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# Note Upgrading min-max spanning tree problem under various cost functions

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## ABSTRACT

This paper addresses upgrading min-max spanning tree problem (MMST). Given a graph G(V, E), the aim of this problem is to modify edge weights under certain limits and given budget so that the MMST with respect to perturbed graph improves as much as possible. We present a complexity result for general non-decreasing cost functions. In special case, it is shown that the problem under linear and sum-type Hamming cost function can be solved in  $O(|E|^2)$  and  $O(|E|\log|E|\log|V|)$  time, respectively.

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# 1. Introduction

Usually the instances of classical network optimization problems are static and realistic, but, applications often admit some improvements of parameters under specific circumstances. This leads to the area of network "upgrading" problems. In these special network modification problems, one may invest a budget in order to change the parameters (weights of edges or vertices) of the given network within certain limits such that the optimal objective value with respect to the modified parameters is minimized while the topological structure of the graph remains unchanged [1,2]. In this paper we consider the upgrading min-max spanning tree (MMST) problem where a budget for reducing the weights of edges is assigned and the edge weights can be modified within given intervals. In this kind of problem, the network is modified before finding the MMST on the network.

Upgrading approach have already been applied to several classical optimization problems. For instance, Fulkerson and Harding [3] and Hambrusch and Hung-Yi Tu [4] investigated upgrading the shortest and longest path problems. Upgrading the network flow problem is considered by Phillips [5]. Gassner dealt with up- and downgrading 1-center and the 1-median problems in [6] and [7,8], respectively. Sepasian and Rahbarnia [9] proposed a linear time algorithm for solving upgrading 1-median on paths. Some authors also investigated upgrading minimum spanning tree problem. Dragmeister et al. [10], Frederickson and Solis-Oba [11] and Krumke et al. [12–14], developed up- and downgrading Steiner and minimum spanning tree problems. Another version of upgrading minimum spanning tree problem is investigated by Krumke et al. [15], [16] and Alvarez et al. [17] in which the weight of each edge is a function of the weights of two vertices connected to it. The aim is modifying the vertex weights in order to upgrading the minimum spanning tree.

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Given a connected graph G = (V, E), with |V| = m and |E| = n, a nonnegative weight  $w_e$  is assigned to each  $e \in E$ . Let  $\mathcal{T}$  denote the collection of all spanning trees of G. The weight of a spanning tree  $T \in \mathcal{T}$  is defined as  $w(T) = \max\{w_e | e \in T\}$ . The MMST is a tree  $T^* \in \mathcal{T}$  with smallest weight, i.e.  $w(T^*) = \min\{w(T) | T \in \mathcal{T}\}$ . The MMST problem has applications in communication network (under the name min–max broadcasting tree) [18–22], and molecular biology [23].

Camerini [24] proposed a linear time algorithm to solve the problem. After that, different versions of the problem was studied. Ishii and Nishida [25] investigated the stochastic version of the problem in which edge weights are random variables. Berman et al. [26] dealt with the constrained version where the sum of edge weights of the tree should not exceed a given upper bound. Afterward, their algorithm was improved by Punnen and Nair [27]. Recently, Anderson and Ras [28] studied the MMST with additional condition that every vertex degree is bounded above in MMST.

In this paper, we investigate upgrading MMST problem. The upgrading MMST problem is aimed at finding new values for edge weights and a new MMST associated with these new parameters, so that this tree is the best over all allowed parameters. Put simply, the aim is to find a MMST of the network when edge weights change.

Some kind of network improvement approaches was applied to MMST by several authors. Liu and Yao [29] introduced the inverse MMST in which the edge weights are modified within given budget so that a candidate spanning tree becomes the MMST. They considered the problem under the sum-type Hamming distance. Also, the constrained inverse MMST under the weighted Hamming distance was studied by Liu and Wang [30].

In the rest of this section the problem is introduced and formulated formally. In section 2 the problem is investigated under non-decreasing and linear cost functions, and a polynomial time algorithm is proposed. In section 3 the problem is considered under sum-type Hamming cost function.

