A Logarithmic Integrality Gap Bound for Directed Steiner Tree in Quasi-bipartite Graphs

Zachary Friggstad\textsuperscript{1}, Jochen K"onemann\textsuperscript{2}, and Mohammad Shadravan\textsuperscript{3}

\textsuperscript{1}Department of Computing Science, University of Alberta, Edmonton, AB, Canada, T6G 2E8, zacharyf@ualberta.ca
\textsuperscript{2}Department of Combinatorics and Optimization, University of Waterloo, Waterloo, ON, Canada, N2L 3G1, jochen@uwaterloo.ca
\textsuperscript{3}Department of Industrial Engineering and Operations Research, Columbia University, New York, NY, USA, 10027, ms4961@columbia.edu

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Abstract

We demonstrate that the integrality gap of the natural cut-based LP relaxation for the directed Steiner tree problem is $O(\log k)$ in quasi-bipartite graphs with $k$ terminals. Such instances can be seen to generalize set cover, so the integrality gap analysis is tight up to a constant factor. A novel aspect of our approach is that we use the primal-dual method; a technique that is rarely used in designing approximation algorithms for network design problems in directed graphs.

1 Introduction

In an instance of the \textit{directed Steiner tree} (DST) problem, we are given a directed graph $G = (V,E)$, non-negative costs $c_e$ for all $e \in E$, \textit{terminal} nodes $X \subseteq V$, and a root $r \in V$. The remaining nodes in $V - (X \cup \{r\})$ are the \textit{Steiner nodes}. The goal is to find the cheapest collection of edges $F \subseteq E$ such that for every terminal $t \in X$ there is an $r,t$-path using only edges in $F$. Throughout, we let $n$ denote $|V|$ and $k$ denote $|X|$.

If $X \cup \{r\} = V$, then the problem is simply the \textit{minimum-cost arborescence} problem which can be solved efficiently \cite{4}. However, the general case is well-known to be NP-hard. In fact, the problem can be seen to generalize the \textit{set-cover} and \textit{group Steiner tree} problems. The latter cannot be approximated within $O(\log^{2-\epsilon}(n))$ for any constant $\epsilon > 0$ unless $\text{NP} \subseteq \text{DTIME}(n^{\text{polylog}(n)})$ \cite{10}.

For a DST instance $G$, let $\text{OPT}_G$ denote the value of the optimum solution for this instance Say that an instance $G = (V,E)$ of DST with terminals $X$ is $\ell$-\textit{layered} if $V$ can be partitioned as $V_0, V_1, \ldots, V_\ell$ where $V_0 = \{r\}, V_\ell = X$ and every edge $uv \in E$ has $u \in V_i$ and $v \in V_{i+1}$ for some $0 \leq i < \ell$. Zelikovsky showed for any DST instance $G$ and integer $\ell \geq 1$ that we can compute an $\ell$-layered DST instance $H$ in poly$(n,\ell)$ time such that $\text{OPT}_G \leq \text{OPT}_H \leq \ell^{1/\ell} \cdot \text{OPT}_G$ and that a DST solution in $H$ can be efficiently mapped to a DST solution in $G$ with the same cost \cite{2} \cite{17}.

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Charikar et al. [3] exploited this fact and presented an $O(\ell^2 \cdot k^{1/\ell} \cdot \log k)$-approximation with running time \(\text{poly}(n,k^\ell)\) for any integer \(\ell \geq 1\). In particular, this can be used to obtain an $O(\log^3 k)$-approximation in quasi-polynomial time and a polynomial-time $O(k^\epsilon)$-approximation for any constant $\epsilon > 0$. Finding a polynomial-time polylogarithmic approximation remains an important open problem.

For a set of nodes $S$, we let $\delta^{in}(S) = \{uv \in V : u \notin S \text{ and } v \in S\}$ be the set of edges entering $S$. The following is a natural linear programming (LP) relaxation for directed Steiner tree.

\[
\begin{align*}
\min & \quad \sum_{e \in E} c_e x_e \\
\text{s.t.} & \quad x(\delta^{in}(S)) \geq 1 \quad \forall S \subseteq V - r, S \cap X \neq \emptyset \\
& \quad x_e \geq 0 \quad \forall e \in E
\end{align*}
\]

This LP is called a relaxation because of the natural correspondence between feasible solutions to a DST instance $G$ and feasible $\{0,1\}$-integer solutions to the corresponding LP (DST-Primal). Thus, if we let $OPT_{L,P}$ denote the value of an optimum (possibly fractional) solution to LP (DST-Primal), then we have $OPT_{L,P} \leq OPT_G$. For a particular instance $G$ we say the integrality gap is $OPT_G/OPT_{L,P}$; we are interested in placing the smallest possible upper bound on this quantity.

