# A Distributed Minimum Cut Approximation Scheme 

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#### Abstract

In this paper, we study the problem of approximating the minimum cut in a distributed message-passing model, the CONGEST model. The minimum cut problem has been well-studied in the context of centralized algorithms. However, there were no known non-trivial algorithms in the distributed model until the recent work of Ghaffari and Kuhn. They gave algorithms for finding cuts of size $O\left(\epsilon^{-1} \lambda\right)$ and $(2+\epsilon) \lambda$ in $O(D)+\tilde{O}\left(n^{1 / 2+\epsilon}\right)$ rounds and $\tilde{O}(D+\sqrt{n})$ rounds respectively, where $\lambda$ is the size of the minimum cut. This matches the lower bound they provided up to a polylogarithmic factor. Yet, no scheme that achieves $(1+\epsilon)$-approximation ratio is known. We give a distributed algorithm that finds a cut of size $(1+\epsilon) \lambda$ in $\tilde{O}(D+\sqrt{n})$ time, which is optimal up to polylogarithmic factors.


## 1 Introduction

The minimum cut problem is a fundamental problem in graph algorithms and network design. Given a weighted undirected graph $G=(V, E)$, a cut $C=(S, V \backslash S)$ where $\emptyset \subset S \subset V$, is a partition of vertices into two non-empty sets. The weight of a cut, $w(C)$, is defined to be the sum of the edge weights crossing $C$. The minimum cut problem is to find a cut with the minimum weight. The exact version of the problem as well as the approximate version have been studied for many years [6, 10, 8, 13, 12, 3, 15, 9] in the context of centralized models of computation, resulting in nearly linear time algorithms [9, 12, 8].

Elkin [2] and Das Sarma et al. 1] addressed the problem in the distributed message-passing model. The problem has trivial time complexity of $\Theta(D)$ (unweighted diameter) in the LOCAL model, where the message size is unlimited. Ghaffari and Kuhn [5] recently developed approximation algorithms for this problem in the CONGEST model where each message is bounded by $\Theta(\log n)$ bits. They assume that the edges of $G$ have integer weights from $\left\{1, \ldots, n^{\Theta(1)}\right\}$ and treat $G$ as an unweighted multigraph, where an edge $e$ with weight $w(e)$ is converted to $w(e)$ parellel edges, while still only $\Theta(\log n)$ bits can be sent over these parallel edges together in each round. Let $\lambda$ be the value of the minimum cut, they give an algorithm that finds a cut of size at most $O\left(\epsilon^{-1} \lambda\right)$ in $O(D)+O\left(n^{1 / 2+\epsilon} \log ^{3} n \log \log n \log ^{*} n\right)$ time. Moreover, they gave an algorithm that finds a cut of size at most $(2+\epsilon) \lambda$ in $O\left(\left(D+\sqrt{n} \log ^{*} n\right) \log ^{2} n \log \log n \frac{1}{\epsilon^{5}}\right)$ time. Das Sarma et al. [1] showed $\alpha$-approximating the minimum cut requires $\tilde{\Omega}(D+\sqrt{n})$ rounds for weighted graphs for any $\alpha \geq 1$. Ghaffari and Kuhn extended their lower bound for unweighted multigraphs (which is equivalent to the setting where one is allowed to send messages of size $w \cdot \Theta(\log n)$ over an edge of weight $w$ in weighted graphs). For unweighted simple graphs, they also gave a lower bound of $\tilde{\Omega}(D+\sqrt{n / \alpha})$. Therefore, the upper bound and lower bound provided by Ghaffari and Kuhn match up to a polylogarithmic factor.

[^0]However, still no approximation algorithms exist for any approximation factor less than 2 . In this paper, we give a simple algorithm that finds a minimum cut of size at most $(1+\epsilon) \lambda$ in $\tilde{O}(D+\sqrt{n})$ time. In particular, our algorithm runs in $O\left(\left(\log ^{11} n / \epsilon^{17}\right)\left(D+\sqrt{n} \log ^{*} n\right)\right)$ rounds.

