

# Better approximation bounds for the network and Euclidean Steiner tree problems

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## Abstract

The network and Euclidean Steiner tree problems require a shortest tree spanning a given vertex subset within a network  $G = (V, E, d)$  and Euclidean plane, respectively. For these problems, we present a series of heuristics finding approximate Steiner tree with performance guarantee coming arbitrary close to  $1 + \ln 2 \approx 1.693$  and  $1 + \ln \frac{2}{\sqrt{3}} \approx 1.1438$ , respectively. The best previously known corresponding values are close to 1.746 and 1.1546.

**Keywords:** Combinatorial problems, approximation algorithms, Steiner trees.

## 1 Introduction

Let  $G = (V, E, d)$  be a graph with a vertex set  $V$ , an edge set  $E$  and distance function  $d : E \rightarrow \mathbb{R}^+$ . A tree  $T$  is called a *Steiner tree* of  $S$ ,  $S \subset V$ , if  $S$  is contained in the vertex set of  $T$ .

**Network Steiner Problem (NSP).** Given  $G$  and  $S$ , find the shortest Steiner tree (also called the *Steiner minimal tree*) of  $S$ .

This problem is *NP*-complete [9], so many approximation algorithms for Steiner minimal trees appeared in the last two decades. The quality of an approximation algorithm is measured by its performance ratio: an upper bound on the ratio between the achieved length and the optimal length.

NSP belongs to *MAX SNP*-class [3], so the constant factor approximation algorithm exists [13] and for some  $\epsilon > 1$ ,  $\epsilon$ -approximation is *NP*-complete [1]. To find such  $\epsilon$  is an important open problem.

Without loss of generality we may assume that  $G$  is complete, i.e. for any  $u - v$ -path in  $G$ , there is an edge  $(u, v) \in E$ . Moreover, the length of any edge in  $G$  coincides with the length of the minimal path between its ends.  $G_S$  denotes a subgraph of  $G$  induced by the set  $S$ .

**Euclidean Steiner Problem (ESP).** Given a point set  $S$  in Euclidean plane, find the shortest Steiner tree spanning  $S$ .

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For ESP, the graph  $G_S$  is the same as for NSP.

A well-known minimum spanning tree heuristic for the Steiner tree problem approximates a Steiner minimal tree with a minimum length spanning tree (MST) of a complete graph  $G_S$ . Du and Hwang [7], Takahashi and Matsuyama [13] proved that the exact performance ratios of this heuristic are equal to  $\frac{2}{\sqrt{3}} \approx 1.1547$  and 2 for ESP and NSP, respectively.

Two better heuristics appeared recently while consideration of so called  $k$ -restricted Steiner trees. The approximation guarantee of the generalized greedy heuristic is bounded by  $11/6$  [15] for NSP and  $1.1546$  for ESP [8]. The series of evaluation heuristics achieves better guarantees while increasing of their runtime. Their performance ratio converges to a value close to  $1.746$  [4, 5] for NSP.

The main result of this paper is the following

**Theorem 1** *There is a polynomial-time  $(1 + \ln r_2)$ -approximation scheme  $\{A_k, k = 2, 3, \dots\}$  for the Steiner tree problems, where  $r_2$  is a performance ratio of the minimum spanning tree heuristic.*

**Corollary 1** *There is a polynomial-time  $(1 + \ln 2)$ - and  $(1 + \ln \frac{2}{\sqrt{3}})$ -approximation scheme for NSP and ESP, respectively.*

In the next section we describe a general framework for different heuristics solving Steiner tree problems and approximation algorithms  $A_k$ . In section 3 we prove the performance guarantee for these algorithms.

## 2 The greedy contraction framework

First we introduce some denotations:  $Smt(S)$  and  $smt(S)$  are a Steiner minimal tree of  $S$  and its length, respectively. For a complete graph  $G_S$ ,  $Mst(S)$  denotes MST of  $G_S$ , and  $M(S)$  denotes its length.  $Smt(S)$  may in general contain vertices of  $V \setminus S$ . So any Steiner tree contains the set  $S$  of *terminal vertices* and some additional vertices. A Steiner tree is called a *full Steiner tree*, if it does not contain internal terminal vertices. If a Steiner tree is not full, then we can split it into the union of edge-disjoint full Steiner subtrees which are called *full components*.

Contraction of a full Steiner tree  $T$  means reducing to 0 the lengths of edges of  $G_S$  connecting terminal vertices of  $T$ . We denote by  $S/T$  the result of contraction. Note that contraction reduces the value  $M(S)$ , i.e.  $M(S/T) \leq M(S)$ .

For all the Steiner tree problems, the following greedy contraction framework is successfully used in approximations.

### Greedy contraction framework (GCF)

- (1) repeat until  $M_0(S) = 0$ 
  - (a) find a full Steiner tree  $T^*$  in a class  $K$  which minimizes a criterion function  $f(T)$ :  $T^* \leftarrow \arg \min_{T \in K} f(T)$ .
  - (b) insert  $T^*$  in  $LIST$ .
  - (c) contract  $T^*$ ,  $S \leftarrow S/T^*$ .
- (2) reconstruct an output Steiner tree from trees of  $LIST$ .

