ECCCFTP:ftp.eccc.uni-trier.de:/pub/eccc/TR95-030WWW:http://www.eccc.uni-trier.de/eccc/Email:ftpmail@ftp.eccc.uni-trier.de with subject 'help eccc'

New Approximation Algorithms for the Steiner Tree Problems

Marek Karpinski * Alexander Zelikovsky [†]

Abstract

The Steiner tree problem asks for the shortest tree connecting a given set of terminal points in a metric space. We design new approximation algorithms for the Steiner tree problems using a novel technique of choosing Steiner points in dependence on the possible deviation from the optimal solutions. We achieve the best up to now approximation ratios of 1.644 in arbitrary metric and 1.267 in rectilinear plane, respectively.

^{*}Dept. of Computer Science, University of Bonn, 53117 Bonn, and International Computer Science Institute, Berkeley, California. Supported in part by the Leibniz Center for Research in Computer Science, by DFG Grant KA 673/4-1, by the ESPRIT BR Grants 7097 and by ECUS030. Email: marek@cs.uni-bonn.de.

[†]Institute of Mathematics, 277028 Kishinev, Moldova. Research partially supported by Volkswagen Stiftung. Parts of this work were done in Max Planck Institute für Informatik, Saarbrücken. Email: 17azz@mathem.moldova.su.

1 Introduction

We consider a metric space with a distance function d. For any set of *terminal* points S one can efficiently find MST(S), a minimum spanning tree of S. Let mst(S, d) be the cost of this tree in metric d. A Steiner tree is a spanning tree of a superset of the terminal points (the extra points are called Steiner points). It was already observed by Pierre Fermat that the cost of a Steiner tree of S may be smaller than mst(S, d). The Steiner tree problem asks for the Steiner minimum tree, that is, for the least cost Steiner tree. However, finding such a tree is NP-hard for almost all interesting metrics, like Euclidean, rectilinear, Hamming distance, shortest-path distance in a graph etc. Because these problems have many applications, they were subject of extensive research cf [13].

In the last two decades many approximation algorithms for finding Steiner minimum trees have been designed. The quality of an approximation algorithm is measured by its performance ratio (PR): an upper bound of the ratio between the achieved length and the optimal length.

The Network Steiner tree problem (NSP) asks for the Steiner minimum tree for a vertex subset $S \subset V$ of a graph G(V, E, d) with cost function d on edges E. Let |V| = v, |E| = e and |S| = n.

In the rectilinear metric, the distance between two points is the sum of the differences of their x- and y-coordinates. The rectilinear Steiner tree problem (RSP) got recently new importance in the development of techniques for VLSI routing [15].

The most obvious heuristic for the Steiner tree problem approximates a Steiner minimum tree of S with the minimum spanning tree of S. While in all metric spaces the performance ratio of this heuristic is at most 2 (Takahashi and Matsuyama [18]) (it can be implemented for NSP in time $O(e + v \log v)$ (Mehlhorn [16])), Hwang [11, 12] proved that this heuristic in the rectilinear plane has the performance ratio exactly 1.5 and can be implemented in time $O(n \log n)$.

Consideration of k-restricted Steiner trees gave several better heuristics. The main their idea is to decrease the cost of MST(S) adding Steiner points to the initial terminal set. For NSP, the performance ratio of the greedy algorithm (GA) (Zelikovsky [19, 21]) is at most $\frac{11}{6} \approx 1.84$ and PR of Berman-Ramaiyer's heuristic (BR) [2] is at most $\frac{16}{9} \approx 1.78$. Their run-times are $O(v^3)$ and $O(\alpha + v^2n^3)$, respectively (here α means time complexity of finding of all pairs shortest paths). The relative greedy heuristic (RGH) (Zelikovsky [22]) with PR converging to $1 + \ln 2 \approx 1.693$ asymptotically beats BR which PR converges to about 1.734 (Brochers and Du [6]).

In the recent paper Berman *et al.* [3] gave a more precise (than in the first papers [20, 2, 8]) analysis of the performance ratio and runtime of BR for RSP. They proved that its performance ratio is at most $\frac{61}{48} \approx 1.271$. BR can run in $O(n^{1.5})$ time and its parameterized version (PBR) approximates BR in time $O(n \log^2 n)$ [3, 8].

Here we introduce a novel approach based on the notion of relative gain (see Section 2). Now the choice of Steiner points also depends on the possible deviation from the optimal solution. We add new preprocessing phases to the algorithms mentioned above. Combined algorithms achieve better performance ratios in the same order of the runtime.

The table above contains approximation algorithms known before to be the best in respect to performance ratios and orders of runtime and new performance ratios after preprocessing. By $+\epsilon$ we mean existence of an algorithm for any $\epsilon > 0$.

In the next section we provide a synopsis of k-restricted Steiner trees and our approach. In Sections 3 and 4 we describe our preprocessing of RGH and BR.