Interestingly, if $|X| = 1$ (the shortest path problem) or $X \cup \{r\} = V$ (the minimum-cost arborescence problem), the extreme points of (DST-Primal) are integral so the integrality gap is 1 ([12] and [4], respectively). However, in the general case Zosin and Khuller showed that (DST-Primal) is not useful for finding polylog($k$)-approximation algorithms for DST [18]. The authors showed that the integrality gap of (DST-Primal) relaxation can, unfortunately, be as bad as $\Omega(\sqrt{k})$, even in instances where $G$ is a 4-layered graph. In their examples, the number of nodes $n$ is exponential in $k$ so the integrality gap may still be $O(\log^c n)$ for some constant $c$.

On the other hand, Rothvoss recently showed that applying $O(l)$ rounds of the semidefinite programming Lasserre hierarchy to (the flow-based extended formulation of) (DST-Primal) yields an SDP with integrality gap $O(\ell \cdot \log k)$ for $\ell$-layered instances [14]. Subsequently, Friggstad et al. [6] showed similar results for the weaker Sherali-Adams and Lovász-Schrijver linear programming hierarchies.

In this paper we consider the class of quasi-bipartite DST instances. An instance of DST is quasi-bipartite if the Steiner nodes $V \setminus (X \cup \{r\})$ form an independent set (i.e., no directed edge has both endpoints in $V \setminus (X \cup \{r\})$). Such instances still capture the set cover problem, and thus do not admit an $(1 - \epsilon) \ln k$-approximation for any constant $\epsilon > 0$ unless $P = \text{NP}$ [13] 5]. Furthermore, it is straightforward to adapt known integrality gap constructions for set cover (e.g. [16]) to show that the integrality gap of (DST-Primal) can be as bad as $(1 - o(1)) \cdot \ln k$ in some instances. Hibi and Fujito [11] give an $O(\log k)$-approximation for quasi-bipartite instances of DST, but do not provide any integrality gap bounds.

Quasi-bipartite instances have been well-studied in the context of undirected Steiner trees. The class of graphs was first introduced by Rajagopalan and Vazirani [13] who studied the integrality gap of (DST-Primal) for the bidirected map of the given undirected Steiner tree instances. Currently, the best approximation for quasi-bipartite instances of undirected Steiner tree is $\frac{73}{60}$ by Goemans et al. [3] who also bound the integrality gap of the bidirected cut relaxation by the same quantity. This is the same LP relaxation as (DST-Primal), applied to the directed graph obtained by replacing each undirected edge $\{u,v\}$ with the two directed edges $uv$ and $vu$. This is a slight improvement over a prior $\left(\frac{73}{60} + \epsilon\right)$-approximation for any constant $\epsilon > 0$ by Byrka et al. [1].

The best approximation for general instances of undirected Steiner tree is $\ln(4) + \epsilon$ for any constant $\epsilon > 0$ [1]. However, the best known upper bound on the integrality gap of the bidirected cut relaxation for non-quasi-bipartite instances is only 2; it is an open problem to determine if this integrality gap is a constant-factor better than 2.
1.1 Our contributions

Our main result is the following. Let $H_n = \sum_{i=1}^{n} 1/i = O(\log n)$ be the $n$th harmonic number.

**Theorem 1.** The integrality gap of LP \((\text{DST-Primal})\) is at most $2H_k = O(\log k)$ in quasi-bipartite graphs with $k$ terminals. Furthermore, a Steiner tree with cost at most $2H_k \cdot \text{OPT}_{LP}$ can be constructed in polynomial time.

As noted above, Theorem 1 is asymptotically tight since any of the well-known $\Omega(\log k)$ integrality gap constructions for set cover instances with $k$ items translate directly to an integrality gap lower bound for \((\text{DST-Primal})\), using the usual reduction from set cover to 2-layered quasi-bipartite instances of directed Steiner tree.

