

An Improved Approximation Algorithm for the Minimum k -Edge Connected Multi-Subgraph Problem

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Abstract

We give a randomized $1 + \sqrt{\frac{8 \ln k}{k}}$ -approximation algorithm for the minimum k -edge connected spanning multi-subgraph problem, k -ECSM.

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1 Introduction

In an instance of the minimum k -edge connected subgraph problem, or k -ECSS, we are given an (undirected) graph $G = (V, E)$ with $n := |V|$ vertices and a cost function $c : E \rightarrow \mathbb{R}_{\geq 0}$, and we want to choose a minimum cost set of edges $F \subseteq E$ such that the subgraph (V, F) is k -edge connected. In its most general form, k -ECSS generalizes several extensively-studied problems in network design such as tree augmentation or cactus augmentation. The k -edge-connected *multi*-subgraph problem, k -ECSM, is a close variant of k -ECSS in which we want to choose a k -edge-connected *multi*-subgraph of G of minimum cost, i.e., we can choose an edge $e \in E$ multiple times. It turns out that one can assume without loss of generality that the cost function c in k -ECSM is a metric, i.e., for any three vertices $x, y, z \in V$, we have $c(x, z) \leq c(x, y) + c(y, z)$.

Around four decades ago, Fredrickson and Jájá [FJ81; FJ82] designed a 2-approximation algorithm for k -ECSS and a $3/2$ -approximation algorithm for k -ECSM. The latter essentially follows by a reduction to the well-known Christofides-Serdyukov approximation algorithm for the traveling salesperson problem (TSP). Over the last four decades, despite a number of papers on the problem [CT00; KR96; Kar99; Gab05; GG08; Gab+09; Pri11; LOS12], the aforementioned approximation factors were only improved in the cases where the underlying graph is unweighted or $k \gg \log n$. Most notably, Gabow, Goemans, Tardos and Williamson [Gab+09] showed that if the graph G is unweighted then k -ECSS and k -ECSM admit $1 + 2/k$ approximation algorithms, i.e., as $k \rightarrow \infty$ the approximation factor approaches 1. The special case of k -ECSM where $k = 2$ received significant attention and better than $3/2$ -approximation algorithms were designed for special cases [CR98; BFS16; SV14; Boy+20].

Motivated by [Gab+09], Pritchard posed the following conjecture:

Conjecture 1.1 ([Pri11]). *The k -ECSM problem admits a $1 + O(1)/k$ approximation algorithm.*

In other words, if true, the above conjecture implies that the $3/2$ -classical factor is not optimal for sufficiently large k , and moreover that it is possible to design an approximation algorithm whose factor gets arbitrarily close to 1 as $k \rightarrow \infty$. In this paper, we prove a weaker version of the above conjecture.

Theorem 1.2 (Main). *There is a randomized algorithm for (weighted) k -ECSM with approximation factor (at most) $1 + \sqrt{\frac{8 \ln k}{k}}$.*

We remark that our main theorem only improves the classical $3/2$ -approximation algorithm for k -ECSM only when $k \geq 164$ (although one can use the more precise expression given in the proof to, for example, improve upon $3/2$ for even values of $k \geq 66$).

For a set $S \subseteq V$, let $\delta(S) = \{\{u, v\} : |\{u, v\} \cap S| = 1\}$ denote the set of edges leaving S . The following is the natural linear programming relaxation for k -ECSM.

$$\begin{aligned}
 \min \quad & \sum_{e \in E} x_e c(e) \\
 \text{s.t.} \quad & x(\delta(v)) = k \quad \forall v \in V \\
 & x(\delta(S)) \geq k \quad \forall S \subseteq V \\
 & x_e \geq 0 \quad \forall e \in E.
 \end{aligned} \tag{1}$$

Note that while in an optimum solution of k -ECSM the degree of each vertex is not necessarily equal to k , since the cost function satisfies the triangle inequality we may assume that in any optimum fractional solution each vertex has (fractional) degree k . This follows from the parsimonious property [GB93].

We prove [Theorem 1.2](#) by rounding an optimum solution to the above linear program. So, as a corollary we also upper-bound the integrality gap of the above linear program.