Problem	Heuristic	Performance Ratio	New PR	Run-time	Reference
NSP	MST	2		$O(v^2)$	[18, 16]
	\mathbf{GA}	$\frac{11}{6} \approx 1.84$		$O(v^3)$	[19, 21]
	BR	$\frac{16}{9} \approx 1.78$	$253/144{pprox}1.757$	$O(v^5)$	[2]
	RGH	$\begin{array}{l} 1 + \ln 2 + \epsilon \\ \approx 1.693 + \epsilon \end{array}$	$pprox$ 1.644 $+\epsilon$	polynomial	[22]
RSP	MST	1.5		$O(n \log n)$	[11, 12]
	BR	$\frac{61}{48} \approx 1.271$	$19/15{pprox}1.267$	$O(n^{1.5})$	[3]
	PBR	$\frac{61}{48} + \epsilon \approx 1.271 + \epsilon$	$pprox$ 1.267 $+\epsilon$	$O(n\log^2 n)$	[3]

2 Gain and Loss of k-Restricted Steiner Trees

2.1 Background

A Steiner tree T of a set of terminals S is *full* if every internal node of T is a *Steiner point*, i.e., not a terminal. If T is not full, it can be decomposed into full Steiner trees for subsets of terminals that overlap only at leaves. Such subtrees are called *full Steiner components* of T [10]. k-trees are full Steiner trees with at most k terminals.

Without loss of generality, we may assume that the metric d on the set of terminals S is the shortest-path distance for the weighted edges D connecting S. This way, MST(S) is the minimum spanning tree of the graph $\langle S, D \rangle$, we denote this tree by MST(D), and its cost with mst(D) or mst(S). If we increase the set of edges D by some extra edges, say forming a set E, the shortest-path distance may decrease; $MST(D \cup E)$ is the minimum spanning tree for the modified metric. For any graph H, d(H) denotes the sum of costs of all edges of H.

Let X(T) be a Steiner tree obtained from a k-tree T by addition of the minimum forest spanning T with the rest of the terminal set S. The cost of this forest equals to $mst(D \cup E(T))$, where E(T) is the set of zero-cost edges between terminals of T. Define a gain of T to be $g(T) = mst(D) - d(X(T)) = mst(D) - mst(D \cup E(T)) - d(T)$. Inductively, the gain of a set of k-trees T_i , i = 1, ..., p, equals to $mst(D) - d(X(\{T_i, i = 1, ..., p\}))$. Let R(T) denote the set of MST-edges substituted with T in the tree X(T). R(T) consists of the edges of the largest cost on the paths in MST(D) connecting pairs of terminals of T [2]. Denote by $m(T) = mst(D) - mst(D \cup E(T))$ the cost of R(T). Thus, g(T) = m(T) - d(T). Note, that addition of any edges to D may only decrease m(T) and the gain of T [2], therefore,

$$g(\{T_i, i = 1, ..., p\}) \le \sum_{i=1}^p g(T_i).$$
(1)

By contraction of T we mean addition of E(T) to D. A greedy algorithm (GA) [19] finds a 3-tree with the biggest gain and contracts it while there are 3-trees with a positive gain. All contracted 3-trees and the rest of MST-edges form the output Steiner tree. The k-restricted relative greedy heuristic (k-RGH) [22] runs similar to GA but maximizing (among all k-trees T) m(T)/d(T) instead of m(T) - d(T). Berman-Ramaiyer [2] suggested a sophisticated generalization of GA for an arbitrary k (k-BR). k-BR processes all *i*-trees, i = 1, ..., k, with a positive gain modifying the set D and forming a stack of *i*-trees chosen. Then it repeatedly pops *i*-trees from the stack re-modifying D and selecting *i*-trees with the current positive gain. The output tree is $X(T_1, ..., T_p)$ for the selected *i*-trees $T_1, ..., T_p$.

To bound PR of GA, k-RGH and k-BR we need the following constants. Let E_k be an arbitrary set of edges such that in $\langle S, D \cup E_k \rangle$ the gain of any k-tree becomes non-positive. We denote by $t_k = t_k(S)$ a supremum of $mst(D \cup E_k)$ over all E_k 's.

The output cost of GA [19] (k = 3) and k-BR [2] is at most

$$t_2 - \sum_{i=3}^k \frac{t_{i-1} - t_i}{i-1} \tag{2}$$

To bound the values t_k , Berman-Ramaiyer [2] introduced the following useful definition. A Steiner tree is *k*-restricted, if every its full component is a *k*-tree. Let $ST_k(S)$ denote a minimal *k*-restricted Steiner tree and $st_k(S)$ denote its cost. This way, $ST_2(S)$ is the minimum spanning tree MST(S). By (1), the gain of any *k*-restricted Steiner tree is nonpositive in $\langle S, D \cup E_k \rangle$, therefore, $st_k(S) \ge t_k(S)$ [2]. These values may not coincide: In the rectilinear plane, for the set $S = \{(\pm 1, 0), (0, \pm 1)\}, st_3(S) = 5$ and $t_3(S) = 4.5$.

A k-Steiner ratio r_k is the supremum of $st_k(S)/s$ over all instances of the Steiner tree problem, where s denotes the cost of the Steiner minimal tree. r_2 (a usual Steiner ratio) equals 2 and 1.5 for NSP and RSP, respectively [18, 11]. For NSP, some r_k were evaluated in [19, 1, 7] and, finally, Brochers and Du [6] proved that for $k = 2^r + l$,

$$r_k = \sup \frac{st_k}{s} = \frac{(r+1)2^r + l}{r2^r + l}.$$
(3)

For the rectilinear metric, $r_k \leq \frac{2k}{2k-1}$ for $r \geq 3$ [2], moreover, for any instance of RSP, $t_2 + t_4 \leq 2.5s$ and $3t_2 + 4t_3 \leq 9s$ [3]. The bounds for t_k and r_k combined with the bound (2) give the performance guarantee of GA and k-BR mentioned in the previous section. It was proved in [22] that the output cost of k-RGH is at most $(1 + \ln(r_2/r_k))r_k$. Since $\lim_{k\to\infty} r_k = 1$, the limit performance ratio of k-RGH for NSP is at most $1 + \ln 2$. Note that the limit performance ratio of k-BR for NSP derived from (2) and (3) is 1.73...

