \cup Z_0$, we obtain $m'(X_0 \cup Z_0) \leq p'(X_0 \cup Z_0) + 2$.

Let $t \in T - t_0$ with $p'(X_t) = p(X_t) > 0$ and let Z be the smallest t -mincut containing z in G_0 . Observe that Z is disjoint from X_0 (use that $d_{G_0}(X_0) + d_{G_0}(Z) \geq d_{G_0}(X_0 - Z) + d_{G_0}(Z - X_0)$, and that $z \in Z - X_0$). This gives that $m'(Z) = m(Z) \leq p(Z) + 2 = p'(Z) + 2$.

By the above arguments, the conditions fail in the obtained instance, which completes the proof of necessity.

Sufficiency: Assume that the conditions hold. If $\min\{1 \cdot x : x \in C(p)\}$ is even then we are done by Corollary 17, so assume that this is not the case. If there exists a $t_0 \in T$ such that $m(X_{t_0}) > \max\{p(X_{t_0}), 0\}$ then we can modify the instance at hand as follows: we increase $r(t_0)$ by $\max\{1, 1 - p(X_{t_0})\}$ (and we leave every other parameter unchanged), and we arrive at the previous case for this modified instance. Finally, if $\min\{1 \cdot x : x \in C(p)\}$ is odd and $m(X_t) = \max\{p(X_t), 0\}$ for every $t \in T$ then, by Condition (2c), there exists a $y \in V - \bigcup_{t \in T} X_t$ and a $t_0 \in T$ such that $m(y) > 0, p(X_{t_0}) > 0$ and any t_0 -mincut X containing y satisfies $m(X) > p(X) + 2$. Choose an arbitrary $x \in X_{t_0}$ with $m(x) > 0$ and consider the following modified instance of Problem DS-GTBP: let $G'_0 = G_0 + xy$, $m' = m - \chi_{\{x,y\}}$ and every other parameter is as in the original instance (that is, we start with a splitting-off at x and y). Let p' be the deficiency function for this modified instance.

Claim 19. *This is an admissible splitting-off, that is $m' \in C(p')$.*

Proof. Assume indirectly that there exists a set $X \subseteq V$ with $x, y \in X$ and $m(X) \leq p(X) + 1$. By Lemma 8, we can assume that $X_t = \{t\}$ is a singleton for every $t \in T$. Since X contains t_0 , we have either $T - X = \{t'\}$ for some $t' \in T - t_0$ or $X \cap T = \{t_0\}$ by the definition of p .

First, suppose that $T - X = \{t'\}$ for some $t' \in T$. Since $p(X) = r(t') - d_{G_0}(X) \leq p(X_{t'})$ and $m(X) \geq \sum\{m(X_t) : t \in T - t'\} + 1 = \sum\{p(X_t) : t \in T - t', p(X_t) > 0\} + 1 > p(X_{t'}) + 1$, where the last inequality follows from Corollary 15, X cannot be dangerous.

Second, suppose that $X \cap T = \{t_0\}$, which implies that $p(X) = r(t_0) - d_{G_0}(X) \leq p(X_{t_0})$. Since $m(X) \geq m(X_{t_0}) + 1 = p(X_{t_0}) + 1 \geq p(X) + 1$, the only way X can be dangerous is that X is a t_0 -mincut in G_0 containing y with $m(X) = p(X) + 1$, contradicting condition (2c). \square

We now finish the proof of the sufficiency by distinguishing the following two cases.

- **Case 1.** There exists no t_0 -mincut in G_0 containing y . In this case, X_{t_0} is a t_0 -mincut in G'_0 and $\min\{1 \cdot x : x \in C(p')\}$ is even, so we are done by induction.
- **Case 2.** There exists a t_0 -mincut in G_0 containing y : let X_0 be the inclusionwise minimal t_0 -mincut in G_0 containing y . In this case the inclusionwise minimal t_0 -mincut in G'_0 is X_0 and $m'(X_0) = m(X_0) - 2 > p(X_0) = p'(X_0) > 0$ by our conditions, so we are again done by induction.

\square

By using this theorem, we can solve Problem MC-GTBP in polynomial time if the weight function is node-induced. Recall that Y_t is defined as the inclusionwise maximal vertex set with $Y_t \cap T = \{t\}$ and $d_{G_0}(Y_t) = d_t$.