This integrality gap bound asymptotically matches the approximation guarantee proven by Hibi and Fujito for quasi-bipartite DST instances \([11]\). We remark that their approach is unlikely to give any integrality gap bounds for \((\text{DST-Primal})\) because they iteratively choose low-density full Steiner trees in the same spirit as \([3]\) and give an $O(\ell \cdot \log k)$-approximation for finding the optimum DST solution $T$ that does not contain a path with $\geq \ell$ Steiner nodes $V \setminus (X \cup \{r\})$. In particular, their approach will also find an $O(\log k)$-approximation to the optimum DST solution in 4-layered graphs and we know the integrality gap in some 4-layered instances is $\Omega(\sqrt{k})$ \([18]\).

We prove Theorem 1 by constructing a directed Steiner tree in an iterative manner. An iteration starts with a partial Steiner tree (see Definition 1 below), which consists of multiple directed components containing the terminals in $X$. Then a set of arcs are purchased to augment this partial solution to one with fewer directed components. These arcs are discovered through a primal-dual moat growing procedure; a feasible solution for the dual \((\text{DST-Primal})\) is constructed and the cost of the purchased arcs can be bounded using this dual solution.

While the primal-dual technique has been very successful for undirected network design problems (e.g., see \([9]\)), far fewer success stories are known in directed domains. Examples include a primal-dual interpretation of Dijkstra’s shortest path algorithm (e.g., see Chapter 5.4 of \([12]\)), and Edmonds’ \([4]\) algorithm for minimum-cost arborescences. In both cases, the special structure of the problem is instrumental in the primal-dual construction. One issue arising in the implementation of primal-dual approaches for directed network design problems appears to be a certain overlap in the moat structure maintained by these algorithms. We are able to handle this difficulty here by exploiting the quasi-bipartite nature of our instances.

2 The integrality gap bound

2.1 Preliminaries and definitions

We now present an algorithmic proof of Theorem 1. As we will follow a primal-dual strategy, we first present the LP dual of \((\text{DST-Primal})\).

\[
\begin{align*}
\max \quad & \sum_{S} yS \\
\text{s.t.} \quad & \sum_{S : e \in \delta^\mu(S)} yS \leq c_e \forall e \in E \\
\quad & y \geq 0
\end{align*}
\]

In \((\text{DST-Dual})\), the sums range only over sets of nodes $S$ such that $S \subseteq V - r$ and $S \cap X \neq \emptyset$.

Our algorithm builds up partial solutions, which are defined as follows.
Definition 1. A partial Steiner tree is a tuple \( T = (\{B_i, h_i, F_i\}_{i=0}^{\ell}, \bar{B}) \) where, for each \( 0 \leq i \leq \ell \), \( B_i \) is a subset of nodes, \( h_i \in B_i \), and \( F_i \) is a subset of edges with endpoints only in \( B_i \) such that the following hold.

- The sets \( B_0, B_1, \ldots, B_\ell, \bar{B} \) form a partition \( V \).
- \( \bar{B} \subseteq V - X - r \) (i.e. \( \bar{B} \) is a subset of Steiner nodes).
- \( h_0 = r \) and \( h_i \in X \) for each \( 1 \leq i \leq \ell \).
- For every \( 0 \leq i \leq \ell \) and every \( v \in B_i \), \( F_i \) contains an \( h_i, v \)-path.

We say that \( \bar{B} \) is the set of free Steiner nodes in \( T \) and that \( h_i \) is the head of \( B_i \) for each \( 0 \leq i \leq \ell \). The edges of \( T \), denoted by \( E(T) \), are simply \( \bigcup_{i=0}^{\ell} F_i \). We say that \( B_0, \ldots, B_\ell \) are the components of \( T \) where \( B_0 \) is the root component and \( B_1, \ldots, B_\ell \) are the non-root components.

Figure 1 illustrates a partial Steiner tree. Note that if \( T \) is a partial Steiner tree with \( \ell = 0 \) non-root components, then \( E(T) \) is in fact a feasible DST solution.

Finally, for a subset of edges \( F \) we let \( \text{cost}(F) = \sum_{e \in F} c_e \).