2.2 A New Approach

The algorithms described above try to maximize the total gain. But every time they accept a k-tree, they also accept all its Steiner points. This may increase the cost of the cheapest solution achievable at the current step. The main idea of our approach is to minimize this possible increase.

Let K be a k-tree and V(K) be its Steiner point set. A forest $K' \subset K$ is called *spanning* if for any $v \in V(K)$, there is a path in K' connecting v with S. The cost of the minimum spanning forest in K is called a *loss* of K and denoted by l(K). The main property of the loss of a k-tree is in the following

Lemma 1 Let P be the set of the Steiner points of an r-tree T. Then $t_k(S \cup P) \leq t_k(S) + l(T)$.

Proof. Let $\langle S \cup P, D_P \rangle$ be a complete graph on the set of terminals $S \cup P$ and edges from D_P have costs equal to the shortest-path distances. Let E'_k be an arbitrary set of edges such that $G = \langle S \cup P, D_P \cup E'_k \rangle$ does not contain k-trees with a positive gain. To prove Lemma it is sufficient to show that $mst(G) \leq t_k(S) + l(T)$. For every pair of vertices $u, v \in S$, we add an edge f = (u, v) such that d(f) is equal to the largest cost of an edge on the path in MST(G) between u and v. In the graph G' obtained, we can choose a minimum spanning tree M in which any pair $u, v \in S$ is connected by paths containing only terminals of S. It is proved in [2] that the d(M) = mst(G') = mst(G) and for any k-tree K, the cost of R(K) is the same in G and G'.

Consider a subgraph H of G' induced by the vertex set S. Since MST(H) is a subgraph of M, for any k-tree K, R(K) is the same in G' and H. This implies that g(K) is non-positive in H and $mst(H) \leq t_k(S)$. From the other side, since $S \cup P$ can be spanned with MST(H) and a spanning forest for T, $mst(G) = mst(G') \leq mst(H) + l(T) \leq t_k(S) + l(T) \diamond$

For any $\alpha \geq 0$, the value $g'(\alpha, K) = g(K) - \alpha l(K)$ will be called a α -relative gain of K. Further we omit α if $\alpha = 1$. Similarly to the definition of $t_k(S)$, we define $t^k(\alpha) = t^k(\alpha, S)$ to be a supremum of $mst(D \cup E^k)$ over all edge sets E^k 's such that addition of E^k to D makes the α -relative gain of any k-tree non-positive.

Lemma 2 $t^k(\alpha, S) \leq (1 + \frac{\alpha}{2})st_k(S)$

Proof. Let T_i be a full component of an optimal k-restricted Steiner tree T. We transform T_i to the form of a binary tree by replicating certain internal vertices, so that copies of the same vertex are connected with zero-cost edges.

The loss of T_i can be bounded in the following way. For any inner vertex of T_i , choose the cheapest edge among two edges going to its two children. It is easy to see, that the forest F obtained spans all inner vertices of T_i . d(F) is at most half of $d(T_i)$, since F contains exactly half of all edges of T_i and $T_i - F$ contains longer edges. This means, that $l(T_i) \leq 0.5d(T_i)$.

Let $g(K) \leq \alpha l(K)$ for any k-tree K in $\langle S, D \cup E^k \rangle$. By (1), $mst(D \cup E^k) - d(T) = g(T) \leq \sum_{i=1}^p g(T_i) \leq \sum_{i=1}^p \alpha l(T_i) = 0.5\alpha d(T)$. Therefore, $mst(D \cup E^k) \leq (1 + 0.5\alpha)d(T)$. Since this is true for any E^k , $t^k(\alpha, S) \leq (1 + \frac{\alpha}{2})d(T) = (1 + \frac{\alpha}{2})st_k \diamond$.

Theorem 2 shows that $\lim_{k\to\infty} t^k(\alpha) = (1 + \frac{\alpha}{2})s$. The relative gain of any triple is non-positive, therefore, $t^3 = t_2$. In Sections 5 and 6, we find the tight bounds for t^4 in the case of NSP and RSP, respectively.

Lemma 3 For NSP, $t^4 \leq \frac{15}{8}s$.

Lemma 4 For RSP, $t^4 \leq \frac{7}{5}s$.

The main idea of preprocessing k-BR and k-RGH is to find some k-trees which are good in respect to the relative gain and to add its Steiner points to initial terminal set before running usual k-BR and k-RGH. Using Lemmas 2, 3 and 4, in Sections 3 and 4, we derive the record performance ratios claimed in Introduction.

Theorem 1 For NSP, there is a polynomial-time approximation algorithm with the performance ratio at most $1.644... + \epsilon$ for any $\epsilon > 0$.

Theorem 2 For NSP, there is an 1.757...-approximation algorithm with a runtime $O(\alpha + v^2n^3)$.