Theorem 20. *Given Problem GTBP and node weights $w(v)$ for every node $v \in V$, we can find a solution G minimizing $\sum_{v \in V} w(v)d_G(v)$ in polynomial time.*

Proof. By Corollary 6, we can find a vector $m \in C(p) \cap \mathbb{Z}_+^V$ minimizing $\sum_{v \in V} w(v)m(v)$. If $m(V)$ is even, then there is an optimal solution G with $d_G = m$ by Theorem 18. Otherwise, by the conditions (2b) and (2c) of Theorem 18, there exists an optimal solution G of the problem such that either

- there exists $v \in \bigcup_{t \in T} X_t$ such that $d_G = m + \chi_v$,
- there exists $v \in V - Y_t$ for some $t \in T$ with $p(X_t) > 0$ such that $d_G = m + \chi_v$,
or
- there exists $v \in Y_t - X_t$ for some $t \in T$ with $p(X_t) > 0$ such that $d_G = m + 3\chi_v$.

Note that the first case corresponds to the condition (2b) and the second and third cases correspond to the condition (2c). With this observation, when $m(V)$ is odd, we can solve the problem by the following algorithm. Let $V_1 = \bigcap \{Y_t : t \in T, p(Y_t) > 0\}$. Note that there exist at least two terminals $t \in T$ with $p(Y_t) > 0$, since $m(V)$ is odd, therefore $V_1 \cap X_t = \emptyset$ for every $t \in T$ (that is, $V - V_1 = (\bigcup_{t \in T} X_t) \cup (\bigcup_{t \in T, p(X_t) > 0} (V - Y_t))$). Let x be the vertex in $V - V_1$ with the smallest weight, and let y be the vertex in V_1 with the smallest weight. Define $m' \in \mathbb{Z}_+^V$ by $m' := m + \chi_x$ if $w(x) \leq 3w(y)$ and $m' := m + 3\chi_y$ otherwise. By Theorem 18, we can find a graph G with $d_G(v) = m'(v)$ at every $v \in V$ satisfying the requirements of Problem GTBP, which is a desired graph. \square

We mention the following related result. In our problem setting (Problem 1) we insist that G has to be a graph. If we allow hyperedges in G then we arrive at a different problem, but it is not clear how to choose the objective function. A natural candidate is to minimize the **total size** of G (where the total size of a hypergraph is the sum of the sizes of its hyperedges: note that this is twice the number of edges, if the hypergraph is in fact a graph). A more general version would consider a **node-induced cost function**, as in Theorem 20: given node weights $w(v)$ for every node $v \in V$, the cost of choosing a hyperedge is the sum of the weights of the nodes contained in that hyperedge. This general problem is solved by Szigeti in [23], as it is contained in the framework of covering a **skew-supermodular function by hyperedges**.

4 Solution of the survivable network design problem

In this section we solve the minimum cost version of Problem GTBP in the special case when G_0 is the empty graph. Let us formulate this problem separately.

Problem 21. *What is the minimum cost of a multigraph $G = (V, E)$ such that $\lambda_G(t, T - t) \geq r(t)$ for any $t \in T$, given a terminal set $T \subseteq V$ ($|T| \geq 2$), a requirement $r(t) \in \mathbb{Z}_+$ for every $t \in T$, and a cost $c(uv) \geq 0$ for every pair $u, v \in V$.*

We observe that Problem 21 is polynomially solvable if $T = V$, because now the question is to find a smallest cost multigraph $G = (V, E)$ so that the degree $d_G(v)$ of each node v is at least $r(v)$. This is a minimum-cost b -edge cover problem with

$b = r$ (which is equivalent to the maximum-weight b -matching problem with a simple reduction, see [22, Section 34.4]).

We also note that the special case $r \equiv 1$ of Problem 21 is known as the **Terminal Backup Problem**, and is shown to be polynomially solvable in [1]. It seems that the methods of [1] also apply to the case when G_0 is not an empty graph (and $r(t) = 1$ for every $t \in T$), but the details need to be clarified.

The solution for the Terminal Backup Problem in [1] is based on a polynomial time algorithm for the **Simplex Matching Problem**. To formulate this problem let us give some definitions. A hypergraph that only has hyperedges of size 2 and 3 is called a **2-3 hypergraph**. A **perfect matching** in a hypergraph $H = (U, \mathcal{E})$ is a subset of hyperedges $\mathcal{F} \subseteq \mathcal{E}$ so that each node is contained in exactly one member of \mathcal{F} . In an instance of the Simplex Matching Problem we are given a simple 2-3 hypergraph $H = (U, \mathcal{E})$ with edge costs $\gamma : \mathcal{E} \rightarrow \mathbb{R}_+$, and the objective is to find a perfect matching of H with minimum total cost. Since this problem is NP-hard in general, we consider instances with the **simplex condition**, which states that for any hyperedge $\{u_1, u_2, u_3\} \in \mathcal{E}$ of size 3, $\{u_1, u_2\}, \{u_2, u_3\}, \{u_3, u_1\} \in \mathcal{E}$ and

$$\gamma(\{u_1, u_2\}) + \gamma(\{u_2, u_3\}) + \gamma(\{u_3, u_1\}) \leq 2\gamma(\{u_1, u_2, u_3\}).$$

To simplify the terminology, the Simplex Matching Problem is meant as a problem **with the simplex condition**. The main theorem in [1] is as follows.