Our approach uses the semi-duality between minimum cuts and tree packings as in (9, 16]. Karger [9] showed that if we greedily pack enough trees, then for any minimum cut, there is a tree crossing the cut at most twice. However, it is technically not easy to utilize this fact to find minimum cuts in the distributed model. Instead, we use a lemma by Thorup [16], which shows that if we pack more trees then there is at least one minimum cut that is crossed by a tree exactly once. We take some ingredients from Ghaffari and Kuhn's algorithm and Thurimella's algorithm [17] for identifying biconnected components to devise a procedure that is able to simultanously test the values of the $n-1$ cuts induced by deleting one of the $n-1$ edges in a tree. Note that the number of trees we have to pack is polynomial in the value of the minimum cut. Thus, we will first use the sampling lemma of Karger [7] to obtain a sampled graph that scales the value of the minimum cut down to $O\left(\log n / \epsilon^{2}\right)$. Then we only have to pack polylogarithmic number of trees. Finally, we combine the resampling procedure, the tree packing, and the procedure for testing tree-induced-cuts to find an approximate minimum cut.

## 2 Distributed Minimum Cut Approximation

Let $G$ be a connected graph with integer weights from $\{1, \ldots, W\}$, where $W=n^{\Theta(1)}$. We will treat $G$ as a multigraph with uniform edge weights. Let $\lambda$ be the weight of the minimum cut of $G$. We show how to find such an approximate minimum cut whose weight is at most $(1+\epsilon) \lambda$.

An edge $e$ is a bridge if it does not exist a cycle in $G$ passing $e$ (or equivalently, deleting $e$ breaks $G$ into two connected components). Given two graph $A$ and $B$ with the same vertex set, $A+B$ is the multigraph obtained by including edges in $A$ and edges in $B$.

A tree packing $\mathcal{T}$ is a multiset of spanning trees. The load of an edge $e$ with respect to $\mathcal{T}$ is the number of trees in $\mathcal{T}$ containing $e$. Given a tree $T$, we say a cut is induced by $T$ if such a cut is obtained by deleting an edge $e \in T$. We will denote this cut by $C(T, e)$. A tree packing $\mathcal{T}=\left\{T_{1}, \ldots, T_{k}\right\}$ is greedy if each $T_{i}$ is a minimum spanning tree with respect to the loads induced by $\left\{T_{1}, \ldots, T_{i-1}\right\}$. Let $\epsilon^{\prime}=\Theta(\epsilon)$ such that $\left(1+\epsilon^{\prime}\right)^{3} /\left(1-\epsilon^{\prime}\right)=1+\epsilon$.
Lemma 2.1 (Thorup [16]). A greedy tree packing with $96(\lambda+1)^{7} \log ^{3} m$ trees contains a tree crossing some min-cut only once.

Remark 2.2. The number of trees in the original statement of the lemma is $\omega\left(\lambda^{7} \log ^{3} m\right)$, though the proof actually implies that $\Theta\left(\lambda^{7} \log ^{3} m\right)$ is enough. In particular, Thorup showed $24 \lambda \ln m / \epsilon^{2}$ trees is sufficient, where $\epsilon$ satisfies $\frac{\epsilon\left(3+\log _{1+\alpha} m\right)}{\lambda}+\alpha<\frac{2}{\lambda(\lambda+1)}$ for some $\alpha<1$. We can choose $\alpha=\frac{1}{\lambda(\lambda+1)}$ and $\epsilon=\frac{1}{2(\lambda+1)^{3} \ln m}$ to make the inequality hold. Therefore, $96(\lambda+1)^{7} \ln ^{3} m$ trees is sufficient.

We describe our algorithm in Algorithm [1. The subroutine $\operatorname{Test}(T, \kappa)$ returns a cut whose weight is at most $\left(1+\epsilon^{\prime}\right) \kappa$ w.h.p. if there exists a cut in $G$ induced by $T$ with weight at most $\kappa$.