Corollary 1.3. *The integrality gap of LP (1) is at most $1 + \sqrt{\frac{8 \ln k}{k}}$.*

1.1 Proof Overview

Before explaining our algorithm, we recall a randomized rounding approach of Karger [Kar99]. Karger showed that if we choose every edge e independently with probability x_e , then the sample is $k - O(\sqrt{k \log n})$ -edge connected with high probability. He then fixes the connectivity of the sample by adding $O(\sqrt{k \log n})$ copies of the minimum spanning tree of G . This gives a $1 + O(\sqrt{\log n/k})$ approximation algorithm for the problem.

First, we observe that where x is a solution to the LP (1), the vector $2x/k$ is in the spanning tree polytope (after modifying x slightly, see [Fact 2.1](#) for more details). Following a recent line of works on the traveling salesperson problem [OSS11; KKO20] we write $2x/k$ as a λ -uniform spanning tree distribution, μ_λ . Then, we independently sample $k/2$ spanning trees¹ $T_1, \dots, T_{k/2}$. It follows that $T^* = T_1 \uplus T_2 \uplus \dots \uplus T_{k/2}$ has the same expectation across every cut as x , and due to properties of λ -uniform spanning tree distributions it is concentrated around its mean. Unlike the independent rounding procedure, T^* has at least $k/2$ edges across each cut with probability 1. This implies that the number of “bad” cuts of T^* , i.e. those of size strictly less than k , is at most $(n-1)k/2$ (with probability 1). This is because any tree T_i has strictly less than 2 edges in exactly $n-1$ “tree cuts,” and a cut lying on no tree cuts must have at least k edges in T^* .

We divide these potentially $O(nk)$ bad cuts into two types: (i) Cuts S such that $\delta(S)_{T^*} \geq k - \alpha\sqrt{k/2 - 1}$ and (ii) Cuts S where $\delta(S)_{T^*} < k - \alpha\sqrt{k/2 - 1}$, for some $\alpha = \Theta(\sqrt{\ln k})$. We fix all cuts of type (i) by adding $\alpha\sqrt{k/2 - 1}$ copies of the minimum spanning tree of G . To fix cuts S of type (ii), we employ the following procedure: for any tree T_i where $\delta(S)_{T_i} = 1$ and S is of type (ii), we add one extra copy of the unique edge of T_i in $\delta(S)$. To bound the expected cost of our rounded solution, we use the concentration property of λ -uniform trees on edges of T^* to show that the probability any fixed cut $\delta(S)$ is of type (ii) is exponentially small in α , $\leq e^{-\alpha^2/2}$, even if we condition on $\delta(S)_{T_i} = 1$ for a single tree T_i .

1.2 Algorithm

For two sets of edges $F, F' \subseteq E$, we write $F \uplus F'$ to denote the multi-set union of F and F' allowing multiple edges. Note that we always have $|F \uplus F'| = |F| + |F'|$.

Let x be an optimal solution of LP (1). We expand the graph $G = (V, E)$ to a graph G^0 by picking an arbitrary vertex $u \in V$, splitting it into two nodes u_0 and v_0 , and then, for every edge $e = (u, w)$ incident to u , assigning fraction $\frac{x(e)}{2}$ to each of the two edges (u_0, w) and (v_0, w) in G^0 . Call this expanded graph G^0 , its edge set E^0 , and the resulting fractional solution x^0 , where $x^0(e)$

¹If k is odd, we sample $\lceil k/2 \rceil$ trees. The bound remains unchanged relative to the analysis we give below as the potential cost of one extra tree is $O(OPT/k)$.

and $x(e)$ are identical on all other edges. (Note that each of u_0 and v_0 now have fractional degree $k/2$ in x^0 .) In [Fact 2.1](#) below, we show that $\frac{2}{k} \cdot x^0$ is in the spanning tree polytope for the graph G^0 . For ease of exposition, the algorithm is described as running on G^0 (and spanning trees² of G^0), which has the same edge set as G (when u_0 and v_0 are identified).