Theorem 22 (Anshelevich and Karagiozova [1], see also [21]). *There is a polynomial time algorithm for the Simplex Matching Problem.*

In this paper we consider and solve the following generalization of the Simplex Matching Problem that we call the **Simplex b -Edge-Cover Problem**.

Problem 23. *Let $H = (T, \mathcal{E})$ be a simple 2-3 hypergraph, let $\gamma : \mathcal{E} \rightarrow \mathbb{R}_+$ be a cost function satisfying the simplex condition, and let $b(t) \in \mathbb{Z}_+$ be a requirement for $t \in T$. Find a minimum cost multihypergraph $H' = (T, \mathcal{F})$ such that \mathcal{F} is a multiset of \mathcal{E} and $d_{H'}(t) \geq b(t)$ for any $t \in T$.*

4.1 Reduction of Problem 21 to the Simplex b -Edge-Cover Problem

In this section, we show how to reduce Problem 21 to the simplex b -edge-cover problem. We start with the following lemma which will be used in solving the survivable network design problem in the next section. This lemma is a variant of Theorem 2 and we think that it is of independent interest. Given a graph $G = (V, E)$ and some $T \subseteq V$, a **T -path** is a path $P \subseteq E$ so that its endpoints are distinct nodes of T . Similarly, a **T -3-tree** is a tree $P \subseteq E$ that does not necessarily span V , has exactly 3 leaves, these leaves are all in T , and P is not incident with other nodes in T . The unique node with degree 3 in a T -3-tree is called the **hub-node** of the T -3-tree: by the previous definition, this node is not in T .

Lemma 24. *Given a graph $G = (V, E)$ and a set $T \subseteq V$, we can find in polynomial time a set F of mutually edge-disjoint T -paths and T -3-trees so that each $t \in T$ is incident with $\lambda_G(t, T - t)$ members of F and each $v \in V - T$ is the hub node of at most one T -3-tree in F .*

Proof. If there is an edge e in G so that

$$\lambda_{G'}(t, T - t) = \lambda_G(t, T - t) \text{ for every } t \in T \quad (1)$$

holds for $G' = G - e$, then the proof is ready by induction. Similarly, if there exist a pair of edges vx, vy incident to some node $v \in V - T$ so that (1) holds for the graph $G' = (V, E - \{vx, vy\} + \{xy\})$ that we obtain after splitting off the pair vx, vy then the proof is ready by induction. So assume that neither a removable edge, nor an admissible splitting-off exists. For every $v \in V - T$, by applying Theorem 16 for the graph $G - v$, in which $m(x) = d_G(x, v)$ for every $x \in V - v$ and $r(t) = \lambda_G(t, T - t)$ for every $t \in T$, we have $d_G(v) \in \{0, 3\}$.

We claim that there is no edge between two vertices $u, v \in V - T$: this claim clearly finishes the induction. Assume indirectly that uv is such an edge, so $d_G(u) = d_G(v) = 3$. Consider an instance of Problem DS-GTBP defined by the graph $G_0 = G - u$, $m(x) = d_G(x, u)$ for every $x \in V - u$, and $r(t) = \lambda_G(t, T - t)$ for every $t \in T$, and let p be the deficiency function defined by this instance. By Lemma 14 (applied for this instance), we get that there exists a set $X_0 \subseteq V - u$ such that $v \in X_0$, $X_0 \cap T = \{t_0\}$ for some $t_0 \in T$, and $d_G(X_0) = \lambda_G(t_0, T - t_0)$ (the (p, m) -tight set containing v). Now consider the following instance of Problem DS-GTBP. Let $G'_0 = G - v$, $m'(x) = d_G(x, v)$ for every $x \in V - v$, and $r(t) = \lambda_G(t, T - t)$ for every $t \in T$. Let p' be deficiency function defined by this instance: since there is no (p', m') -admissible splitting-off, $p' \leq 1$ by Lemma 9. This together with $d_G(V - X_0) = \lambda_G(t_0, T - t_0)$ implies that $m'(V - X_0) = p'(V - X_0) = 1$ (that is, the only G -neighbour of v in $V - X_0$ is u). By our assumptions, $m'(V - v) = 3$, so let $m'(x_1) = m'(x_2) = 1$ for some distinct $x_1, x_2 \notin V - X_0$, and let X_i be the smallest (p', m') -tight sets containing x_i for $i = 1, 2$. Note that $V - X_0, X_1, X_2$ are mutually disjoint, contradicting that $(V - X_0) \cap T = T - \{t_0\}$ and $|X_i \cap T| \geq 1$ for $i = 1, 2$. \square

We note that another proof of this lemma is given in the conference version [6, Lemma 4.2]. We also note that we will not utilize below the fact that every node of $V - T$ is the hub-node of at most one T -3-tree in the decomposition given by Lemma 24. Now we show the reduction of Problem 21 to the simplex b -edge-cover problem.