We show that w.h.p. the algorithm will output a cut $C$ with $w(C) \leq(1+\epsilon) \lambda$. In particular, consider the iteration $i$ where $\lambda \in\left[X_{i}, X_{i+1}\right]$. Let $\lambda^{\prime}$ denote the value of the minimum cut in the sampled graph $H_{i}$. If $i=0$, then it is clear that $\lambda^{\prime}=\lambda \leq X_{1}=20 \ln n / \epsilon^{\prime 2}$. If $i>0$, since we sampled with probability $1 / 2^{i}=20 \ln n /\left(\epsilon^{\prime 2} X_{i+1}\right)=10 \ln n /\left(\epsilon^{\prime 2} X_{i}\right) \geq 10 \ln n /\left(\epsilon^{\prime 2} \lambda\right)$, we know that w.h.p. for any cut $C$ [7, Corollary 2.4],

$$
\left(1-\epsilon^{\prime}\right) \cdot w_{G}(C) / 2^{i} \leq w_{H_{i}}(C) \leq\left(1+\epsilon^{\prime}\right) \cdot w_{G}(C) / 2^{i}
$$

Therefore, $\lambda^{\prime} \leq\left(1+\epsilon^{\prime}\right) \lambda / 2^{i} \leq\left(1+\epsilon^{\prime}\right) 20 \ln n / \epsilon^{\prime 2}$. If we pack $96\left(\lambda^{\prime}+1\right)^{7} \log ^{3} m$ trees in $\mathcal{T}$, then by Lemma 2.1 there exists a tree crossing some minimum cut $C^{*}$ of $H_{i}$ only once. Notice that for any other cut $C^{\prime}$,

$$
w_{G}\left(C^{*}\right) \leq 2^{i} \cdot \frac{w_{H_{i}}\left(C^{*}\right)}{1-\epsilon^{\prime}}=2^{i} \cdot \frac{\lambda^{\prime}}{1-\epsilon^{\prime}} \leq 2^{i} \cdot \frac{w_{H_{i}}\left(C^{\prime}\right)}{1-\epsilon^{\prime}} \leq \frac{1+\epsilon^{\prime}}{1-\epsilon^{\prime}} \cdot w_{G}\left(C^{\prime}\right)
$$

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\(X_{0} \leftarrow 1\)
\(i \leftarrow 0\)
repeat
    \(X_{i+1} \leftarrow 2^{i} \cdot 20 \ln n / \epsilon^{\prime 2}\)
    (We are assuming \(\lambda \in\left[X_{i}, X_{i+1}\right]\) in this iteration)
    Let \(H_{i}\) be the subgraph sampled with probability \(p=1 / 2^{i}\) on each edge of \(G\).
    Find a greedy tree packing \(\mathcal{T}\) with \(96\left(\left(1+\epsilon^{\prime}\right) 20 \ln n / \epsilon^{\prime 2}+1\right)^{7} \ln ^{3} m\) trees in \(H_{i}\)
    \(\gamma \leftarrow X_{i}\)
    repeat
        for each \(T \in \mathcal{T}\) do
            Call Test \(\left(T,\left(1+\epsilon^{\prime}\right) \gamma\right)\).
            If \(\operatorname{Test}\left(T,\left(1+\epsilon^{\prime}\right) \gamma\right)\) returns a cut \(C\), output \(C\) and terminate.
        end for
        \(\gamma \leftarrow\left(1+\epsilon^{\prime}\right) \gamma\)
    until \(\gamma>\frac{1+\epsilon^{\prime}}{1-\epsilon^{\prime}} \cdot X_{i+1}\)
    \(i \leftarrow i+1\)
until \(X_{i+1}>n W\)
```

Algorithm 1: $(1+\epsilon)$-approximate minimum cut
Therefore, one of the cuts induced by some $T \in \mathcal{T}$ is an $\left(1+\epsilon^{\prime}\right) /\left(1-\epsilon^{\prime}\right)$ approximate minimum cut. Denote this cut by $C^{\prime}$, so $w\left(C^{\prime}\right) \in\left[X_{i},\left(\left(1+\epsilon^{\prime}\right) /\left(1-\epsilon^{\prime}\right)\right) \cdot X_{i+1}\right]$. Therefore in the $i^{\prime}$ th iteration, there exists $\gamma$ in the loop (Line9-Line [15) such that $w\left(C^{\prime}\right) \in\left[\gamma,\left(1+\epsilon^{\prime}\right) \gamma\right]$. So w.h.p. we will output a cut with weight at most $\left(1+\epsilon^{\prime}\right)^{2} \gamma \leq\left(1+\epsilon^{\prime}\right)^{2} w\left(C^{\prime}\right) \leq\left(1+\epsilon^{\prime}\right)^{3} /\left(1-\epsilon^{\prime}\right) w\left(C^{*}\right)=(1+\epsilon) w\left(C^{*}\right)$.

### 2.1 Distributed Implmentation

We have shown the correctness of this algorithm. It remains to show how to implement it in $\tilde{O}(D+\sqrt{n})$ distributed rounds, and in particular, to implement the tree packing (Line 7) and Test $(T, \kappa)$ in Algorithm [1. To pack $k$ trees, it is striaghtfoward to apply $k$ MST computations on the graph where the edge weights are equal to the number of trees including it. This can be done in $O\left(k\left(D+\sqrt{n} \log ^{*} n\right)\right)$ rounds [11].

Given a partition $\mathcal{P}$ of $G$ into components, Ghaffari and Kuhn [5] devised a testing procedure to test if there is a cut induced by a component in $\mathcal{P}$ that has weight less than $\kappa$ in $\widetilde{O}(D+\sqrt{n})$ rounds. Given a spanning tree $T$, we will show how to test the $n-1$ cuts induced by $T$ also in $\widetilde{O}(D+\sqrt{n})$ rounds.

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for \(i \leftarrow 1 \ldots k=\Theta\left(\frac{\log n}{\epsilon^{2}}\right)\) do
    Let \(G_{i}\) be the subgraph obtained by sampling each edge of \(G\) independently with probability
    \(1-2^{-1 / \kappa}\).
3: For each edge \(e \in T\), determine if \(e\) is a bridge in the graph \(G_{i}+T\).
4: Let \(Y_{e, i}= \begin{cases}1 & \text { if } e \text { is not a bridge in the graph } G_{i}+T \text {. } \\ 0 & \text { otherwise. }\end{cases}\)
end for
If there is \(e \in T\) such that \(\sum_{i=1}^{k} Y_{e, i} \leq k / 2+\epsilon^{\prime} k / 8\), then return the cut \(C(T, e)\)
```