Theorem 3 For RSP, for any $\epsilon > 0$, there are $\frac{19}{15}$ and $\frac{19}{15} + \epsilon$ -approximation algorithms with run-times $O(n^{1.5})$ and $O(n \log^2 n)$, respectively.

3 Preprocessing the Relative Greedy Heuristic

We suggest the following generalization of k-RGH (k-RGH(α)): While $mst(D) \neq 0$, find and contract a k-tree T minimizing $p(T) = (d(T) + \alpha l(T))/m(T)$. The union of k-trees T obtained forms the output tree.

Theorem 4 k- $RGH(\alpha)$ finds a tree T such that $d(T) + \alpha l(T) \le (1 + \ln \frac{mst(S)}{t^k(\alpha,S)})t^k(\alpha,S)$.

Proof. Let $T_1, ..., T_a$ be the k-trees chosen by k-RGH (α) including 2-terminal trees (edges). Let M_j denote $mst(D \cup E(T_1) \cup ... \cup E(T_j)), j = 0, ..., a$. Let $p(T_1)D$ be the set of edges D with the cost $p(T_1)$ times the cost of edges of D. Since p(e) = 1 for any MST-edge, $p(T_i) \leq 1$ and $MST(D \cup p(T_1)D) = MST(p(T_1)D)$. By the choice of $T_1, < S, p(T_1)D > \text{does}$ not contain k-trees with the positive α -relative gain. Therefore, $p_1mst(D) = mst(D \cup p_1D) \leq t^k(\alpha)$ and

$$\frac{d(T_1) + \alpha l(T_1)}{m(T_1)} \le \frac{t^k(\alpha)}{M_0}$$

Similarly, after contracting of T_1 and choosing T_2 , we obtain

$$\frac{d(T_2) + \alpha l(T_2)}{m(T_2)} \le \frac{t^k(\alpha)}{M_1}$$

Note, that $M_i = M_{i-1} - m(T_i)$. Inductively we obtain for each $i \ge 1$, $(d(T_i) + \alpha l(T_i))/(M_{i-1} - M_i) \le t^k(\alpha)/M_{i-1}$, or equivalently $M_i \le M_{i-1}(1 - (d(T_i) + \alpha l(T_i))/t^k(\alpha))$. Unraveling these inequalities,

$$M_r \le M_0 \prod_{i=1}^r (1 - \frac{d(T_i) + \alpha l(T_i)}{t^k(\alpha)}).$$

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

$$\frac{\sum_{i=1}^{r} (d(T_i) + \alpha l(T_i))}{t^k(\alpha)} \le \ln \frac{M_0}{M_r}$$

Since $M_{|S|} = 0$, we can choose r such that $M_r > t^k(\alpha, S) \ge M_{r+1}$. We split $d(T_{r+1}) + \alpha l(T_{r+1})$ proportionally by the position of $t^k(\alpha)$ in the interval $[M_{r+1}, M_r]$. We combine the first portion with M_{r+1} to bring this cost up to exactly $t^k(\alpha)$, and combine the second portion with $d(T_r) + \alpha l(T_r)$. We then split $M_r - M_{r+1}$ into the same proportions, and subtract the second portion from M_r so that the last inequality above still holds when we "pretend" that $t^k(\alpha) = M_{r+1}$. We now finish the proof with the sequence of inequalities

$$\frac{\sum_{i=1}^{a} (d(T_i) + \alpha l(T_i))}{t^k(\alpha)}$$

$$\leq \frac{M_{r+1}}{t^k(\alpha)} + \frac{\sum_{i=1}^{r+1} (d(T_i) + \alpha l(T_i))}{t^k(\alpha)}$$

$$\leq 1 + \ln \frac{M_0}{M_{r+1}} = 1 + \ln \frac{mst(S)}{t^k(\alpha, S)} \diamondsuit$$

Now we preprocess k-RGH (k-RGH(0)) with l-RGH(α) in the following way. We run l-RGH(α) obtaining a Steiner tree T and add all Steiner points of T to the initial terminal set S. Then we apply k-RGH to the modified terminal set.

Proof of Theorem 1. Our goal is to obtain the limit performance ratio of k-RGH after preprocessing with l-RGH(α) while $l, k \to \infty$. Denote by S_l the modified terminal set after preprocessing and by s_l the cost of the optimal Steiner tree for S_l . Note that $mst(S_l) = d(T)$.

By Lemma 2 and Theorem 4, while $l \to \infty$, the bound for $(d(T) + \alpha l(T))/s$ converges to

$$B = (1 + \frac{\alpha}{2})(1 + \ln\frac{2}{1 + \frac{\alpha}{2}}).$$
(4)

By Theorem 4 and Lemma 1, the cost of the output of k-RGH applied to S_l is at most

$$(1 + \ln \frac{mst(S_l)}{t_k(S_l)})t_k(S_l)$$

$$\leq (1 + \ln \frac{d(T)}{t_k(S) + l(T)})(t_k(S) + l(T)).$$
(5)

Since $\lim_{k\to\infty} t_k(S) = s$, (4) and (5) imply that the limit output cost is at most

$$(1 + \ln \frac{d(T)}{s + \frac{1}{\alpha}(Bs - d(T))})(s + \frac{1}{\alpha}(Bs - d(T))).$$
(6)

As a function of d(T), (6) has one maximum for d(T) such that

$$\frac{(\alpha+B)s-d(T)}{d(T)} = \ln \frac{\alpha d(T)}{(\alpha+B)s-d(T)}.$$

Denote by $f(\alpha)$ the solution of the equation $x = \ln(\alpha/x)$. Then we obtain the following upper bound for the limit output cost

$$f(\alpha)(1+B/\alpha)$$

The last function has a minimum for $\alpha \approx 0.5$ which is about 1.644... Thus, k-RGH preprocessed with *l*-RGH(0.5) has a limit performance ratio at most 1.644... while $l, k \to \infty$.