Let us model the problem formally. Assume that the weight of each edge  $e \in E$ , can be decreased with a given cost, and the modified weight is  $\tilde{w}_e$ , that is  $\tilde{w}_e \leq w_e$ . Each modified edge weight is restricted by a nonnegative lower bound  $\underline{w}_e$  i.e.  $\tilde{w}_e \geq \underline{w}_e$ . The weight of a tree T with respect to the new vector of edge weights,  $\tilde{w}$ , is denoted by  $\tilde{w}(T)$ . Assume that the cost of reducing the weight of edge  $e \in E$  is given by non-decreasing function  $f_e(x)$ . Moreover, let B be a positive budget which limits the cost of edge weight reductions. Using the above notations, the upgrading MMST problem is formally formulated as:

$$\min_{T \in \mathcal{T}} \quad \tilde{w}(T) \\
\text{s.t.} \quad \sum_{e \in E} f_e(w_e - \tilde{w}_e) \le B \\
\underline{w}_e \le \tilde{w}_e \le w_e, \quad \forall e \in E$$
(1)

# 2. Upgrading MMST problem under non-decreasing cost functions

In this section we assume that cost function  $f_e(x)$  for each edge  $e \in E$  is non-decreasing. Let  $v^*$  be the optimal value of (1). It is easy to see that  $v^*$  is bounded above by the objective value of current MMST,  $\bar{v}$ , and is bounded below by  $\min_{e \in E} \underline{w}$ . A tighter lower bound for  $v^*$  is the weight of MMST,  $\underline{T}$ , corresponding to weight vector  $\underline{w}$ . Denote this lower bound by  $\underline{v}$ . Observe that  $\underline{v}$  and  $\bar{v}$  can be found in linear time by [24]. Thus,  $v^*$  belongs to interval  $\overline{\mathcal{V}} = [\underline{v}, \bar{v}]$ . From now on we assume that  $\sum_{e \in \underline{T}} f_e(w_e - \underline{w}_e) > B$ , since otherwise,  $\underline{T}$  is the upgraded MMST with weight  $v = \underline{v}$  and nothing is left to solve. By this assumption, the whole budget will be used in order to get the optimal solution.

Before we consider how to solve the problem, we concentrate on a problem closely related to problem (1). Indeed, given  $\nu \in \mathcal{V}$ , the problem is to make a MMST with weight  $\nu$  at minimum cost so that new weights satisfy bound restrictions.

For  $\nu \in \mathcal{V}$ , consider the new graph  $G_{\nu} = (V, E)$  where the underlying graph is G and the weight of each  $e \in E$ ,  $\lambda_e(\nu)$ , is defined as

$$\lambda_{e}(\nu) := \begin{cases} 0 & \text{if } \nu \ge w_{e} \\ f_{e}(w_{e} - \nu) & \text{if } \underline{w}_{e} \le \nu < w_{e} \\ B + 1 & \text{if } \nu < \underline{w}_{e} \end{cases}$$
(2)

For a spanning tree *T* of  $G_{\nu}$ , define  $D_T(\nu) = \sum_{e \in T} \lambda_e(\nu)$ . If *T* is a minimum sum spanning tree of  $G_{\nu}$ , then  $D(\nu)$  denotes the sum of the edge weights of *T*. The following lemma is straightforward.

**Lemma 1.** The perturbed G has a MMST with weight v if  $G_v$  has a minimum sum spanning tree, T, with total sum weight  $D(v) \leq B$ . Furthermore, T is a MMST of G with weight v.

According to Lemma 1, one can find a minimum sum spanning tree *T* of  $G_{\nu}$  and change the weight of each edge  $e \in T$  with  $\underline{w}_e \leq \nu < w_e$  to  $\nu$ . Therefore, *T* is the improved MMST with  $w(T) = \nu$  and the total cost is equal to  $D(\nu)$ .

We have the following fact.

**Lemma 2.** D(v) is a non-increasing function of v.

**Proof.** If  $\nu_1 < \nu_2$ , then  $\lambda_e(\nu_1) \ge \lambda_e(\nu_2)$  for each  $e \in E$ , since  $f_e$  is non-decreasing. This implies that  $D(\nu)$  is non-increasing.  $\Box$ 

Let  $\mathcal{R} = \{w_e | e \in E\} \cap \mathcal{V}$  and  $\mathcal{S} = \{\underline{w}_e | e \in E\} \cap \mathcal{V}$ . Define  $\mathcal{P} = \mathcal{R} \cup \mathcal{S} \cup \{\underline{\nu}, \overline{\nu}\}$ . Assume that the elements of  $\mathcal{P}$  are sorted increasingly, i.e.,  $\underline{\nu} = p_1 < p_2 < \cdots < p_r = \overline{\nu}$ . We find an interval  $[p_k, p_{k+1}]$  with  $D(p_k) > B$  and  $D(p_{k+1}) \leq B$  by applying binary search algorithm. By Lemma 2, the optimal solution of (1),  $\nu^*$ , belongs to  $(p_k, p_{k+1}]$ . If  $D(p_{k+1}) = B$ , then  $p_{k+1}$  is the optimal value. Otherwise,  $\nu^*$  is in  $(p_k, p_{k+1})$  with  $D(\nu^*) = B$ . Since this interval does not contain any element of  $\mathcal{P}$ , each edge weight  $\lambda_e(\nu)$  is not piecewise. More precisely, for each *e* in optimal MMST,  $\lambda_e(\nu)$  is identical to either  $f_e(w_e - \nu)$  or 0 on  $(p_k, p_{k+1})$ .