Algorithm 2: $\operatorname{Test}(T, \kappa)$. $\operatorname{Test}(T, \kappa)$ returns a cut whose weight is at most $\left(1+\epsilon^{\prime}\right) \kappa$ w.h.p. if there exists a cut in $G$ induced by $T$ with weight at most $\kappa$. Note that the sample probability $1-2^{-1 / \kappa}=\Theta(1 / \kappa)$.

Lemma 2.3. If $T$ induces a cut $C(T, e)$ with weight at most $\kappa$, then $T$ est $(T, \kappa)$ returns a cut w.h.p. Moreover, any cut returned by the algorithm has weight at most $\left(1+\epsilon^{\prime}\right) \kappa$ w.h.p.

Proof. Consider a cut $C(T, e)$. First observe that $G_{i}$ contains an edge crossing $C(T, e)$ if and only if $e$ is not a bridge in the graph $G_{i}+T$. Therefore, $\mathrm{E}\left[Y_{e, i}\right]=1-\left(1-\left(1-2^{-1 / \kappa}\right)\right)^{w(C(T, e))}=$ $1-2^{-w(C(T, e)) / \kappa}$.

If there is $C(T, e) \leq \kappa$, then $\mathrm{E}\left[Y_{e, i}\right] \leq 1 / 2$ and $\mathrm{E}\left[\sum_{i} Y_{e, i}\right] \leq k / 2$. By Hoeffiding's inequality, $\operatorname{Pr}\left(\sum_{i} Y_{e, i}>k / 2+\epsilon^{\prime} k / 8\right) \leq \operatorname{Pr}\left(\sum_{i} Y_{e, i}>\mathrm{E}\left[\sum_{i} Y_{e, i}\right]+\epsilon^{\prime} k / 8\right) \leq e^{-\frac{2\left(\epsilon^{\prime} k / 8\right)^{2}}{k}}=e^{-\epsilon^{\prime 2} k / 32}=1 / \operatorname{poly}(n)$. By taking the union bound over the $n-1$ cuts induced by $T$, we conclude that w.h.p. the algorithm will return a cut if there is cut whose weight is at most $\kappa$.