Figure 1: A partial Steiner tree with \( \ell = 3 \) non-root components (the root is pictured at the top). The only edges shown are those in some \( F_i \). The white circles are the heads of the various sets \( B_i \) and the black circles are terminals that are not heads of any components. The squares outside of the components are the free Steiner nodes \( \bar{B} \). Note, in particular, that each head can reach every node in its respective component. We do not require each \( F_i \) to be a minimal set of edges with this property.
2.2 High-level approach

Our algorithm builds up partial Steiner trees in an iterative manner while ensuring that the cost does not increase by a significant amount between iterations. Specifically, we prove the following lemma in Section 3. Recall that $OPT_{LP}$ refers to the optimum solution value for $(\text{DST-Primal}).$

**Lemma 1.** Given a partial Steiner tree $T$ with $\ell \geq 1$ non-root components, there is a polynomial-time algorithm that finds a partial Steiner tree $T'$ with $\ell' < \ell$ non-root components such that

$$\text{cost}(E(T')) \leq \text{cost}(E(T)) + 2 \cdot OPT_{LP} \cdot \frac{\ell - \ell'}{\ell}.$$ 

Theorem 1 follows from Lemma 1 in a standard way.

**Proof of Theorem 1** 

Initialize a partial Steiner tree $T_k$ with $k$ non-root components as follows. Let $B = \{r\}$, and $F_0 = \emptyset$. Furthermore, label the terminals as $t_1, \ldots, t_k \in X$ and for each $1 \leq i \leq k$ let $B_i = \{t_i\}, h_i = t_i$ and $F_i = \emptyset$. Note that $\text{cost}(E(T_k)) = 0$.

Iterate Lemma 1 to obtain a sequence of partial Steiner trees $T_0, T_1, T_2, \ldots, T_a$ where $T_{\ell_i}$ has $\ell_i$ non-root components such that $k = \ell_0 > \ell_1 > \ldots > \ell_a = 0$ and

$$\text{cost}(E(T_{i+1})) \leq \text{cost}(E(T_i)) + 2 \cdot OPT_{LP} \cdot \frac{\ell_i - \ell_{i+1}}{\ell_i}$$

for each $0 \leq i < a$. Return $E(T_a)$ as the final Steiner tree.

That $E(T_a)$ can be found efficiently follows simply because we are iterating the efficient algorithm from Lemma 1 at most $k$ times. The cost of this Steiner tree can be bounded as follows.

$$\text{cost}(E(T_a)) \leq 2 \cdot OPT_{LP} \cdot \sum_{i=0}^{a-1} \frac{\ell_i - \ell_{i+1}}{\ell_i} = 2 \cdot OPT_{LP} \cdot \sum_{j=0}^{a-1} \sum_{i=j+1}^{a} \frac{1}{\ell_i} \leq 2 \cdot OPT_{LP} \cdot \sum_{j=0}^{k} \frac{1}{\ell_j} = 2 \cdot OPT_{LP} \cdot H_k.$$

The idea presented above resembles one proposed by Guha et al. [7] for bounding the integrality gap of a natural relaxation for undirected node-weighted Steiner tree by $O(\log k)$ [7]. Like our approach, Guha et al. also build a solution incrementally. In each phase of the algorithm, the authors reduce the number of connected components of a partial solution by adding vertices whose cost is charged carefully to the value of a dual LP solution that the algorithm constructs simultaneously.

3 A primal-dual proof of Lemma 1

Consider a given partial Steiner tree $T = (\{B_i, h_i, F_i\}_{i=0}^\ell, \bar{B})$ with $\ell \geq 1$ non-root components. Lemma 1 promises a partial Steiner tree $T'$ with $\ell' < \ell$ non-root components with $\text{cost}(E(T')) \leq \text{cost}(E(T)) + 2 \cdot OPT_{LP} \cdot \frac{\ell - \ell'}{\ell}$. In this section we will present an algorithm that *augments* forest $T$ in the sense that it computes a set of edges to *add* to $T$. The proof presented here is constructive: we will design a primal-dual algorithm that maintains a feasible dual solution for $(\text{DST-Dual})$, and uses the structure of this solution to guide the process of adding edges to $T$. 

5
3.1 The algorithm

For any two nodes \( u, v \in V \), let \( d(u, v) \) be the cost of the cheapest \( u, v \)-path in \( G \). More generally, for a subset \( \emptyset \subseteq S \subseteq V \) and a node \( v \in V \) we let \( d(S, v) = \min_{u \in S} d(u, v) \). We will assume that for every \( 0 \leq i \leq \ell \) and \( 1 \leq j \leq \ell, j \neq i \) that \( d(B_i, h_j) > 0 \) as otherwise, we could merge \( B_i \) and \( B_j \) by adding the 0-cost \( B_i, h_j \)-path to \( T \).