When  $\nu$  changes in  $(p_k, p_{k+1})$ , a sequence of minimum spanning trees are constructed. In [31,32], the problem of finding all such intermediate minimum spanning trees is solved when the edge weight functions are all linear. In our problem, edge weight functions are decreasing, however, with a similar method in linear case one can find all minimum spanning trees when  $\nu$  varies in  $(p_k, p_{k+1})$ . Here, we discuss the main idea of the solution method. Suppose *T* is a minimum spanning tree corresponding to the given  $\nu_0 \in (p_k, p_{k+1})$ . Let  $e \notin T$ . When  $\nu$  is increased from its current value,  $\nu_0$ , the weight of each edge is decreased, since  $\lambda_g = f_g(w_e - \nu)$  and for all edge g,  $f_g$  is non-decreasing by assumption. Now, e enters *T* only if its weight becomes less than at least one of edges, say  $h \in T$ , in the cycle induced by adding e to *T*. That is, h replaced by e only if  $f_e(w_e - \nu) = f_h(w_h - \nu)$ . Therefore, the new spanning tree could be found by inspecting between all intersection points of weight functions.

Our main goal is to find  $v \in (p_k, p_{k+1})$  with D(v) = B. Thus, it is not necessary to compute all intermediate spanning trees. Indeed, we can choose all intersection points belonged to  $(p_k, p_{k+1})$ , sort them increasingly, and use binary search algorithm in order to find two points  $q_t$  and  $q_{t+1}$  with  $D(q_t) > B$  and  $D(q_{t+1}) \leq B$ . Note that the minimum spanning tree does not change on interval  $(q_t, q_{t+1}]$ ; this unique minimum spanning tree is the solution of the problem. It is left to calculate the minimum weight v by solving the equation  $D(v) = \sum_{e \in T} \lambda_e(v) = B$ . Here we assume that this equation could be solved by calling function solve(). We also assume that finding the number of all intersection points of cost functions, s, needs O(g(s)) time. We have the following result.

**Theorem 1.** The upgrading MMST problem under non-decreasing cost functions can be solved in  $O(g(s) + n \log m (\log s + \log n))$  time plus O(1) call of function solve().

**Proof.** Computing  $\bar{\nu}$  and  $\underline{\nu}$  needs linear time. The elements of  $\mathcal{P}$  can be sorted in  $O(n \log n)$  time. Then, binary search algorithm is used for determining interval  $(p_k, p_{k+1})$ . Each iteration of binary search algorithm needs solving a MMST which can be solved by well-known Kruskal's algorithm in  $O(n \log m)$ . Thus, finding interval  $(p_k, p_{k+1})$  needs  $O(n \log m \log n)$ . Similarly, finding interval  $[q_t, q_{t+1}]$  needs  $O(g(s) + n \log m \log s)$ , where O(g(s)) is included for finding all intersections points of cost functions. Thus, the problem is solvable in  $O(g(s) + n \log m (\log n + \log s))$  time, plus O(1) call of function solve().  $\Box$ 

**Corollary 1.** If the cost functions are linear, then upgrading MMST can be solved in  $O(n^2)$  time.

**Proof.** Since all cost functions are linear, then,  $s = O(n^2)$  and each intersection point can be found in O(1) time. In addition, equation D(v) = B is linear and function solve() runs in O(1). According to Theorem 1 and the fact that  $n = O(m^2)$ , we conclude that the problem can be solved in

$$O(g(s) + n \log m(\log n + \log s)) + O(1) \text{ call of solve}() \text{ function} = O(n^2 + n \log m \log n^2 + n \log m \log n) + O(1) = O(n^2). \quad \Box$$

**Remark 1.** There is another improvement network problem named downgrading MMST problem. The aim of this problem is increasing the edge weights under the given budget so that the weight of MMST problem with respect to new edge weights is maximized. We can solve this problem by a similar method mentioned in this paper with minor changes. We ignore details because of avoiding of duplication.