On the other hand if $w(C(T, e))>\left(1+\epsilon^{\prime}\right) \kappa$, then $\mathrm{E}\left[Y_{e, i}\right]=1-2^{-1-\epsilon^{\prime}} \geq 1 / 2+\epsilon^{\prime} / 4$ when $\epsilon^{\prime} \leq 1$, since $2^{-\epsilon^{\prime}} \leq 1-\epsilon^{\prime} / 2$ when $\epsilon^{\prime} \leq 1$. So $E\left[\sum_{i} Y_{e, i}\right] \geq k / 2+\epsilon^{\prime} k / 4$. By Hoeffiding's inequality, $\operatorname{Pr}\left(\sum_{i} Y_{e, i} \leq k / 2+\epsilon^{\prime} k / 8\right) \leq \operatorname{Pr}\left(\sum_{i} Y_{e, i} \leq \mathrm{E}\left[\sum_{i} Y_{e, i}\right]-\epsilon^{\prime} k / 8\right) \leq e^{-\frac{2\left(\epsilon^{\prime} k / 8\right)^{2}}{k}}=e^{-\epsilon^{\prime 2} k / 32}=1 / \operatorname{poly}(n)$. By taking the union bound over the $n-1$ cuts induced by $T$, we conclude the cut returned by the algorithm has weight at most $\left(1+\epsilon^{\prime}\right) \kappa$ w.h.p.

### 2.2 Computing the Bridges

Given a subgraph $G_{i}$ of $G$, it remains to show how to determine what edges of $T$ are bridges in the subgraph $T+G_{i}$ in $\tilde{O}(D+\sqrt{n})$ rounds. Thurimella 17] gave an algorithm for computing the biconnected components of the underlying graph in $\tilde{O}(D+\sqrt{n})$ rounds. With simple modifications, it can be applied to compute which edges of $T$ are bridges in the subgraph $G_{i}$ of the underlying graph $G$. Note that even we have the algorithm for computing the bridges of $T$ in $G+T$, it is not clearly whether we can directly simulate it to compute the bridges of $T$ in $G_{i}+T$, because we want the running time to depend on the diameter of $G$ rather than that of $G_{i}$. Therefore, we describe the algorithm and necessary changes below.

Fix a root $r$ in $T$. Let $\operatorname{pre}(u) \in[0, n-1]$ be the preorder number which denote the time $u$ is visited if we perform a depth-first search on $T$ starting at $r$. Denote the subtree rooted at $u$ by $T_{u}$
and let size ( $u$ ) be the size of $T_{u}$. Let

$$
\begin{gathered}
\operatorname{low}(u) \stackrel{\text { def }}{=} \min \begin{cases}\operatorname{pre}(u) \\
\operatorname{low}(v) & v \text { is a child of } u \text { in } T \\
\operatorname{pre}(v) & u v \in G_{i} \stackrel{\dagger}{ }\end{cases} \\
\operatorname{high}(u) \stackrel{\text { def }}{=} \max \begin{cases}\operatorname{pre}(u) \\
\operatorname{high}(v) & v \text { is a child of } u \text { in } T \\
\operatorname{pre}(v) & u v \in G_{i} \stackrel{\dagger}{\leftrightharpoons}\end{cases}
\end{gathered}
$$

Lemma 2.4. Let $u v \in T$ with $v$ being a child of $u$, i.e. pre $(u)<\operatorname{pre}(v)$. uv is a bridge if and only if $\operatorname{low}(v) \geq \operatorname{pre}(v)$ and $\operatorname{high}(v) \leq \operatorname{pre}(v)+\operatorname{size}(v)-1$.

Proof. First notice that every vertex $x \in T_{v}$ must have $\operatorname{pre}(x) \in[\operatorname{pre}(v), \operatorname{pre}(v)+\operatorname{size}(v)-1]$. If $u v$ is a bridge, then no descendent of $v$ will be adjacent to anything outside the subtree rooted at $v$, for otherwise a cycle passing $u v$ will be created. Therefore, $\operatorname{low}(v), \operatorname{high}(v) \in[\operatorname{pre}(v), \operatorname{pre}(v)+$ $\operatorname{size}(v)-1]$.