The usual conventions of primal-dual algorithms will be adopted. We think of such an algorithm as a continuous process that increases the value of some dual variables over time. At time \( t = 0 \), all dual variables are initialized to a value of 0. At any point in time, exactly \( \ell \) dual variables will be raised at a rate of one unit per time unit. We will use \( \Delta \) for the time at which the algorithm terminates. As is customary, we will say that an edge \( e \) goes tight if the dual constraint for \( e \) becomes tight as the dual variables are being increased. When an edge goes tight, we will perform some updates to the various sets being maintained by the algorithm. Again, the standard convention applies that if multiple edges go tight at the same time, then we process them in any order.

Algorithm 1\[\text{Algorithm 1}\]describes the main subroutine that augments the partial Steiner tree \( T \) to one with fewer components. It maintains a collection of moats \( M_i \subseteq V - \{r\} \) and edges \( F_i \) for each \( 1 \leq i \leq \ell \), while ensuring that the dual solution \( y \) it grows remains feasible. Mainly to aid notation, our algorithm will maintain a so called virtual body \( \beta_i \) for all \( 0 \leq i \leq \ell \) such that \( B_i \subseteq \beta_i \subseteq B_i \cup B \). We will ensure that each \( v \in \overline{B} \cap \beta_i \) has a mate \( u \in B_i \) such that the edge \( uv \) has cost no more than \( \Delta \). For notational convenience, we will let \( \beta_0 = B_0 \) be the virtual body of the root component. The algorithm will not grow a moat around the root since dual variables do not exist for sets containing the root.

Our algorithm will ensure that moats are pairwise terminal-disjoint. In fact, we ensure that any two moats may only intersect in \( B \). Terminal-disjointness together with the quasi-bipartite structure of the input graph will allow us to charge the cost of arcs added in the augmentation process to the duals grown.

An intuitive overview of our process is the following. At any time \( t \geq 0 \), the moats \( M_i \) will consist of all nodes \( v \) with \( d(v, h_i) \leq t \). The moats \( M_i \) will be grown until, at some time \( \Delta \), for at least one pair \( i, j \) with \( i \neq j \), there is a tight path connecting \( \beta_j \) to \( h_i \). At this point the algorithm stops, and adds a carefully chosen collection of tight arcs to the partial Steiner tree that merges \( B_j \) and \( B_i \) (and potentially other components). Crucially, the cost of the added arcs will be charged to the value of the dual solution grown around the merged components.

Due the structure of quasi-bipartite graphs, we are able to ensure that in each step of the algorithm the active moats pay for at most one arc that is ultimately bought to form \( T' \). Also, if \( T' \) has \( \ell' < \ell \) non-root components then each arc was paid for by moats around at most \( \ell - \ell' + 1 \leq 2(\ell - \ell') \) different heads. So, the total cost of all purchased arcs is at most \( 2(\ell - \ell') \cdot \Delta \). Finally, the total dual grown is \( \ell \cdot \Delta \), which is \( \leq OPT_{LP} \) due to feasibility, so the cost of the edges bought can be bounded by \( 2(\ell - \ell') \cdot OPT_{LP} \).

3.2 Algorithm and invariants

Now we will be more precise. The primal-dual procedure is presented in Algorithm 1. The following invariants will be maintained at any time \( 0 \leq t \leq \Delta \) during the execution of Algorithm 1.

1. For each \( 1 \leq i \leq \ell \), \( h_i \in M_i \) and \( M_i \subseteq V - \{r\} \) (so there is a variable \( y_{M_i} \) in the dual).
2. For each \( 1 \leq i \leq \ell \), \( M_i = \{v \in V : d(v, h_i) < t\} \cup S \) where \( S \subseteq \{v \in V : d(v, h_i) = t\} \).
3. \( M_i \cap M_j \subseteq \overline{B} \) and both \( \beta_i \cap \beta_j = M_i \cap M_j = \emptyset \) for distinct \( 0 \leq i, j \leq \ell \).
4. For each \( 1 \leq i \leq \ell \) we have \( B_i \subseteq \beta_i \subseteq B_i \cup \overline{B} \). Furthermore, for each \( v \in \beta_i - B_i \) there is a mate \( u \in B_i \) such that \( uv \in E \) and \( c_{uv} \leq t \).
5. \( y \) is feasible for LP (DST-Dual) with value exactly \( \ell \cdot t \).