Lemma 25. *We can reduce Problem 21 to the simplex b -edge-cover problem (Problem 23) in polynomial time.*

Proof. For a given instance I of Problem 21, define the family $\mathcal{E} = \binom{T}{2} \cup \binom{T}{3} \subseteq 2^T$, where $\binom{T}{2} = \{\{t_1, t_2\} \mid t_1, t_2 \in T, t_1 \neq t_2\}$, and $\binom{T}{3} = \{\{t_1, t_2, t_3\} \mid t_1, t_2, t_3 \in T, t_1 \neq t_2 \neq t_3 \neq t_1\}$, let $b = r$, and let $\gamma : \mathcal{E} \rightarrow \mathbb{R}_+$ be the cost function such that $\gamma(\{t_1, t_2\})$ is the minimum cost of a t_1 - t_2 path (with respect to the cost function c) and $\gamma(\{t_1, t_2, t_3\})$ is the minimum cost of a Steiner tree spanning t_1, t_2 and t_3 (with respect to the cost function c). Since a minimum cost Steiner tree spanning t_1, t_2 and t_3 consists of (at

most) three paths each connecting a hub vertex $v \in V$ and each t_i , it can be computed in polynomial time by guessing the hub vertex v and using a shortest path algorithm. The family \mathcal{E} and the cost function γ define an instance I' of the simplex b -edge-cover problem (note that the simplex condition naturally holds).

Clearly, any solution of instance I' of the simplex b -edge-cover problem gives rise to a solution of instance I of Problem 21 with the same cost. On the other hand, if we have a solution of instance I of Problem 21, then by Lemma 24 it defines edge-disjoint T -paths and T -3-trees: substitute each T -path with a minimum c -cost T -path between the same nodes, and substitute each T -3-tree with a minimum c -cost Steiner tree with the same leaves. This way we obtain a solution of instance I' of the simplex b -edge-cover problem with γ -cost not higher than the c -cost of the solution instance I of Problem 21 that we started with. This relation finishes the proof of this lemma. \square

We note that the corresponding result is given in [26] for the special case $r \equiv 1$.

In the rest of the paper we give a polynomial time algorithm for the simplex b -edge-cover problem. This will be done in two steps: in Section 4.2 we obtain a pseudo-polynomial time algorithm, and based on this algorithm we show in Section 4.3 how to obtain a strongly polynomial time algorithm.

4.2 Pseudo-polynomial time algorithm

Suppose that we are given an instance of the simplex b -edge-cover problem (Problem 23) consisting of a simple 2-3 hypergraph $H = (T, \mathcal{E})$, requirement $b : T \rightarrow \mathbb{Z}_+$ and cost $\gamma : \mathcal{E} \rightarrow \mathbb{R}_+$. Let us introduce the notation $B = \sum_{t \in T} b(t)$. Our pseudo-polynomial time algorithm for this problem is as follows.

Pseudo-polynomial time algorithm for Problem 23

Step 1 Construct a Simplex Matching Problem instance consisting of the simple 2-3 hypergraph $(T^+, \mathcal{E}_+ \cup \mathcal{E}_0)$ and costs as follows.

Step 1-1 The ground set is $T^+ = \{t^{(1)}, t^{(2)}, \dots, t^{(B+2)} : t \in T\}$, that is we introduce $B + 2$ copies of each node of T .

Step 1-2 The hyperedges in \mathcal{E}_+ and their costs are the following. For each $\{t_1, t_2\} \in \mathcal{E}$, add edges $\{t_1^{(i)}, t_2^{(j)}\}$ with cost $\gamma(\{t_1, t_2\})$ for all $i, j \in \{1, 2, \dots, B+2\}$. Similarly, for each $\{t_1, t_2, t_3\} \in \mathcal{E}$, add edges $\{t_1^{(i)}, t_2^{(j)}, t_3^{(k)}\}$ with cost $\gamma(\{t_1, t_2, t_3\})$ for all $i, j, k \in \{1, 2, \dots, B+2\}$.

Step 1-3 The hyperedges in \mathcal{E}_0 and their costs are the following. For each $t \in T$, add edges $\{t^{(i)}, t^{(j)}\}$ with cost 0 for $b(t) + 1 \leq i < j \leq B + 2$, and add edges $\{t^{(i)}, t^{(j)}, t^{(k)}\}$ with cost 0 for $b(t) + 1 \leq i < j < k \leq B + 2$.