On the other hand, if $\operatorname{low}(v)<\operatorname{pre}(v)$ or if $\operatorname{high}(v) \geq \operatorname{pre}(v)+\operatorname{size}(v)$, then there exists a vertex $y \in T_{v}$ and $z \notin T_{v}$ such that $y$ and $z$ are adjacent. Since $z \notin T_{v}$, there must exists a path from $z$ to $u$ such that it does not pass $u v$. Therefore, $u \rightarrow v \rightsquigarrow y \rightarrow z \rightsquigarrow u$ forms a cycle and $u v$ is not a bridge.

Remark 2.5. Note that the second condition $\operatorname{high}(v) \leq \operatorname{pre}(v)+\operatorname{size}(v)-1$ is needed because $T$ is not necessarily a DFS tree.

Now it remains to show how to compute $\operatorname{pre}(u)$, $\operatorname{low}(u)$, and $\operatorname{high}(u)$ in $\tilde{O}(D+\sqrt{n})$ time. It is explicitly described in 17 how to compute $\operatorname{pre}(u)$. Note that $\operatorname{pre}(u)$ is independent of the $G_{i}$. Although $\operatorname{low}(u)$ and $\operatorname{high}(u)$ depend on $G_{i}$, they can be computed in a similar way in $\tilde{O}(D+\sqrt{n})$ time. For completeness, we describe how to compute these functions in the following.

Lemma 2.6 ( $4, ~[11)$. A tree of $n$ vertices can be divided into $O(\sqrt{n})$ connected subgraphs each of diameter $O(\sqrt{n})$ in $O\left(\sqrt{n} \log ^{*} n\right)$ time

First, use Lemma 2.6 to decompose the rooted tree $T$ into components $F_{1}, \ldots F_{O(\sqrt{n})}$. For each component $F_{i}$, there is a root $r_{i}$ which is either the root of $T, r$, or the unique vertex in $F_{i}$ connecting to its parent outside $F_{i}$. It is shown in 14 that the root $r$ is able to downcast distinct messages of size $O(\log n)$ to each of $r_{i}$ in $O(D+\sqrt{n})$ time. Conversely, it is possible for each of the $r_{i}$ to upcast a message of size $O(\log n)$ to the root $r$ in $O(D+\sqrt{n})$ time.

Suppose each vertex has a unique ID. The component ID of $F_{i}$ is defined to be the ID of $r_{i}$. The component ID can be broadcast to every vertex in the component in $O(\sqrt{n})$ rounds. We can then assume that the root $r$ knows the topology of the contracted tree where each component is contracted into a single vertex. This can be done if every root $r_{i}$ upcasts a message about the component ID of its parent and itself.

To compute pre $(u)$, each root $r_{i}$ in each component first calculate the size of $F_{i}$ then upcast it to $r$. Since $r$ knows the topology of the contracted tree, $r$ can calculate the size of each subtree rooted at each of $r_{i}$. Then $r$ downcasts the size of subtree rooted at $r_{i}$ back to $r_{i}$. Now each $F_{i}$

[^1]computes its preorder number internally in $O(\sqrt{n})$ time assuming $r_{i}$ has number 0 . During the computation, each $r_{i}$ also records what its preorder number is supposed to be if the depth-first search started from the root of its parent component. Finally, each $r_{i}$ upcasts this number to $r$ and then $r$ computes the correct offset for each subtree and downcasts the offsets back to the $r_{i}$. After adding the offset internally, we get the correct preorder number.

To compute low $(u)$, initially each vertex $u$ computes $\min \left(\operatorname{pre}(u), \min _{u v \in G_{i}} \operatorname{pre}(v)\right)$ in constant rounds. Then the problem becomes aggregating the minimum in the subtree $T_{u}$ for each $u$. First, each $r_{i}$ computes the minimum in $F_{i}$ in $O(\sqrt{n})$ time and then upcasts to $r$. Using the information, $r$ calculates the minimum of the subtrees rooted at each $r_{i}$ and downcasts to each $r_{i}$. Now each $r_{i}$ sends the minimum to its parent via the inter-component links. The parent replace its minimum if it is smaller. Finally, each component $F_{i}$ internally updates the minimum toward the root $r_{i}$. Then each vertex has the correct minimum. $\operatorname{high}(u)$ can be computed in the same way.