Our algorithm is as follows:

Algorithm 1 An Approximation Algorithm for k -ECSM

- 1: Let x^0 be an optimum solution of (1) extended to the graph G^0 as described above.
 - 2: Find weights $\lambda : E^0 \rightarrow \mathbb{R}_{\geq 0}$ such that for any $e \in E^0$, $\mathbb{P}_{\mu_\lambda}[e] \leq \frac{2}{k} x_e^0 (1 + 2^{-n})$. ▷ By [Theorem 2.2](#)
 - 3: Sample $k/2$ spanning trees $T_1, \dots, T_{k/2} \sim \mu_\lambda$ (in G^0) independently and let $T^* \leftarrow T_1 \uplus \dots \uplus T_{k/2}$.
 - 4: Let B be $\alpha\sqrt{k/2 - 1}$ copies of the MST of G^0 . ▷ $\alpha = \Theta(\sqrt{\ln k})$ is a parameter we choose later.
 - 5: **for** $i \in [\frac{k}{2}]$ and $e \in T_i$ **do**
 - 6: **if** $C_{T_i}(e)_{T^*} < k - \alpha\sqrt{k/2 - 1}$ and $(u_0, v_0) \notin C_{T_i}(e)$ **then**
 - 7: $F \leftarrow F \uplus \{e\}$.
 - 8: **end if**
 - 9: **end for**
 - 10: **Return** $T^* \uplus B \uplus F$.
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2 Preliminaries

For any set of edges $F \subseteq E$ and a set of edges $T \subseteq E$, we write

$$F_T := |F \cap T|.$$

Also, for any edge weight function $x : E \rightarrow \mathbb{R}$, we write $x(F) := \sum_{e \in F} x(e)$.

For any spanning tree T of G^0 , and any edge $e \in T$, we write $C_T(e) \subseteq E$ to denote the set of edges in the unique cut obtained by deleting e from T . Of particular interest to us below will be $C_{T_i}(e)_{T^*} = |C_{T_i}(e) \cap T^*|$ where e is an edge in T_i .

We will also use the notation $(u_0, v_0) \notin C$ to indicate that u_0 and v_0 are on the same side of the cut C .

2.1 Random Spanning Trees

Edmonds [[Edm70](#)] gave the following description for the convex hull of the spanning trees of any graph $G = (V, E)$, known as the *spanning tree polytope*.

$$\begin{aligned} z(E) &= |V| - 1 \\ z(E(S)) &\leq |S| - 1 && \forall S \subseteq V \\ z_e &\geq 0 && \forall e \in E. \end{aligned} \tag{2}$$

Edmonds also [[Edm70](#)] proved that the extreme point solutions of this polytope are the characteristic vectors of the spanning trees of G .

²A spanning tree in G^0 is a 1-tree in G , that is, a tree plus an edge.

Fact 2.1 ([KKO20]). Let x be the optimal solution of LP (1) and x^0 its extension to G^0 as described above. Then $\frac{2}{k} \cdot x^0$ is in the spanning tree polytope (2) of G^0 .

Proof. For any set $S \subseteq V(G^0)$ with $u_0, v_0 \notin S$, $x^0(E(S)) = \frac{k|S| - x(\delta(S))}{2} \leq \frac{k}{2}(|S| - 1)$. If $u_0 \in S, v_0 \notin S$, then $x^0(\delta(S)) \geq k/2$, so $x^0(E(S)) \leq \frac{k|S| - k/2 - x^0(\delta(S))}{2} \leq \frac{k}{2}(|S| - 1)$. Finally, if $u_0, v_0 \in S$, then $x^0(\delta(S)) \geq k$. Thus, $x^0(E(S)) = \frac{k|S| - k - x^0(\delta(S))}{2} \leq \frac{k}{2}(|S| - 2)$. The claim follows because $x^0(E) = \frac{k|V(G^0)|}{2} = \frac{k}{2}(|V(G^0)| - 1)$. \square