Step 2 Solve the obtained Simplex Matching Problem instance using Theorem 22. Then, from the optimal solution of the Simplex Matching Problem, we can construct a solution of Problem 23 by ignoring the hyperedges in \mathcal{E}_0 and contracting $t^{(1)}, t^{(2)}, \dots, t^{(B+2)}$ to a single vertex for each $t \in T$.

Before proving the correctness of this algorithm, we give a small claim on the optimal solutions of the simplex b -edge-cover problem.

Claim 26. *Problem 23 always has an optimal solution $H' = (T, \mathcal{F})$ such that $d_{H'}(t) \in \{b(t), b(t) + 1, \dots, B\}$ for any $t \in T$, where $B = \sum_{t \in T} b(t)$.*

Proof. It suffices to show that any minimal solution $H' = (T, \mathcal{F})$ of Problem 23 satisfies that $|\mathcal{F}| \leq B$. Define $\mathcal{F}_{\text{tight}} := \{e \in \mathcal{F} : \text{there exists } t \in T \text{ such that } d_{H'}(t) = b(t) \text{ and } e \text{ enters } t\}$. By the minimality of \mathcal{F} , we have $\mathcal{F} = \mathcal{F}_{\text{tight}}$. Therefore, we have

$$|\mathcal{F}| = |\mathcal{F}_{\text{tight}}| \leq \sum_{t \in T: d_{H'}(t)=b(t)} d_{H'}(t) \leq \sum_{t \in T} b(t) = B.$$

□

Now we are ready to prove the following theorem, which will be improved in Section 4.3.

Theorem 27. *Our algorithm solves the simplex b -edge-cover problem (Problem 23) in polynomial time in $|T|$ and $B = \sum_{t \in T} b(t)$. Furthermore, we can solve Problem 21 in polynomial time in $|V|$ and $R = \sum_{t \in T} r(t)$.*

Proof. We show the optimality of the output of our algorithm. Without the set of hyperedges \mathcal{E}_0 added in Step 1-3 in our algorithm, Step 2 would find a minimum cost multihypergraph $H' = (T, \mathcal{F})$ such that \mathcal{F} is a multiset of \mathcal{E} and $d_{H'}(t) = B + 2$ for any $t \in T$. By using edges in \mathcal{E}_0 , we can cover k vertices in $t^{(b(t)+1)}, t^{(b(t)+2)}, \dots, t^{(B+2)}$, where k can be $0, 2, 3, 4, \dots, B + 2 - b(t)$ (note that we cannot cover exactly one vertex with a zero cost hyperedge). Therefore, in Step 2 of our algorithm, we obtain a minimum cost multihypergraph $H' = (T, \mathcal{F})$ such that \mathcal{F} is a multiset of \mathcal{E} and $d_{H'}(t) \in \{b(t), b(t) + 1, \dots, B\}$ for any $t \in T$, which is an optimal solution of Problem 23 by the above argument and Claim 26. We note that since we introduced $B + 2$ vertices for each vertex $u \in T$ in Step 1-1, the running time of our algorithm is polynomial in $|T|$ and $B = \sum_{t \in T} b(t)$.

Finally, by Lemma 25, the above algorithm solves Problem 21 in polynomial time in $|V|$ and $R = \sum_{t \in T} r(t)$. □

Note that in the SODA version [6] of this paper we have shown the following strengthening of Claim 26: if an instance of the simplex b -edge-cover problem is obtained from an instance of Problem 21 using Lemma 25, then it has an optimal solution $H' = (T, \mathcal{F})$ such that $d_{H'}(t) \leq \max\{r(t) : t \in T\}$. This observation reduces the running time of the algorithm when used for solving Problem 21: just use $\max\{r(t) : t \in T\}$ in the algorithm instead of $R = \sum_{t \in T} r(t)$.

4.3 Strongly polynomial time algorithm

In this subsection, we improve the running time of Theorem 27 to strongly polynomial time.

Let $H = (T, \mathcal{E})$ be a simple 2-3 hypergraph and let \mathcal{E}_2 and \mathcal{E}_3 be the sets of hyperedges in \mathcal{E} of sizes 2 and 3, respectively. A multihypergraph $H' = (T, \mathcal{F})$, where \mathcal{F} is a multiset of \mathcal{E} , is represented by a pair (x, y) with $x \in \mathbb{Z}_+^{\mathcal{E}_2}$ and $y \in \mathbb{Z}_+^{\mathcal{E}_3}$, where $x(e)$ is the multiplicity of $e \in \mathcal{E}_2$ contained in \mathcal{F} and $y(e)$ is the multiplicity of $e \in \mathcal{E}_3$ contained in \mathcal{F} . The cost of (x, y) is denoted by

$$\gamma(x, y) := \sum_{e \in \mathcal{E}_2} x(e)\gamma(e) + \sum_{e \in \mathcal{E}_3} y(e)\gamma(e).$$

For $t \in T$, define $d_x(t) := \sum\{x(e) \mid e \in \mathcal{E}_2, t \in e\}$ and $d_y(t) := \sum\{y(e) \mid e \in \mathcal{E}_3, t \in e\}$.