Therefore, the step of computing the bridges in $T$ of $G_{i}+T$ takes $O\left(D+\sqrt{n} \log ^{*} n\right)$ time. Each invocation of $\operatorname{Test}(T, \kappa)$ takes $O\left(\frac{\log n}{\epsilon^{2}}\left(D+\sqrt{n} \log ^{*} n\right)\right)$ time.

### 2.3 Running Time

Now we analyze the running time of Algorithm [1. The outerloop runs for $O(\log n)$ iterations. Therefore, the tree packing, Line [7, is executed $O(\log n)$ times, each taking $O\left(\log ^{10} n / \epsilon^{14}(D+\right.$ $\left.\sqrt{n} \log ^{*} n\right)$ ) rounds.

Let $k=O(\log (n W))$ be the largest index such that $X_{k} \leq n W$. The total number of iterations that the innerloop runs is at most

$$
\sum_{i=0}^{k} \log _{1+\epsilon^{\prime}}\left(\frac{1+\epsilon^{\prime}}{1-\epsilon^{\prime}} \cdot \frac{X_{i+1}}{X_{i}}\right)=O(k)+\sum_{i=0}^{k} \log _{1+\epsilon^{\prime}} \frac{X_{i+1}}{X_{i}}=O(k)+\log _{1+\epsilon^{\prime}}\left(X_{k+1}\right)=O(\log n / \epsilon)
$$

Therefore, $\operatorname{Test}(T, \kappa)$ is invoked at most $O\left((\log n / \epsilon) \cdot\left(\log ^{10} n / \epsilon^{14}\right)\right)$ times, each taking $O\left(\left(\log n / \epsilon^{2}\right)(D+\right.$ $\left.\sqrt{n} \log ^{*} n\right)$ ) rounds.

The total running time is

$$
\begin{aligned}
& O\left(\log n \cdot\left(\log ^{10} n / \epsilon^{14}\right)\left(D+\sqrt{n} \log ^{*} n\right)\right)+\left(\log ^{11} n / \epsilon^{15}\right) \cdot\left(\left(\log n / \epsilon^{2}\right)\left(D+\sqrt{n} \log ^{*} n\right)\right) \\
& =O\left(\left(\log ^{12} n / \epsilon^{17}\right) \cdot\left(D+\sqrt{n} \log ^{*} n\right)\right)=\tilde{O}(D+\sqrt{n})
\end{aligned}
$$

Remark 2.7. The total iterations of the outerloop and innerloop in Algorithm 1 can be reduced to $O(1)$ and $O(1 / \epsilon)$ by first approximating $\lambda$ within constant factor by Ghaffari and Kuhn's algorithm. Then, we can reduce our running time to $O\left(\left(\log ^{11} n / \epsilon^{17}\right)\left(D+\sqrt{n} \log ^{*} n\right)\right)$.

The exponent of the $\log n$ and the $\epsilon$ in our running time depends heavily on the size of the greedy tree packing in Lemma 2.1. If one can show that $O\left(\lambda^{a} \log ^{b} n\right)$ trees is sufficient, then our running time can be improved to $O\left(\left(\log ^{2+a+b} n / \epsilon^{2 a+3}\right) \cdot\left(D+\sqrt{n} \log ^{*} n\right)\right)$ rounds. Using Ghaffari and Kuhn's algorithm to approximate $\lambda$ within a constant (Remark 2.7), we can get a running time of $O\left(\left(\log ^{1+a+b} n / \epsilon^{2 a+3}+\left(\log ^{2} n \log \log n\right) / \epsilon^{5}\right) \cdot\left(D+\sqrt{n} \log ^{*} n\right)\right)$. For comparison, Karger [9] showed that a greedy tree packing of size $O(\lambda \log n)$ is enough for any minimum cut to be crossed at most twice by some tree. It will be interesting to see if the number of trees in Lemma 2.1] can be reduced.

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[^1]:    ${ }^{\dagger}$ It can be the case that $v$ is the parent of $u$ in $T$, which happens when there are parallel edges between $u$ and $v$ in $G_{i}+T$, and one of them is in $T$. Note that an edge is not a bridge if it is a multiedge.