To determine a performance guarantee of an algorithm  $A$  embedded in GCF we may bound the following two ratios:

$$a_1 = \frac{smt_{\tilde{K}}}{smt},$$

where  $smt_{\tilde{K}}$  is the minimum tree length in the family  $\tilde{K}$  containing all Steiner trees with full components belonging to  $K$ , and

$$a_2 = \frac{st_A}{smt_{\tilde{K}}},$$

where  $st_A$  is the length of the output Steiner tree of  $A$ . Thus, a performance ratio  $pr(A)$  equals to their product  $pr(A) = a_1 \cdot a_2$ .

Many famous heuristics can be embedded in this framework considering different definitions of a class  $K$  and a criterion function  $f$ .

**The minimum spanning tree heuristic.** The class  $K$  consists of all paths in  $G$  and  $f(T) = d(T)$ . The exact bounds for  $a_1$  were proved in [7, 13]. An equality  $a_2 = 1$  follows from the famous fact that the greedy algorithm finds exact MST of a graph.

**The Rayward-Smith's heuristic (RSH)** [12]. The class  $K$  contains stars and  $f(T) = \frac{d(T)}{r-1}$ , where  $r$  is the number of leaves of  $T$ . For NSP, exact upper bounds  $a_1 \leq 5/3$  and  $a_1 \cdot a_2 \leq 2$  were proved in [15, 16] and [14], respectively.

**The generalized greedy heuristic (GGH)** [16]. The class  $K$  consists of trees with  $k$  terminal vertices and  $f(T) = d(T) - (M(S) - M(S/T))$ . For NSP, the same (as for RSH) exact upper bound  $a_1 \leq 5/3$  [15, 16] holds. For  $k = 3$ , the upper bound  $a_1 \cdot a_2 \leq 11/6$  [16] may be not exact. Berman [6] found an instance of NSP such that  $a_1 \cdot a_2 > 5/3$  and conjectured that  $a_1 \cdot a_2 \leq 43/24$  ( $k = 3$ ). For ESP,  $a_1 \cdot a_2 \leq 1.1546$  ( $k = 128$ ) [8].

In this paper we present

**The relative greedy heuristic (RGH).** The class  $K = K_k$  contains all trees with at most  $k$  terminal vertices. The criterion function  $f$  is defined slightly different than for GGH.

$$f(T) = \frac{d(T)}{M(S) - M(S/T)} \quad (1)$$

Now we describe the class  $K = K_k$  and review known facts about it. A Steiner tree is called  $k$ -restricted if all its full components have at most  $k$  terminal vertices. Let the shortest  $k$ -restricted Steiner tree for the set  $S$ , denoted by  $Smt_k(S)$ , have the length  $smt_k(S)$ . Note, that  $Smt_2(S) = M(S)$ .

Below we give constants for ESP in brackets, they should be less than for NSP. Let  $r_k = \sup\{smt_k(S)/smt(S)\}$ . The bound for the MST-heuristic implies  $r_2 = 2$  [13] ( $r_2 = \frac{2}{\sqrt{3}}$  [7]). As mentioned above,  $r_3 = 5/3$  [15, 16] (conjectured value  $\approx 1.073$  [8]),  $r_4 \leq \frac{3}{2}$  and  $r_8 \leq \frac{4}{3}$  [4]. Moreover,  $r_{2k} \leq 1 + \frac{1}{k}$  [8]. Unfortunately, even when  $k = 4$ , computing optimal  $k$ -restricted Steiner tree is NP-hard [9]. For  $k = 3$ , this problem is still open although V. Arora et al. [2] applied so called backtrack greedy approach and conjectured that it gave polynomial-time approximation scheme.

GGH and RGH approximate  $k$ -restricted Steiner minimal trees. Berman and Ramayer suggested a more complicated way to approximate such trees [5]. They construct a family of *evaluation* heuristics  $B_k$  and bound a performance ratio of  $B_k$  with the following value

$$pr(B_k) \leq r_2 - \sum_{i=3}^k \frac{r_{i-1} - r_i}{i-1}.$$

Knowing bounds for  $r_{2^k}$  we may count an upper bound for  $\lim_{k \rightarrow \infty} pr(B_k)$  which is close to 1.746. Since  $r_k$  is not evaluated for  $k$  unequal to powers of 2, this value might be less. The next step [10] decreases the bound for Berman/Ramayer algorithm by  $\frac{1}{48} = 0.02$ .

For RGH,  $\lim_{k \rightarrow \infty} a_1 = \lim_{k \rightarrow \infty} r_k = 1$ . In the next section we prove that  $a_2 \leq 1 + \ln 2$ . In other words, it induces a polynomial  $(1 + \ln 2)$ -approximation scheme for STP. Again we are not sure that  $1 + \ln 2$  is an exact upper bound for this scheme.