These concepts are illustrated in Figure 2.

Figure 2: The moats around the two partial Steiner trees are depicted by the gray circles. The dashed edges are those bought by the first moat and the solid edges are those bought by the second moat. Note the moats only intersect in \( \overline{B} \) (in particular, \( v \) is the only lying in both moats). Also, \( u \) lies in the virtual body for the left partial Steiner tree and the dashed arc entering \( u \) is coming from its mate. The edges \( F_i \) from the original partial Steiner trees are not shown. Observe that if any edge entering \( v \) goes tight then it must be from either \( r \) or some terminal (because \( G \) is quasi-bipartite). This would allow us to merge at least one partial Steiner tree into the body of another.

**Algorithm 1** Dual Growing Procedure

1: \( M_i \leftarrow \{ v \in \hat{V} : d(v, h_i) = 0 \}, 1 \leq i \leq \ell \)
2: \( \beta_i \leftarrow B_i \) for \( 0 \leq i \leq \ell \)
3: \( y \leftarrow 0 \)
4: Raise \( y_{M_i} \) uniformly for each moat \( M_i \) until some edge \( uv \) goes tight
5: if \( u \in \beta_j \) for some \( 0 \leq j \leq \ell \) and \( v \in M_{i'} \) for some \( i' \neq j \) then
6: return the partial Steiner tree \( \mathcal{T}' \) described in Lemma 3
7: else
8: Let \( M_i \) be the unique moat with \( uv \in \delta^{in}(M_i) \) \( \triangleright \text{c.f. Proposition 1} \)
9: \( M_i \leftarrow M_i \cup \{ u \} \)
10: if \( u \in \beta_i \) then
11: \( \beta_i \leftarrow \beta_i \cup \{ v \} \)
12: go to Step 4

3.3 Invariant analysis

**Lemma 2.** Invariants 1-5 are maintained by Algorithm \( \square \) until the condition in the if statement in Step 5 is true. Furthermore, the algorithm terminates in \( O(n \cdot k) \) iterations.

**Proof.** Clearly the invariants are true after the initialization steps (at time \( t = 0 \)), given that \( d(B_i, h_i) > 0 \) for any \( i \neq j \). To see why Algorithm \( \square \) terminates in a polynomial number of iterations, note that each
iteration increases the size some moat by 1 and does not decrease the size of any moats. So after at most $kn$ iterations some moat will grow to include the virtual body of another moat, at which point the algorithm stops.

Assume now that the invariants are true at some point just before Step (4) is executed and that the condition in Step (5) is false after Step (4) finishes. We will show that the invariants continue to hold just before the next iteration starts. We let $uv$ denote the edge that went tight that is considered in Step (4). We also let $t$ denote the total time the algorithm has executed (i.e. grown moats) up to this point.

Before proceeding with our proof, we exhibit the following useful fact. In what follows, let $X_t$ denote the edge that went tight that is considered in Step (4). We also let $t'$ denote the time the algorithm has executed (i.e. grown moats) up to this point.

Assume now that the invariants are true at some point just before Step (4) is executed and that the terminating condition in Step (5) would have been satisfied as $u \notin X_t$ or $j' \neq j$, then the terminating condition in Step (5) would have been satisfied as $u \notin X_t$ or $j' \neq j$. A contradiction.

Following Proposition 1, we let $i$ be the unique index such that $uv \in \delta^i(M_i)$ as in Step (8).

Invariant 1

First note that $M_i$ never loses vertices during the algorithm’s execution, and it therefore always contains head vertex $h_i$. Also, vertex $u$ is not part of $B_0$ as otherwise the algorithm would have terminated in Step (5). Hence $M_i \cup \{u\}$ also does not contain the root node $r$.

Invariant 2

This is just a reinterpretation of Dijkstra’s algorithm in the primal-dual framework (e.g. Chapter 5.4 of [12]), coupled with the fact that no edge considered in Step (4) in some iteration crosses more than one moat at any given time (Proposition 1).

Invariant 3

Suppose $(M_i \cup \{u\}) \cap M_j \subseteq \bar{B}$ for some $i \neq j$. This implies $u \in M_j \setminus \bar{B}$, and hence $u \in B_j \subseteq \beta_j$. Thus, the termination condition in Step (5) was satisfied and the algorithm should have terminated; contradiction.