## 3. Upgrading MMST problem under sum-type Hamming cost function

In this section we assume that for each  $e \in E$ , the cost function  $f_e(x_e)$  is of the form  $c_e H(w_e, \tilde{w}_e)$  where  $c_e$  is positive scalar and H(x, y) is the Hamming distance between x and y, i.e.,

$$H(x, y) := \begin{cases} 0 & \text{if } x = y \\ 1 & \text{if } x \neq y \end{cases}$$
(3)

Let  $\underline{T}$  be the MMST with weight  $\underline{\nu}$  with respect to weight vector  $\underline{w}$ . We assume that  $\sum_{e \in \underline{T}} c_e H(w_e, \tilde{w}_e) > B$ . Since, otherwise,  $\underline{T}$  is upgraded MMST and nothing is left to solve. So, whole budget will be used. Therefore  $\underline{\nu}$  is a lower bound for the optimal value of problem (1),  $\nu^*$ , which can be found in linear time.

On the other hand, suppose that T' is the MMST with respect to the initial weight vector w. It is easy to see that the weight of T' is an upper bound for optimal solution of (1). However, a tighter upper bound can be found by improving the weight of T' as much as possible. Let  $\bar{\nu}$  be the weight of the improved T' in the perturbed G. We can find  $\bar{\nu}$  with a greedy approach as follows. First, notice that when the Hamming distance is used, we pay the whole cost to modify an edge weight without attention to the magnitude. So the best choice for new weights are the lower bounds. Let  $\underline{w}_{max} = max\{\underline{w}_e | e \in T'\}$ . It is clear that  $\bar{\nu} \ge \underline{w}_{max}$ . Thus,  $\bar{\nu}$  belongs to  $P = \{w_e | e \in T', w_e > \underline{w}_{max}\} \cup \{\underline{w}_{max}\}$ . We first sort the elements of P in non-increasing order, i.e.  $P = \{p_1, \ldots, p_k\}$  with  $p_1 \ge p_2 \ge \ldots \ge p_k = \underline{w}_{max}$ . For simplicity, we denote by  $c_k$  the constant in Hamming function cost corresponding to  $p_k$ . Now, if  $c_1 \le B$ , then we reduce  $p_1$  to its lower bound and modify the budget. We continue this way until  $c_i$  exceeds the remaining budget for some i. Therefore obtaining  $\bar{\nu}$  needs  $O(|E(T')|\log(|E(T')|)) = O(n \log n)$  time.

Again, we consider a problem closely related to problem (1). The aim is to answer the question of finding the MMST with weight  $v \in \mathcal{V} = [\underline{v}, \overline{v}]$  under bound restrictions and budget constraint. Consider the graph  $G_v$  introduced in section 2 where edge weights are defined as:

$$\lambda_{e}(\nu) := \begin{cases} 0 & \text{if } \nu \ge w_{e} \\ c_{e} & \text{if } \underline{w}_{e} \le \nu < w_{e} \\ B+1 & \text{if } \nu < \underline{w}_{e} \end{cases}$$
(4)

and D(v) is defined as before.

We have the following lemma.

**Lemma 3.** The perturbed G has a MMST with weight v if  $G_v$  has a minimum spanning tree T of weight  $D(v) \leq B$  and T is a MMST in G with weight v.

Similar to Lemma 2, D(v) is a non-increasing function. We recall that again the edge weights does not change unless it reaches its lower bound. Therefore the new weights belong to the following set:

 $\mathcal{Q} = \left( \{ w_e | e \in E \} \cup \{ \underline{w}_e | e \in E \} \cup \{ \underline{\nu}, \overline{\nu} \} \right) \cap [\underline{\nu}, \overline{\nu}].$ 

Assume that the elements of Q are sorted in increasing order, i.e.  $Q = \{q_1, q_2, \dots, q_s\}$  with  $\underline{v} = q_1 < \dots < q_s = \overline{v}$ . Now,  $q_k$  is the optimum value of the problem where k is the smallest index k with  $D(q_k) \leq B$ . It can be found by binary search. Therefore, we have the following result.

**Theorem 2.** The upgrading MMST problem under sum-type Hamming cost function can be solved in O (n logm logn) time.

**Proof.** We need to solve a MMST to find  $\underline{v}$ ; it needs linear time. Finding  $\overline{v}$  needs sorting the elements of P, which runs in  $O(n \log n)$  time. Sorting the elements of Q needs  $O(n \log n)$  time. The last binary search runs in  $O(n \log m \log n)$ , since it calls Kruskal's algorithm at most  $\log n$  times. Thus, problem can be solved in  $O(n \log n) + O(n \log n) + O(n \log m \log n) = O(n \log m \log n)$ .  $\Box$ 

# 4. Conclusion

In this paper the upgrading MMST problem has been investigated. It is proved that this problem could be transformed into parametric minimum sum spanning tree problem. It is shown that the upgrading MMST under linear and sum-type Hamming cost functions are solvable in polynomial time. Further research on the upgrading other versions of MMST problems (stochastic or constrained) on networks under various objective functions, seems to be promising.

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