4 Improved Version of Berman-Ramaiyer's Algorithm

An *r*-restricted Berman-Ramaiyer's preprocessing (r-BRP) differs from the usual *r*-BR only in the gain function substituted with the relative gain function.

Lemma 5 Let T(r) be an output tree of r-BRP. Then $g'(T(r)) \ge \sum_{i=3}^{r} \frac{t^{i-1}-t^{i}}{i-1}$.

Proof (Sketch). For any *r*-tree K, denote d'(K) = d(K) + l(K). For any *r*-restricted Steiner tree T, $d'(T) = \sum_{i \in A} d'(K_i)$, where $K_i, i \in A$ are the full components of T. Since d'(T) = d(T) for any 2-restricted Steiner tree T, an optimal in respect to d' Steiner tree has a cost at most the minimum spanning tree cost.

r-BRP coincides with *r*-BR applied to the modified cost function d' instead of d. Berman and Ramaiyer [2] proved that the output tree of the usual *r*-BR has a gain at least $\sum_{i=3}^{r} \frac{t_{i-1}-t_i}{i-1}$ (compare with (2)). This proof does not use any properties of the cost function d on k-trees except the properties above. Thus, we may conclude that the same fact is true for *r*-BR applied to the cost function d'.

Since the gain function in respect to d' equals to the relative gain function in respect to d, the relative gain of the the output tree of r-BRP is at least $\sum_{i=3}^{r} \frac{t'_{i-1}-t'_{i}}{i-1}$, where t'_{i} means

the value of t_i in respect to the cost function d'. Lemma follows from the fact that the value t'_i coincides with the value t^i in respect to the function d for any i = 1, ..., r.

Let S_r be the union of the terminal set S with the set of all Steiner points of T(r). Denote by G, L and G' = G - L the total gain, loss and relative gain of T(r), respectively. Then $t_2(S_r) = t_2(S) - G$ and $t_i(S_r) \le t_i(S) + L$ by Lemma 1.

Let bound the cost of the output of k-BR applied to S_r . By (2), it is at most

$$\sum_{i=3}^{k-1} \frac{t_2(S_r) + t_i(S_r)}{(i-1)i} + \frac{t_2(S_r) + t_k(S_r)}{k-1}$$

$$\leq \sum_{i=3}^{k-1} \frac{t_2(S) - G' + t_i(S)}{(i-1)i} + \frac{t_2(S) - G' + t_k(S)}{k-1}$$

$$= \frac{t_2 - G'}{2} - \sum_{i=3}^k \frac{t_{i-1} - t_i}{i-1}$$
(7)

Lemma 5 and (7) imply

Theorem 5 The cost of the output Steiner tree of k-BR preprocessed with r-BRP is at most

$$\frac{t_2}{2} - \sum_{i=3}^k \frac{t_{i-1} - t_i}{i-1} - \frac{1}{2} \sum_{i=3}^r \frac{t^{i-1} - t^i}{i-1} \tag{8}$$

Proof of Theorems 2 and 3. Note that r-BRP has the same order of runtime as r-BR since r-trees with a positive relative gain should have a positive gain and a loss of an r-tree can be found very fast using a greedy algorithm. By Lemma 3 and Theorem 5, 4-BR preprocessed with 4-BRP satisfies Theorem 2.

In the rectilinear metric, the output length of 4-BR preprocessed with 4-BRP can be bounded using Lemma 4 and inequalities (8), $3t_2 + 4t_3 \leq 9s$ and $2t_2 + 2t_4 \leq 5s$. Indeed, this length is at most

$$t_2 - \frac{t_2 - t_3}{2} - \frac{t_2 - t_4}{3} - \frac{1}{2} \frac{t_2 - t^4}{3} = \frac{3}{8}s + \frac{5}{6}s + \frac{7}{120}s = \frac{19}{15}s$$

5 The value of t^4 for NSP

Proof of Lemma 3. Further assume that some terminals are connected with short edges such that $g(K) \leq l(K)$ for any 4-tree K. We may prove Lemma for each full Steiner component separately. We transform such a component to the form of the complete binary tree by replicating certain vertices, so that copies of the same vertex are connected with zero-cost edges. Note that all terminals are leaves of this tree.

Let k be the depth of this tree. We label its vertices with words from $B^* = \{\alpha \in B^* : |\alpha| \le k\}$, where $B = \{0, 1\}$. Let ρ be the root and α have children $\alpha 0$, $\alpha 1$. The set of terminals with the common ancestor α is denoted by α also.