If $v$ is not added to $\beta_i$, and thus $\beta_i$ remains unchanged, $\beta_i \cap \beta_j = M_j \cap \beta_i = \emptyset$ continues to hold for $j \neq i$. We also must have that $(M_i \cup \{u\}) \cap \beta_j = \emptyset$ for $i \neq j$, as otherwise $u \in \beta_j$ and this would violate the termination condition in Step (5).

Now suppose that $v$ is added to $\beta_i$. Then for $j \neq i$ we still have $(\beta_i \cup \{v\}) \cap \beta_j = \emptyset$ as otherwise $v \in \beta_j$ which contradicts $v \in M_i$ and the fact that Invariant 3 holds at the start of this iteration. We also have that $M_j \cap (\beta_i \cup \{v\}) = \emptyset$ as otherwise $v \in M_j$. But this would mean that $u \in M_j$ as well by Proposition 1. We established above that $(M_i \cup \{u\}) \cap M_j \subseteq \bar{B}$. However, $(\{u, v\}) \subseteq (M_i \cup \{u\}) \cap M_j \subseteq \bar{B}$ contradicts the fact that $G$ is quasi-bipartite.

Invariant 4

That $B_i \subseteq \beta_i$ is clear simply because we only add nodes to the sets $\beta_i$. Suppose now that $v$ is added to $\beta_i$. In this case, $v \notin B_i$ as $B_i \subseteq \beta_i$ from the start. We claim that $v$ can also not be part of $B_j$ for some $j \neq i$, since otherwise $\emptyset \neq \emptyset \cap M_i \subseteq \beta_j \cap M_i$, contradicting Invariant 3. Hence $v \in \bar{B}$. Note that the quasi-bipartiteness of $G$ implies that $u \in X$, and hence $u \in B_i$. Proposition 1 finally implies that only the moats crossed by $uv$ are moats around $i$, so since the algorithm only grows one moat around $i$ at any time we have $c_{uv} \leq t$, and this completes the proof of Invariant 4.
Lemma 3
When Step \(4\) goes tight and was considered in Step \((5)\).

\[c \text{ to } \in \epsilon \text{ for some } w\]

Proof. Let \(T\) be the sum of the dual variables, then the value of the dual is just \(\ell\) times the total time spent growing dual variables.

3.4 Augmenting \(\mathcal{T}\)

To complete the final detail in the description of the algorithm, we now show how to construct the partial Steiner tree after Step \((5)\) has been reached. Lemma \(2\) shows that Invariants 1 through 5 hold just before Step \((4)\) in the final iteration. Say the final iteration executes for \(\delta\) time units and that \(uv\) is the edge that goes tight and was considered in Step \((5)\).

Lemma 3. When Step \((6)\) is reached in Algorithm \(1\) we can efficiently find a partial Steiner tree \(\mathcal{T}'\) with \(\ell' < \ell\) non-root components such that \(\text{cost}(E(\mathcal{T}')) \leq \text{cost}(E(pt)) + 2\frac{\ell - \ell'}{\ell} \cdot \text{OPT}_{LP}\).

Proof. Let \(j\) be the unique index such that \(u \in \beta_j\) at time \(\Delta\). There is exactly one such \(j\) because \(\beta_i \cap \beta_j = \emptyset\) for \(i \neq j\) is ensured by the invariants. Next, let \(J = \{i' \neq j : v \in M_{i'}\}\) and note that \(J\) consists of all indices \(i'\) (except, perhaps, \(j\)) such that \(uv \in \delta^m(M_{i'}).\) By the termination condition, \(J \neq \emptyset\). Vertex \(u\) lies in \(\beta_j\) by definition. If \(u \notin B_j\) then we let \(w\) be the mate of \(u\) as defined in Invariant 4. Otherwise, if \(w \in B_j\), we let \(w = u\).

For notational convenience, we let \(P_j\) be the path consisting of the single edge \(wu\) (or just the trivial path with no edges if \(w = u\)). In either case, say cost of \(P_j\) is \(\Delta - \epsilon_j\) where \(\epsilon_j \geq 0\) (c.f. Invariant 4). For each \(i' \in J\), let \(P_{i'}\) be a shortest \(v, h_{i'}\)-path. Invariant 2 implies that

\[c(P_{i'}) = \Delta - \epsilon_{i'},\]

for some \(\epsilon_{i'} \geq 0\). Observe also that the tightness of \(uv\) at time \(\Delta\) and the definition of \(J\) imply that

\[\sum_{i' \in J \cup \{j\}} \epsilon_{i'} \geq c_{uv}.\]

In fact, precisely a \(\epsilon_{i'}\)-value of the dual variables for \(i' \neq j\) contribute to \(c_{uv}\); the contribution of \(j\)'s variables to \(c_{uv}\) is at most \(\epsilon_j\).