Given nonnegative edge weights $\lambda : E \rightarrow \mathbb{R}_{\geq 0}$, we say a distribution μ_λ over spanning trees of G is λ -uniform, if for any spanning tree T ,

$$\mathbb{P}_{T \sim \mu_\lambda} [T] \propto \prod_{e \in T} \lambda(e).$$

Theorem 2.2 ([Asa+17]). There is a polynomial-time algorithm that, given a connected graph $G = (V, E)$, and a point $z \in \mathbb{R}^{|E|}$ in the spanning tree polytope (2) of $G = (V, E)$, returns $\lambda : E \rightarrow \mathbb{R}_{\geq 0}$ such that the corresponding λ -uniform spanning tree distribution μ_λ satisfies

$$\sum_{T \in \mathcal{T}: e \in T} \mu_\lambda(T) \leq (1 + 2^{-n})z_e, \quad \forall e \in E,$$

i.e., the marginals are approximately preserved. In the above \mathcal{T} is the set of all spanning trees of G .

2.2 Bernoulli-Sum Random Variables

Definition 2.3 (Bernoulli-Sum Random Variable). We say $BS(q)$ is a Bernoulli-Sum random variable if it has the law of a sum of independent Bernoullis, say $B_1 + B_2 + \dots + B_t$ for some $t \geq 1$, with $\mathbb{E}[B_1 + \dots + B_t] = q$.

Fact 2.4. If $X = BS(q_1)$ and $Y = BS(q_2)$ are two independent Bernoulli-sum random variables then $X + Y = BS(q_1 + q_2)$.

Lemma 2.5 ([BBL09; Pit97]). Given $G = (V, E)$ and $\lambda \in E \rightarrow \mathbb{R}_{\geq 0}$, let μ_λ be the λ -uniform spanning tree distribution of G . Let T be a sample from μ_λ . Then for any fixed $F \subseteq E$, the random variable F_T is distributed as $BS(\mathbb{E}[F_T])$.

Theorem 2.6 (Multiplicative Chernoff-Hoeffding Bound for BS Random Variables). Let $X = BS(q)$ be a Bernoulli-Sum random variable. Then, for any $0 < \epsilon < 1$ and $q' \leq q$

$$\mathbb{P}[X < (1 - \epsilon)q'] \leq e^{-\frac{\epsilon^2 q'}{2}}.$$

3 Analysis of the Algorithm

In this section we prove [Theorem 1.2](#). We first observe that the cuts of G are precisely the cuts of G^0 that have u_0 and v_0 on the same side of the cut, and for any such cut the set of edges crossing the cut in G and in G^0 is the same (once u_0 and v_0 are contracted). We begin by showing that the output of [Algorithm 1](#) is k -edge connected (in G) with probability 1.

Lemma 3.1 (*k*-Connectivity of the Output). For any $\alpha \geq 0$, the output of *Algorithm 1*, $F \uplus B \uplus T^*$ is a *k*-edge connected subgraph of *G*.

Proof. Fix spanning trees $T_1, \dots, T_{k/2}$ in G^0 and let $C = \delta(S)$ for some S , where $(u_0, v_0) \notin C$. We show that $C_{T^* \uplus F \uplus B} \geq k$. If $C_{T^*} \geq k - \alpha\sqrt{k/2 - 1}$, then since B has $\alpha\sqrt{k/2 - 1}$ copies of the minimum spanning tree, $C_{T^* \uplus B} \geq k$ and we are done. Otherwise $C_{T^*} < k - \alpha\sqrt{k/2 - 1}$. Then, we know that for any tree T_i , either $C_{T_i} \geq 2$ or $C_{T_i} = 1$. If $C_{T_i} = 1$, since $(u_0, v_0) \notin C_{T_i}$, F has one extra copy of the unique edge of T_i in C . Therefore, including those cases where an extra copy of the edge e is added, each T_i has at least two edges in C , so $C_{T^* \uplus F} \geq k$ as desired. \square

Lemma 3.2. For any $0 \leq \alpha \leq \sqrt{k/2 - 1}$, $1 \leq i \leq k/2$, and any $e \in E$,

$$\mathbb{P} \left[C_{T_i}(e)_{T^*} \leq k - \alpha\sqrt{k/2 - 1} \mid e \in T_i \wedge (u_0, v_0) \notin C_{T_i}(e) \right] \leq e^{-\alpha^2/2}.$$

where the randomness is over spanning trees $T_1, \dots, T_{i-1}, T_{i+1}, \dots, T_{k/2}$ independently sampled from μ_λ .