We first show that there exists an optimal solution of the simplex b -edge-cover problem that contains not so many hyperedges of size 3.

Lemma 28. *There exists an optimal solution (x^*, y^*) of the simplex b -edge-cover problem (Problem 23) such that $d_{y^*}(t) \leq 1$ for any $t \in T$.*

Proof. Let (x^*, y^*) be an optimal solution of the simplex b -edge-cover problem that minimizes $\|y^*\|_1$, i.e., it contains a minimum number of hyperedges of size 3. In what follows, we show that $d_{y^*}(t) \leq 1$ for any $t \in T$.

Assume that $y^*(e) \geq 2$ for some $e = \{u_1, u_2, u_3\}$. By decreasing $y^*(e)$ by two and increasing $x^*(\{u_1, u_2\})$, $x^*(\{u_2, u_3\})$, and $x^*(\{u_3, u_1\})$ by one, we obtain a feasible solution of the problem. Furthermore, since the total cost is not increased by the simplex condition, the obtained solution is also an optimal solution, which contradicts that (x^*, y^*) is an optimal solution minimizing $\|y^*\|_1$.

Assume that $y^*(\{u_1, u_2, u_3\}) \geq 1$ and $y^*(\{u_1, u_2, u_4\}) \geq 1$ for distinct u_1, u_2, u_3 and u_4 . Let (x_1, y_1) and (x_2, y_2) be feasible solutions of the problem such that

- (x_1, y_1) is obtained from (x^*, y^*) by replacing $\{u_1, u_2, u_3\}$ and $\{u_1, u_2, u_4\}$ with $\{u_1, u_2\}$, $\{u_1, u_3\}$, and $\{u_2, u_4\}$, and
- (x_2, y_2) is obtained from (x^*, y^*) by replacing $\{u_1, u_2, u_3\}$ and $\{u_1, u_2, u_4\}$ with $\{u_1, u_2\}$, $\{u_1, u_4\}$, and $\{u_2, u_3\}$.

Since

$$\begin{aligned} & 2\gamma(\{u_1, u_2, u_3\}) + 2\gamma(\{u_1, u_2, u_4\}) \\ & \geq 2\gamma(\{u_1, u_2\}) + \gamma(\{u_1, u_3\}) + \gamma(\{u_1, u_4\}) + \gamma(\{u_2, u_3\}) + \gamma(\{u_2, u_4\}) \end{aligned}$$

by the simplex condition, we have

$$2\gamma(x^*, y^*) \geq \gamma(x_1, y_1) + \gamma(x_2, y_2),$$

which implies that $\gamma(x^*, y^*) = \gamma(x_1, y_1) = \gamma(x_2, y_2)$ by the optimality of (x^*, y^*) . This contradicts that (x^*, y^*) is an optimal solution minimizing $\|y^*\|_1$.

Assume that $y^*(\{u_1, u_2, u_3\}) \geq 1$ and $y^*(\{u_1, u_4, u_5\}) \geq 1$ for distinct u_1, u_2, u_3, u_4 and u_5 . Let (x_1, y_1) and (x_2, y_2) be feasible solutions of the problem such that

- (x_1, y_1) is obtained from (x^*, y^*) by replacing $\{u_1, u_2, u_3\}$ and $\{u_1, u_4, u_5\}$ with $\{u_1, u_2\}$, $\{u_1, u_3\}$, and $\{u_4, u_5\}$, and
- (x_2, y_2) is obtained from (x^*, y^*) by replacing $\{u_1, u_2, u_3\}$ and $\{u_1, u_4, u_5\}$ with $\{u_1, u_4\}$, $\{u_1, u_5\}$, and $\{u_2, u_3\}$.

Since

$$\begin{aligned} & 2\gamma(\{u_1, u_2, u_3\}) + 2\gamma(\{u_1, u_4, u_5\}) \\ & \geq \gamma(\{u_1, u_2\}) + \gamma(\{u_1, u_3\}) + \gamma(\{u_4, u_5\}) + \gamma(\{u_1, u_4\}) + \gamma(\{u_1, u_5\}) + \gamma(\{u_2, u_3\}) \end{aligned}$$

by the simplex condition, we have

$$2\gamma(x^*, y^*) \geq \gamma(x_1, y_1) + \gamma(x_2, y_2),$$

which implies that $\gamma(x^*, y^*) = \gamma(x_1, y_1) = \gamma(x_2, y_2)$ by the optimality of (x^*, y^*) . This contradicts that (x^*, y^*) is an optimal solution minimizing $\|y^*\|_1$.