Construct a partial Steiner tree \(\mathcal{T}'\) obtained from \(\mathcal{T}\) and Algorithm \(1\) as follows.

- The sets \(B_{j'}, F_{j'}\) and head \(h_{j'}\) are unchanged for all \(j' \notin J \cup \{j\}\).
- Replace the components \(\{B_{i'}\}_{i' \in J \cup \{j\}}\) with a component \(B := \bigcup_{i' \in J \cup \{j\}} (B_{i'} \cup V(P_{i'}))\) having head \(h := h_j\). The edges of this component in \(\mathcal{T}'\) are \(F := \bigcup_{i' \in J \cup \{j\}} (F_{i'} \cup E(P_{i'})) \cup \{uv\}\).
- The free Steiner nodes \(B'\) of \(\mathcal{T}'\) are the Steiner nodes not contained in any of these components.

Namely, \(B'\) consists of those nodes in \(B\) that are not contained on any path \(P_{i'}, i' \in J \cup \{j\}\).

We show that Steiner tree \(\mathcal{T}'\) as constructed above satisfies the conditions stated in Lemma \(1\). We first verify that \(\mathcal{T}'\) as constructed above is indeed a valid partial Steiner tree. Clearly the new sets \(B', \{B_i\}_{i \notin J + j}\) and \(B\) partition \(V\) and \(B'\) is a subset of Steiner nodes.

Note that if \(0 \in J \cup \{j\}\) in the above construction, then \(j = 0\) because no moat contains \(r\). Thus, if \(B_0\) is replaced when \(B\) is constructed, then \(r\) is the head of this new component.
Next, consider any $b \in B$. If $b \in B_j$ then there is an $h_j, b$-path in $F_j \subseteq F$. If $b \in B_{i'}$, $i' \neq j$ then it can be reached from $h_j$ in $(B, F)$ as follows. Follow the $h_j, w$-path in $F_j$, then the $w, u$ path $P_j$, cross the edge $uv$, follow $P_{i'}$ to reach $h_{i'}$, and finally follow the $h_{i'}, b$-path in $F_{i'}$. Finally, if $b \notin B_{i'}$ for any $i' \in J + j$ then $b$ lies on some path $P_{i'}$, in which case it can be reached in a similar way.

It is also clear that $E(T) \subseteq E(T')$ and that the number of non-root components in $T'$ is $\ell - |J| < \ell$.

Also, $\delta_i - \gamma_i \leq \epsilon_i$. It now easily follows from (3) and (4) that

$$\sum_{i' \in J \cup \{j\}} \delta_i - \gamma_i \leq \epsilon_i.$$  

The last bound follows because the feasible dual we have grown has value $\ell \cdot \Delta \leq \OPT_{LP}$. Let $\ell' = \ell - |J|$ be the number of non-root components in $T'$. Conclude by observing $|J| + 1 = \ell - \ell' + 1 \leq 2(\ell - \ell')$.

To wrap things up, executing Algorithm[1] and constructing the partial Steiner tree as in Lemma[3] yields the partial Steiner tree that is promised by Lemma[1].

\section{Conclusion}

We have shown that the integrality gap of LP relaxation [DST-Primal] is $O(\log k)$ in quasi-bipartite instances of directed Steiner tree. The gap is known to be $\Omega(\sqrt{k})$ in 4-layered instances [18] and $O(\log k)$ in 3-layered instances [6]. Since quasi-bipartite graphs are a generalization 2-layered instances, it is natural to ask if there is a generalization of 3-layered instances which has an $O(\log k)$ or even $o(\sqrt{k})$ integrality gap.

One possible generalization of 3-layered graphs would be when the subgraph of $G$ induced by the Steiner nodes does not have a node with both positive indegree and positive outdegree. None of the known results on directed Steiner tree suggest such instances have a bad gap.

Even when restricted to 3-layered graphs, a straightforward adaptation of our algorithm that grow moats around the partial Steiner tree heads until some partial Steiner trees absorbs another fails to grow a sufficiently large dual to pay for the augmentation within any reasonable factor. A new idea is needed.

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