Some more denotations: Let $s = s(\rho)$ denote the cost of the Steiner minimal tree, $t = t(\rho)$ be the cost of MST for the whole terminal set, $s_i(\alpha) = \sum_{|\beta|=i,b\in B} d(\alpha\beta, \alpha\beta b)$, $H = H(\rho) = s_0(\rho) + s_1(\rho)$, $P(\alpha)$ denote the cost of the cheapest path from α to S.



Figure 1: A full component

An average path cost is defined to be

$$\bar{P} = \bar{P}(\rho) = \frac{\sum_{i=1}^{k-1} 2^{k-i} s_i(\rho)}{2^k} = \sum_{i=1}^{k-1} 2^{-i} s_i(\rho)$$

This cost has the following two obvious properties:

$$\bar{P}(\alpha) \ge P(\alpha) \tag{9}$$

$$2\bar{P}(\alpha) = s_0(\alpha) + \bar{P}(\alpha 0) + \bar{P}(\alpha 1).$$
(10)

Since $\bar{P} \geq \frac{H}{4}$, the following inequality is slightly stronger than Lemma.

$$t \le 2s - 2\bar{P} - \frac{s - H}{8} \tag{11}$$

We will prove (11) by induction on k. Indeed, for $k \leq 2$, (11) is trivially true. Let (11) be true for all trees of depth at most k. We will prove it for a tree of depth k + 1 (Fig. 1). Further assume that $s_1(0) \geq s_1(1)$.

Now we partition $s(\rho)$ into five subtrees:

$$s(\rho) = \sum_{\alpha \in A} s(\alpha) + D,$$

where $\alpha \in A = \{000, 001, 01, 1\}$ and $D = s_0(\rho) + s_0(0) + s_0(00)$ (thick lines on Fig. 1).

These five parts correspond to some spanning tree:

$$t(\rho) \le \sum_{\alpha \in A} t(\alpha) + t', \tag{12}$$

where t' is the cost of three cheapest edges connecting four MST for the sets $\alpha \in A$. By induction, inequality (11) holds for every $\alpha \in A$:

$$t(\alpha) \le 2s(\alpha) - 2\bar{P}(\alpha) - \frac{s(\alpha) - H(\alpha)}{8}$$
(13)

Substituting (13) into (12) we obtain

$$t(\rho) - (2s - 2\bar{P} - \frac{s - H}{8}) \le t' + 2\bar{P} + \frac{s - H}{8}$$

$$-2D - 2\sum_{\alpha \in A} \bar{P}(\alpha) - \sum_{\alpha \in A} \frac{s(\alpha) - H(\alpha)}{8}.$$

To prove (11) we show that the RHS of the last inequality is non-positive, i.e,

$$\frac{1}{8} \left(s - H - \sum_{\alpha \in A} (s(\alpha) - H(\alpha)) \right)$$

$$\leq 2D + 2 \sum_{\alpha \in A} \bar{P}(\alpha) - (t' + 2\bar{P})$$
(14)

Claim 1 The RHS of (14) is at least $\bar{P}(0) - d(0, 00)$.

Proof. Consider an arbitrary 4-tree q with Steiner points 0 and 00 and four terminals achievable from 000, 001, 01 and 1, respectively. Note, that $t' \leq t(q)$, where t(q) = d(q)+g(q) is the cost of three corresponding longest edges on paths connecting terminals of q. Let terminals of q be the nearest to the corresponding vertices of A. Since $g(q) \leq l(q) \leq d(0,00) + P(00)$, we obtain

$$t' \le D + \sum_{\alpha \in A} P(\alpha) + d(0, 00) + P(00)$$

Now Claim can be proved strait-forward using the properties (9) and (10) of the average path cost:

$$2D + 2\sum_{\alpha \in A} P(\alpha) - (t' + 2P) \ge$$

$$s_0(0) + s_0(00) + \bar{P}(000) + \bar{P}(001) + \bar{P}(01)$$

$$-P(00) - \bar{P}(0) - d(0,00) \ge \bar{P}(0) - d(0,00) \quad \diamondsuit$$

The LHS of (14) equals to $\frac{1}{8}(D + \sum_{\alpha \in A} H(\alpha) - H) = \frac{1}{8}(s_1(1) + s_0(01) + s_1(01) + s_0(00) + s_1(00) + s_2(00))$. By Claim and our assumption of $s_0(00) + s_0(01) = s_1(0) \ge s_1(1)$, (14) follows from the following inequality

$$\frac{1}{8}(2s_0(01) + s_1(01) + 2s_0(00) + s_1(00) + s_2(00))$$

$$\leq \bar{P}(0) - d(0,00)$$
(15)

Similarly, the corresponding partition of the Steiner minimal tree induced by the 4-tree with Steiner points 0 and 01 implies that it is sufficient to prove

$$\frac{1}{8}(2s_0(00) + s_1(00) + 2s_0(01) + s_1(01) + s_2(01))$$

$$\leq \bar{P}(0) - d(0, 01)$$
(16)

Thus, to prove (11) we may show that one of the inequalities (15) or (16) is true. This follows from the fact that their sum is true. Indeed, summing (15) and (16) we obtain

$$\frac{1}{8}(4s_0(00) + 2s_1(00) + s_2(00) + 4s_0(01) + 2s_1(01) + s_2(01)) \le 2\bar{P}(0) - s_0(0) = \bar{P}(00) + \bar{P}(01),$$

which trivially follows from the definition of the average path cost. \Diamond



Figure 2: Two types of a full component



Figure 3: The partition of the Steiner segment

6 The value of t^4 for RSP

Hwang [11] proved that there is a Steiner minimum tree where every full component has one of the shapes shown in Fig. 2. It was suggested in [3] some partition of a full component into so called Steiner segments. Below we briefly describe this useful technique.