Therefore, any two hyperedges of size 3 do not contain a common vertex, that is, $d_{y^*}(t) \leq 1$ for any $t \in T$. \square

Let (x^*, y^*) be an optimal solution of the simplex b -edge-cover problem such that $d_{y^*}(t) \leq 1$ for any $t \in T$ as in Lemma 28. Let $b^* \in \mathbb{Z}_+^T$ be the vector defined by $b^*(t) = \min\{d_{x^*}(t), b(t)\}$ for $t \in T$. Then, we can see that $b \geq b^* \geq b - \mathbf{1}$, where $\mathbf{1} \in \mathbb{Z}^T$ is the all 1's vector, and x^* is a minimum cost b^* -edge-cover (in the graph (T, \mathcal{E}_2)) by the optimality of (x^*, y^*) . Here, for a graph $G = (V, E)$ and a vector $b \in \mathbb{Z}_+^V$, we say that $z \in \mathbb{Z}_+^E$ is a b -edge-cover if $\sum_{e: v \in e} z(e) \geq b(v)$ for any $v \in V$.

Since there exists a strongly polynomial time algorithm that computes a b -edge-cover with minimum total cost (see e.g. [22, Chapter 34]), we would like to utilize it to compute a minimum cost b^* -edge-cover x^* . However, since b^* is not known in advance, we cannot compute x^* directly. Our idea is to use a minimum cost b -edge-cover, which can be computed in strongly polynomial time, instead of the minimum cost b^* -edge-cover x^* . The following lemma guarantees that the minimum cost b -edge-cover is close to x^* to some extent (for a vector $p \in \mathbb{R}^A$ let $\|p\|_\infty = \max_{a \in A} |p(a)|$ and $\|p\|_1 = \sum_{a \in A} |p(a)|$).

Lemma 29 (see [22, Lemma 31.4 β]). *Let $G = (V, E)$, let $b, b' \in \mathbb{Z}_+^V$ and let $c : E \rightarrow \mathbb{R}_+$ be a cost function. Then, for any minimum cost b -edge-cover $z \in \mathbb{Z}^E$, there exists a minimum cost b' -edge-cover $z' \in \mathbb{Z}^E$ satisfying that $\|z - z'\|_\infty \leq 2\|b - b'\|_1$.*

Note that this lemma is stated in terms of a *maximum weight b -matching* in [22, Lemma 31.4 β]. Since the minimum cost b -edge-cover problem and the maximum weight b -matching problem are equivalent by considering the complement, we can see that Lemma 29 is equivalent to [22, Lemma 31.4 β]. To make the paper self-contained, we give a sketch of the proof of Lemma 29.

Proof Sketch of Lemma 29. It suffices to consider the case when $\|b - b'\|_1 = 1$. By symmetry, we may assume that there exists a vertex $u \in V$ such that $b'(u) = b(u) + 1$ and $b'(v) = b(v)$ for $v \in V \setminus \{u\}$. Let $z \in \mathbb{Z}^E$ be a minimum cost b -edge-cover, let

$z_1 \in \mathbb{Z}^E$ be a minimum cost b' -edge-cover, and suppose that z is not a b' -edge-cover. By a standard alternating path argument, we can find a walk $P = (v_0, e_1, v_1, \dots, e_l, v_l)$ such that

1. $v_0 = u$, $z_1(e_i) > z(e_i)$ if i is odd, and $z_1(e_i) < z(e_i)$ if i is even,
2. each edge e is traversed at most $\min\{|z_1(e) - z(e)|, 2\}$ times, and
3. $\sum_{e: v_l \in e} z_1(e) < \sum_{e: v_l \in e} z(e)$ if l is even, and $\sum_{e: v_l \in e} z_1(e) > \sum_{e: v_l \in e} z(e)$ if l is odd (if $v_l = v_0$ and l is odd, then $\sum_{e: v_l \in e} z_1(e) \geq \sum_{e: v_l \in e} z(e) + 2$).

Let $z_P \in \mathbb{Z}^E$ be the vector defined by

$$z_P(e) := \begin{cases} |\{i : e_i = e\}| & \text{if } z_1(e) > z(e), \\ -|\{i : e_i = e\}| & \text{if } z_1(e) < z(e), \\ 0 & \text{otherwise.} \end{cases}$$

Then, $z + z_P$ is a b' -edge-cover, $z_1 - z_P$ is a b -edge-cover, and $\|z_P\|_\infty \leq 2$. Since z is a minimum cost b -edge-cover, we have $c \cdot z \leq c \cdot (z_1 - z_P)$, and hence $c \cdot (z + z_P) \leq c \cdot z_1$. This shows that $z' := z + z_P$ is a minimum cost b' -edge-cover satisfying that $\|z' - z\|_\infty = \|z_P\|_\infty \leq 2$. \square

We now describe our strongly-polynomial time algorithm for the simplex b -edge-cover problem (Problem 23).