Proof. Condition on tree T_i such that $e \in T_i$ and $(u_0, v_0) \notin C_{T_i}(e)$.

By *Lemma 2.5*, for any $j \leq k/2$ such that $j \neq i$, $C_{T_i}(e)_{T_j}$ is a $BS(\mathbb{E} [C_{T_i}(e)_{T_j}])$ random variable, with $\mathbb{E} [C_{T_i}(e)_{T_j}] = \frac{2}{k}x(C_{T_i}(e)) \geq 2$. Also, by definition, $C_{T_i}(e)_{T_i} = 1$ (with probability 1). Since $T_1, \dots, T_{k/2}$ are independently chosen, by *Fact 2.4* the random variable $C_{T_i}(e)_{T^*}$ is distributed as $BS(q)$ for $q \geq k - 1$. Since each T_j has at least one edge in $C_{T_i}(e)$, $C_{T_i}(e)_{T^*} \geq k/2$ with probability 1. So, by *Theorem 2.6*, with $q' = k - 1 - k/2$, when $0 \leq \alpha \leq \sqrt{k/2 - 1}$,

$$\begin{aligned} & \mathbb{P} \left[C_{T_i}(e)_{T^*} < k - \alpha\sqrt{k/2 - 1} \mid e \in T_i \wedge (u_0, v_0) \notin C_{T_i}(e) \right] \\ &= \mathbb{P} \left[C_{T_i}(e)_{T^*} - k/2 < k/2 - \alpha\sqrt{k/2 - 1} \mid e \in T_i \wedge (u_0, v_0) \notin C_{T_i}(e) \right] \leq e^{-\frac{(\alpha\sqrt{k/2-1})^2(k/2-1)}{2}} = e^{-\alpha^2/2}. \end{aligned}$$

Averaging over all realizations of T_i satisfying the required conditions proves the lemma. \square

Proof of Theorem 1.2. Let x be an optimum solution of LP (1). Since the output of the algorithm is always *k*-edge connected we just need to show $\mathbb{E} [c(F \cup T^* \cup B)] \leq \left(1 + \sqrt{\frac{8 \ln k}{k}}\right) c(x)$. By linearity of expectation,

$$\mathbb{E} [c(T^*)] = \sum_{i \in [\frac{k}{2}]} \mathbb{E} [c(T_i)] = \frac{k}{2} \sum_{e \in E} c(e) \mathbb{P}_{\mu_\lambda} [e] = \frac{k}{2} \sum_{e \in E} c(e) \cdot \frac{2}{k} \cdot x_e = c(x),$$

where for simplicity we ignored the $1 + 2^{-n}$ loss in the marginals. On the other hand, since by *Fact 2.1*, $\frac{2x}{k}$ is in the spanning tree polytope of G^0 , $c(B) \leq \frac{2c(x)}{k} \cdot \alpha\sqrt{k/2 - 1} \leq \frac{\alpha c(x)}{\sqrt{k/2}}$. It remains to bound the expected cost of F . By *Lemma 3.2*, we have,

$$\begin{aligned} \mathbb{E} [c(F)] &= \sum_{e \in E} c(e) \sum_{i=1}^{k/2} \mathbb{P} [e \in T_i \wedge (u_0, v_0) \notin C_{T_i}(e)] \mathbb{P} \left[C_{T_i}(e)_{T^*} < k - \alpha\sqrt{k/2 - 1} \mid e \in T_i \wedge (u_0, v_0) \notin C_{T_i}(e) \right] \\ &\leq \sum_{e \in E} c(e) x_e e^{-\alpha^2/2} \leq e^{-\alpha^2/2} c(x). \end{aligned}$$

Putting these together we get, $\mathbb{E} [c(T^* \cup B \cup F)] \leq (1 + \alpha/\sqrt{k/2} + e^{-\alpha^2/2})c(x)$. Setting $\alpha = \sqrt{\ln \left(\frac{k}{2}\right)}$ finishes the proof. \square

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