Let a_1, \ldots, a_k and $b_0 = 0, b_1, \ldots, b_k$ be the lengths of horizontal and vertical lines of a full Steiner component F with terminals s_0, \ldots, s_k . The horizontal lines form its spine. Moreover, in case (i) $b_k < b_{k-2}$ holds. In case (ii) assume that $b_k = 0$. Consider the sequences $b_0, b_1, b_3, \ldots, b_{2i+1}, \ldots$ and $b_0, b_2, \ldots, b_{2i}, \ldots$ Let

$$b_{h(0)} = b_0, b_{h(1)}, \dots, b_{h(p+1)} = b_k$$
(17)

be the sequence of local minima of these sequences, i.e. $b_{h(j)-2} \ge b_{h(j)} < b_{h(j)+2}$. If h(p) = k - 1, we exclude the member $b_{h(p)}$ from (17). For the case of h(j + 1) = h(j) + 1, $(j = 1, \ldots, p-1)$, we exclude arbitrarily either $b_{h(j+1)}$ or $b_{h(j)}$. So, we get $h(j+1)-h(j) \ge 3$. The elements of the refined sequence (17) are called *hooks*. Further we assume that a full Steiner tree non-trivially contains at least 4 terminals $(k \ge 4)$. A Steiner segment K is a part of a full Steiner component bounded by two sequential hook terminals. So two neighboring Steiner segments have a common hook. K contains the two furthest terminals below and above the spine called *top* and *bottom*, respectively.

Now we are ready to start the following

Proof of Lemma 4. (Sketch.) Further assume that some terminals are connected with short edges such that $g(K) \leq l(K)$ for any 4-tree K. It is sufficient to prove Lemma for a full Steiner component F with a terminal set Set. Let $F = \bigcup_{i=0}^{k} K_i$ be a partition of F into Steiner segments. Then $d(F) = \sum_{i=0}^{k} d(K_i) - \sum_{i=1}^{k-1} h_i$, where h_i are hooks. Consider some

Steiner segment $K = K_i$ of F with terminal set $S = S_i$, hooks $hl = h_i$ and $hr = h_{i+1}$ and the length s = d(K). Similarly to Section 5, denote the MST-length for a terminal set X by t(X). We intend to prove that

$$t(S) - s \le \frac{2}{5}s - \frac{7}{10}(hl + hr)$$
(18)

This inequality yields Lemma, since then

$$\sum_{i=0}^{k} t(S_i) \le \frac{7}{5} \sum_{i=0}^{k} d(K_i) - \frac{7}{10} \sum_{i=0}^{k-1} (h_i + h_{i+1}) \le \frac{7}{5} (\sum_{i=0}^{k} d(K_i) - \sum_{i=1}^{k-1} h_i) = \frac{7}{5} d(F)$$

Let top of K be to the left of its bottom. We partition S into three parts $S = L \cup C \cup R$, where L is the set of terminals from the left hook till the first before top, C contains all terminals from the the first before top till the next after bottom and R contains ones from the next after bottom till the right hook. Similarly, we partition F into three corresponding parts

$$s = left + center + right,$$

where *center* contains all edges spanning C, and *left* and *right* consists of the rest of the Steiner segment to the left and right of *center* (Fig. 3). Denote by vl and vr the lengths of two vertical lines which bound *center* from the left and the right. Note that K should contain *center*, but *left* and *right* might be empty.

For brevity we consider only the case of *bottom* be the next to top (Fig. 4). For this case we need the following useful

Lemma 6 [3] There are two trees (Fig. 4(i)) Top (dotted lines) and Bot (dashed lines) spanning terminals of K with a total length

$$d(Top) + d(Bot) = 3s - 2(hl + hr) - Rest;$$

Rest sums the lengths of the thin drawn Steiner tree lines.

Lemma 6 says that $t \leq \frac{3}{2}s - \frac{Rest}{2} - (hl + hr)$. It is easy to see that (18) holds if *Rest* is big enough, i.e. $Rest \geq \frac{s}{5} - \frac{3}{5}(hl + hr)$. So further assume that

$$Rest \le \frac{s}{5} - \frac{3}{5}(hl + hr). \tag{19}$$

We may span R and L with the alternative chains (Fig. 3), therefore,

$$t(L) + t(R) \le left + right + Rest - x,$$
(20)

where x is the horizontal edge length of *Rest*.