Strongly-polynomial time algorithm for Problem 23

Step 1 Let $x_0 \in \mathbb{Z}_+^{\mathcal{E}_2}$ be a minimum γ -cost b -edge-cover in the graph (T, \mathcal{E}_2) , which can be computed in strongly polynomial time.

Step 2 Define $x_1 \in \mathbb{Z}_+^{\mathcal{E}_2}$ by $x_1(e) = \max\{x_0(e) - 2|T|, 0\}$ for $e \in \mathcal{E}_2$.

Step 3 Let $(x_2, y_2) \in \mathbb{Z}_+^{\mathcal{E}_2} \times \mathbb{Z}_+^{\mathcal{E}_3}$ be a minimum cost pair such that $d_{x_2}(t) + d_{y_2}(t) \geq \max\{b(t) - d_{x_1}(t), 0\}$ for any $t \in T$.

Step 4 Output $(x_1 + x_2, y_2)$.

Theorem 30. *Our algorithm solves Problem 23 in strongly polynomial time.*

Proof. Let (x^*, y^*) be an optimal solution of the simplex b -edge-cover problem such that $d_{y^*}(t) \leq 1$ for any $t \in T$ as in Lemma 28. Let $b^* \in \mathbb{Z}_+^T$ be the vector defined by $b^*(t) = \min\{d_{x^*}(t), b(t)\}$ for $t \in T$. As noted earlier, $b \geq b^* \geq b - \mathbf{1}$, where $\mathbf{1} \in \mathbb{Z}^T$ is the all 1's vector, and x^* is a minimum cost b^* -edge-cover (in the graph (T, \mathcal{E}_2)) by the optimality of (x^*, y^*) . By Lemma 29, there exists a minimum cost b^* -edge-cover x^{**} (which might coincide with x^*) such that

$$\|x_0 - x^{**}\|_\infty \leq 2\|b - b^*\|_1 \leq 2|T|.$$

By the above inequality, it holds that $x^{**} \geq x_1 \geq \mathbf{0}$.

Obviously, $(x_1 + x_2, y_2)$ is a feasible solution of the simplex b -edge-cover problem. Since $(x, y) = (x^{**} - x_1, y^*)$ satisfies that $d_x(t) + d_y(t) \geq \max\{b(t) - d_{x_1}(t), 0\}$ for any $t \in T$, it holds that $\gamma(x_2, y_2) \leq \gamma(x^{**} - x_1, y^*)$ by the choice of (x_2, y_2) . Hence, we have $\gamma(x_1 + x_2, y_2) \leq \gamma(x^{**}, y^*) = \gamma(x^*, y^*)$, which means that $(x_1 + x_2, y_2)$ is an optimal solution of the problem.

The only thing left is to show is that Step 3 can be implemented in strongly polynomial time. We use our pseudo-polynomial time algorithm for the problem and note that it runs in polynomial time in $|T|$ and $\sum_{t \in T} (b(t) - d_{x_1}(t))$ by Theorem 27. Since

$$b(t) - d_{x_1}(t) \leq d_{x_0}(t) - d_{x_1}(t) \leq (|T| - 1) \|x_0 - x_1\|_\infty \leq 2|T|^2$$

for any $t \in T$, the running time of this part is indeed polynomial in $|T|$. \square

By Lemma 25, we have the following as a corollary.

Corollary 31. *Problem 21 can be solved in strongly polynomial time.*

5 Concluding remarks

Note that in Problem GTBP we allow an arbitrary number of parallel copies of any edge in G , therefore our problem is an **uncapacitated network design problem**. A natural capacitated extension of our problem would be the following (we only formulate the minimum cost version here).

Problem 32. *In the minimum cost version of Problem 1, find a graph $G = (V, E)$ also satisfying that the number of parallel copies of an edge $e \in E$ is at most some capacity $cap(e) \in \mathbb{Z}_+$, that is given in advance.*

This problem can also be seen as a **minimum cost subgraph problem** by introducing a supply graph with edge-multiplicities $cap(uv)$ for every $u, v \in V$. Note that Problem 1 is the special case of this problem by setting $cap(uv) = \sum_{t \in T} r(t)$ for every pair $u, v \in V$. We could not extend our results to Problem 32. The problem is open even if G_0 is the empty graph. Note that Jain's framework implies a 2-approximation algorithm for this problem in the case when the capacities do not exceed some fixed constant (that is not part of the input).

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