# Distributed Computing Made Secure: A New Cycle Cover Theorem* 

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#### Abstract

In the area of distributed graph algorithms a number of network's entities with local views solve some computational task by exchanging messages with their neighbors. Quite unfortunately, an inherent property of most existing distributed algorithms is that throughout the course of their execution, the nodes get to learn not only their own output but rather learn quite a lot on the inputs or outputs of many other entities. This leakage of information might be a major obstacle in settings where the output (or input) of network's individual is a private information (e.g., distributed networks of selfish agents, decentralized digital currency such as Bitcoin, voting systems).

While being quite an unfamiliar notion in the classical distributed setting, the notion of secure multi-party computation (MPC) is one of the main themes in the Cryptography community. Yet despite all extensive work in the area, no existing algorithm fits the framework of classical distributed models in which there are no assumptions on the graph topologies and only messages of bounded size are sent on the edges in each round.

In this paper, we introduce a new framework for secure distributed graph algorithms and provide the first general compiler that takes any "natural" non-secure distributed algorithm that runs in $r$ rounds, and turns it into a secure algorithm that runs in $\widetilde{O}(r \cdot D \cdot$ poly $(\Delta))$ rounds where $\Delta$ is the maximum degree in the graph and $D$ is its diameter. We also show that this is nearly (existentially) optimal for any round-by-round compiler for bounded degree graphs.

The main technical part of our compiler is based on a new cycle cover theorem: We show that the edges of every bridgeless graph $G$ of diameter $D$ can be covered by a collection of cycles such that each cycle is of length $\widetilde{O}(D)$ and each edge of the graph $G$ appears in $\widetilde{O}(1)$ many cycles. In fact, our construction can be made instance optimal with respect to each single edge. Letting $C_{e}$ be the shortest cycle containing $e$ in $G$, our cycle collection contains a cycle of length $\widetilde{O}\left(\left|C_{e}\right|\right)$ that covers $e$ for every $e \in G$, and in addition, each edge appears on $\widetilde{O}(1)$ many cycles. As a result, our compiler becomes instance optimal for bounded degree graphs.


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## Contents

1 Introduction ..... 1
1.1 Our Approach and Results ..... 2
1.1.1 From Security Requirements to Graph Structures ..... 2
1.1.2 Secure Simulation ..... 4
2 Our Techniques ..... 6
2.1 Low Congestion Cycle Covers ..... 6
2.2 From Low Congestion to Secure Simulation ..... 8
3 Preliminaries and Model ..... 10
3.1 Distributed Algorithms ..... 11
3.2 Cryptography with Perfect Privacy ..... 12
4 Low-Congestion Covers ..... 14
4.1 Cycle Cover ..... 14
4.1.1 Covering Non-Tree Edges ..... 15
4.1.2 Covering Tree Edges ..... 20
4.2 Private Neighborhood Cover ..... 28
$4.3 \quad P_{2}$ Cycle Cover ..... 31
5 Universally (Nearly) Optimal Covers ..... 32
5.1 Cycle Covers ..... 33
5.2 Private Neighborhood Cover and $P_{2}$ Covers ..... 35
6 Secure Simulation via Low-Congestion Covers ..... 36
6.1 Our Framework ..... 36
6.2 Secure Simulation of a Single Round ..... 37
6.3 The Final Secure Algorithm ..... 40
7 Distributed Computation of Low Congestion Covers ..... 42
7.1 Cycle Cover ..... 42
7.1.1 Distributed Algorithm for Covering Non-Tree Edges ..... 43
7.1.2 Analysis of Algorithm DistNonTreeCover ..... 48
7.1.3 Distributed Algorithm for Covering Tree Edges ..... 52
7.2 Additional Low Congestion Covers ..... 54
7.3 Pre-processing for Improved Cover Structures ..... 55
8 Discussion and Future Work ..... 56
A Balanced Partitioning of a Tree ..... 60
B Missing Details for Algorithm DistCycleCover ..... 61
C Distributed Construction of Neighborhood Cover in the Congest Model ..... 62

## 1 Introduction

In distributed graph algorithms (or network algorithms) a number of individual entities are connected via a potentially large network. Starting with the breakthrough by Awerbuch et al. [ALGP], and the seminal work of Linial [Lin92], Peleg [Pel00] and Naor and Stockmeyer [NS95], the area of distributed graph algorithm is growing rapidly. Recently, it receives considerably more theoretical and practical attention motivated by the spread of multi-core computers, cloud computing, and distributed databases. We consider the standard synchronous message passing model where in each round $O(\log n)$ bits can be transmitted over every edge (the CONGEST model).

The common principle underlying all distributed algorithms (regardless of the model specification) is that the input of the algorithm is given in a distributed format, and consequently the goal of each vertex is to compute its own part of the output, e.g., whether it is a member of a computed maximal independent set, its own color in a valid coloring of the graph, its incident edges in the minimum spanning tree, or its chosen edge for a maximal matching solution. In most distributed algorithms, throughout execution, vertices learn much more than merely their own output but rather collect additional information on the input or output of (potentially) many other vertices in the network. This seems inherent in many distributed algorithms, as the output of one node is used in the computation of another. For instance, most randomized coloring (or MIS) algorithms [Lub86, BE13, BEPS16, HSS16, Gha16, CPS17] are based on the vertices exchanging their current color with their neighbors in order to decide whether they are legally colored.