Let q be the quadruple with terminals from C (Fig. 4 (ii)). The loss of q is at most x plus the length of the shortest among four dotted lines (we may shift the central edge up or down till dashed lines). Lemma assumes that g(q) = t(C) - center is at most

$$l(q) \le x + \frac{center - (2vl + 2vr + x)}{4}$$



Figure 4: top besides bottom: the whole segment (i) and its center (ii)



Figure 5: An instance of RSP with $t^4 = \frac{7}{5}$

$$\leq x + \frac{s - Rest - (hl + hr)}{4} \tag{21}$$

Thus, (18) follows from (19), (20), (21):

$$\begin{split} t(S) - s &= t(C) - center + t(L) - left + t(R) - right \\ &\leq x + \frac{s - Rest - (hl + hr)}{4} + Rest - x \\ &\leq \frac{s}{4} + \frac{3}{4}Rest - \frac{hl + hr}{4} \\ &\leq \frac{s}{4} + \frac{3}{4}(\frac{s}{5} - 3\frac{hl + hr}{5}) - \frac{hl + hr}{4} \\ &= \frac{2}{5}s - \frac{7}{10}(hl + hr) \quad \diamondsuit$$

Note that Lemma 4 cannot be improved. Fig. 5 illustrates the following **Remark 1** The bound of $\frac{7}{5}$ is tight for t^4 in the rectilinear plane.

7 Conclusion and Open Problems

The main open question remaining for the Network Steiner Tree Problem is to compute the exact value of a constant c which separates polynomial approximability from nonapproximability (*NP*-hardness) of this problem. Such a constant c must exist since NSP is *MAX SNP*-complete [5]. We prove that c lies somewhere below 1.644... for that problem. Note that we do not know at the moment whether RSP is also *MAX SNP*-complete, and therefore it could have a polynomial time approximation scheme. At the end a word about achieved heuristics: Our paper shows for the first time that we are able to solve with at most 26.7% error any practical instance of RSP of size, say, up to 10^5 in 1 h, whereas all other known algorithms of the same quality are able to solve RSP only for about 30 points in 24 hours on a SUN3 workstation (see [17]).

References

- P. Berman, V. Ramaiyer. An approximation algorithm for the Steiner tree problem. TR #CS-91-05, The Pennsylvania State University, 1991.
- [2] P. Berman, V. Ramaiyer. Improved approximations for the Steiner tree problem. J. of Algorithms 17: 381-408, 1994.
- [3] P. Berman, U. Fößmeier, M. Karpinski, M. Kaufmann, A. Zelikovsky. Approaching the 5/4-Approximations for Rectilinear Steiner Trees. 3d European Symposium on Algorithms. LNCS 855, 60-71, 1994.
- [4] M. Bern and D. Eppstein. Approximation Algorithms for Geometric Problems. In "Approximation Algorithms for NP-Complete Problems", D. Hochbaum, (ed.), PWS Publications, 1995. (to appear)
- [5] M. Bern and P. Plassmann. The Steiner problems with edge lengths 1 and 2. Inform. Process. Lett. 32: 171-176, 1989.
- [6] A. Borchers and D. Du. The k-Steiner Ratio in Graphs. 27th STOC, 641-649 1995.
- [7] D. Du, Y. Zhang, Q. Feng. On better heuristic for Euclidean Steiner minimum trees. 32nd FOCS, 431-439, 1991.
- [8] U. Fößmeier, M. Kaufmann, A. Zelikovsky. Faster Approximation Algorithms for the Rectilinear Steiner Tree Problem. 4th ISSAC. LNCS 762, 533-542, 1993.
- M. R. Garey, D. S. Johnson. The Rectilinear Steiner Problem is NP-Complete. SIAM J. Appl. Math., 32, 826-834, 1977.
- [10] E. N. Gilbert, H. O. Pollak. Steiner Minimal Trees. SIAM Appl. Math., 16, 1-29, 1968.
- [11] F. K. Hwang. On Steiner Minimal Trees with Rectilinear Distance. SIAM J. Appl. Math., 30, 104-114, 1976.
- [12] F. K. Hwang. An o(n log n) algorithm for rectilinear minimum spanning trees. J. ACM 26, 177–182, 1979.
- [13] F. K. Hwang, D. Richards, P. Winter. The Steiner Tree Problem. Annals of Disc. Math. 53, 1992.
- [14] M. Karpinski, A. Zelikovsky. 1.757 and 1.267-Approximation Algorithms for the Network and Rectilinear Steiner Tree Problems. Technical Report TR-95-010, International Computer Science Institute, Berkeley, 1995.
- [15] B.Korte, H.J.Prömel, A.Steger. Steiner Trees in VLSI-Layouts. In Korte et al.: Paths, Flows and VLSI-Layout, Springer, 1990.
- [16] K.Mehlhorn. A faster approximation algorithm for the Steiner problem in graphs. Inf. Process. Lett. 27: 125–128, 1988.
- [17] J.S.Salowe and D.M.Warme. An exact rectilinear Steiner tree algorithm. Proc. of the Int. Conf. on Comp. Design, 472–475, 1993.
- [18] H. Takahashi, A. Matsuyama. An approximate solution for the Steiner problem in graphs. Math. Japonica, 24: 573-577, 1980.

- [19] A.Zelikovsky. An 11/6-approximation Algorithm for the network Steiner Problem Algorithmica, 9: 463-470, 1993.
- [20] A. Zelikovsky. An 11/8-approximation Algorithm for the Steiner Problem on Networks with Rectilinear Distance. In Coll. Math. Soc. J. Bolyai 60: 733-745, 1992.
- [21] A. Zelikovsky. A faster approximation algorithm for the Steiner Tree Problem in Graphs. Inf. Process. Lett. 46: 79-83, 1993.
- [22] A. Z. Zelikovsky. Better approximation bounds for the network and Euclidean Steiner tree problems. Manuscript, 1993.