### 3 The performance guarantee

In this section we determine a performance ratio of a relative greedy heuristic  $A_k$  while it approximates a minimal  $k$ -restricted Steiner tree. In other words, we bound the value  $a_2 = a_2(A_k)$ . For brevity, below we omit  $S$  in the following denotations:  $smt(S)$ ,  $smt_k(S)$ ,  $M(S)$ .

**Lemma 1**  $a_2(A_k) \leq 1 + \ln \frac{M}{smt_k}$ .

**Proof.** We need the following denotations. Let  $T$  be the output tree of RGH  $A_k$  and  $T = T_1 \cup \dots \cup T_l$  be a partition of  $T$  into its full components, i.e.  $LIST = \{T_1, \dots, T_l\}$ . We denote the length of  $T_i$  by  $d_i = d(T_i)$ ,  $M_0 = M(S)$ . Recursively,  $M_i = M(S/T_i)$  and

$$m_i = M_{i-1} - M_i \tag{2}$$

So the criterial function (1) equals  $f(T_i) = \frac{d_i}{m_i}$ .

We also partition the optimal  $k$ -restricted Steiner tree  $T^*$  into its full components:  $T^* = \cup T_j^*$ . It is easy to see that

$$f(T^*) \geq \min d(T_j^*) \geq f(T_1),$$

since  $T_1 = \arg \min f(T)$ . So we obtain

$$\frac{d_1}{m_1} \leq \frac{smt_k}{M_0}$$

We may apply the same argument to the  $k$ -restricted Steiner tree after contraction of  $T_1$ .

$$\frac{d_2}{m_2} \leq \frac{smt_k(S/T_1)}{M_1} \leq \frac{smt_k}{M_1}$$

Inductively we obtain

$$\frac{d_i}{m_i} \leq \frac{smt_k}{M_{i-1}}, \quad i = 1, \dots, l \tag{3}$$

Now we apply an analysis technique due to Leighton and Rao [11] to prove Inequality (5). Substituting Equality (2) into (3), we get

$$M_i \leq M_{i-1} \left(1 - \frac{d_i}{smt_k}\right) \tag{4}$$

Unraveling (4), we obtain

$$M_r \leq M_0 \prod_{i=1}^r \left(1 - \frac{d_i}{smt_k}\right).$$

Taking natural logarithm on both sides and simplifying using the approximation  $\ln(1+x) \leq x$ , we obtain

$$\ln \frac{M_0}{M_r} \geq \frac{\sum_{i=1}^r d_i}{smt_k} \quad (5)$$

Note that  $M_l = M_{l-1} - m_l = 0$ , since Step (1) of GCF interrupts when  $M(S) = 0$ . So we can choose  $r$  such that  $M_r > smt_k \geq M_{r+1}$ . We partition  $m_{r+1}$  into two parts  $m_{r+1} = m^* + m'$  such that  $m^* = M_r - smt_k$  and  $m' = smt_k - M_{r+1}$ . We also partition  $d_{r+1}$  in the same proportion, i.e.  $d_{r+1} = d^* + d'$  and  $\frac{d_{r+1}}{m_{r+1}} = \frac{d^*}{m^*} = \frac{d'}{m'}$  ( $d' = 0$  if  $m' = 0$ ). Denote  $M_{r+1}^* = M_r - m^*$ .

These denotations allows us to put down (5) for  $M_{r+1}^*$

$$\ln \frac{M_0}{M_{r+1}^*} \geq \frac{\sum_{i=1}^r d_i + d^*}{smt_k}$$

Note that  $\frac{d_i}{m_i} \leq 1$ . Indeed,  $T_i = \arg \min f(T)$ , but  $f(e) = 1$  for any nonzero edge  $e \in Mst(S)$ . Therefore,

$$d(T) \leq \sum_{i=1}^{r+1} d_i + M_{r+1} = \sum_{i=1}^r d_i + d^* + d' + M_{r+1}$$

Thus, we are ready to finish the proof of Lemma:

$$\begin{aligned} a_2(A_k) &= \frac{d(T)}{smt_k} \leq \\ &\frac{\sum_{i=1}^r d_i + d^*}{smt_k} + \frac{d' + M_{r+1}}{smt_k} \leq \\ \ln \frac{M_0}{M_{r+1}^*} + \frac{m' + M_{r+1}}{smt_k} &= \ln \frac{M_0}{smt_k} + 1 \quad \diamond \end{aligned}$$

To prove Theorem 1 we only need to note that

$$\frac{M}{smt_k} \leq \frac{M}{smt} \quad \diamond$$

## 4 Conclusion

In this paper we introduced a greedy contraction framework for a wide class of heuristics for the Steiner tree problem. A series of new *relative greedy* heuristics embedded in this framework has a better approximation ratio which tends to  $1 + \ln r_2$ , where  $r_2$  is the approximation ratio for the MST-heuristic.

We applied relative greedy heuristics to the Steiner tree problems in networks and in Euclidean plane. The performance guarantee of these heuristics comes close to  $1 + \ln 2 \approx 1.693$  and  $1 + \ln \frac{2}{\sqrt{3}} \approx 1.1438$ , respectively. These values improved the best previously known bounds of 1.746 and 1.1546.

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