In cases where the data is sensitive or private, these algorithms may raise security concerns. To exemplify this point, consider a voting task for a distributed network, where the goal is to elect the candidate with most votes in the network. This is a rather simple distributed task: construct a BFS tree and let the nodes send the votes from the leaves to the root where each intermediate node sends to its parent in the tree, the sum of all votes for each candidate. While the output goal has been achieved, privacy has been compromised as intermediate nodes learn more information regarding the votes of their subtrees. As privacy in election processes is a fundamental property of any "democratic system", it is desirable to design a secure distributed voting algorithm, in which information of other nodes' votes do not get revealed throughout the course of execution. Additional motivation for secure distributed computation are settings that involve private medical data, networks of selfish agents with private utility functions or even decentralized digital currency such as the Bitcoin.

Whereas much effort in recent years has been devoted to improving the round complexities of various distributed algorithms, e.g., coloring, MIS, Lovász Local lemma [Suo13, BEK14, FHK16, BEPS16, Gha16, CPS17, CHL ${ }^{+}$17] and many more, the challenge of making these algorithms secure - was left behind. To this point, we have no distributed algorithms for general graphs in the standard CONGEST model that are secure and efficient compared to their non-secure counterparts.

While being a rather virgin objective in the distributed graph algorithm setting, the notion of secure multi-party computation (MPC) is one of the main themes in the Cryptography community. The goal of an MPC protocol is to allow parties to jointly compute a function of their inputs without revealing any about their input except the output of the function. There has been tremendous work on MPC protocols, from general feasibility results [Yao82, GMW87, BGW88, CCD88] to efficient protocols for specific functionalities [BNP08, BLO16]. Despite the large progress, almost all prior protocols require a private channel between every two nodes in the network in order to secure compute the desired function. This, of course, defeats the whole purpose of distributed computing.

Only a handful of works have considered a general graph interaction patterns [HLP11, HIJ ${ }^{+}$16, $\mathrm{GGG}^{+} 14, \mathrm{BGI}^{+}$14], or locality of MPC protocols [BGT13, CCG ${ }^{+}$, BIPW]. Unfortunately, they all have many major drawbacks: they rely on heavy computational assumptions (e.g., obfuscation), they assume a trusted setup phase (which is often not reusable), they require many rounds of interaction, they do not obey bandwidth limitations and they assume specific interaction patterns (e.g., star topology).

Despite some common interests, up to this point, both areas of distributed graph algorithms and secure multi-party computation have been developed in almost total isolation, each in its own community with different requirements and goals in mind. The distributed area puts most emphasis on locality, while the study of MPC focuses mostly on security. In this paper, we aim towards bridging this gap and combine the goods from both worlds, by tackling the following question:

How to design distributed algorithms that are both efficient (in terms of round and bandwidth complexity) and secure (where nothing is learned but the desired output)?

One tedious way to attack this challenge is to go through the most popular distributed problems (e.g., coloring, MIS, matching) and design a secure distributed algorithm for each of them one by one. Much more desirable, however, is to have a general recipe for adding security to existing algorithms, while incurring a small overhead in the round complexity.

Towards this end, we introduce the first general compiler that can take any (possibly insecure) distributed algorithm to one that has perfect security (a notion that will be explained next). The compiled algorithm respects the same bandwidth limitations, relies on no setup phase nor on any computational assumption and works for (almost) any graph, while paying a small overhead only in the number of rounds.

This quite general and powerful framework is made possible due to fascinating connections between "secure cryptographic definitions" and natural combinatorial graph properties. Most notably is the cycle cover of a graph. While cycle covers have been studied in the literature, e.g., the wellknown double cycle cover conjecture by Szekeresand and Seymour [JT92]; the Chinese postman problem by Edmond [EJ73], none of the known results satisfy our requires. Instead, we prove a new theorem regarding cycle covers with low congestion which we foresee being of independent interest and exploited in future work.

### 1.1 Our Approach and Results

### 1.1.1 From Security Requirements to Graph Structures

We demonstrate our approach with a simple example. Let $G$ be an $n$-vertex graph and suppose that each vertex $u$ has an input $x_{u}$. Suppose that each pair of neighbors $u, v$ in the graph (with inputs $\left.x_{u}, x_{v}\right)$ wishes to compute a function $f\left(x_{u}, x_{v}\right)$ securely, i.e., each party should learn the output $f\left(x_{u}, x_{v}\right)$ but "nothing more" (the precise notion of security will be later elaborated).

Kushilevitz [Kus89] showed that almost all (non-trivial) functions cannot be computed between $u$ and $v$ with this notion of privacy, and this was generalized for weaker notions of (information theoretic) privacy as well [FJS14]. To circumvent these barriers, Feige, Kilian and Naor [FKN94] (later generalized by [IK97]) suggested a "minimal model" called PSM ${ }^{1}$ where a third party $s$ aids

[^1]the computation. In the model, $u$ and $v$ share private randomness (not known to $s$ ) and each sends a single message to $s$ which depends on its own input and the shared randomness. This allows $s$ to compute the output $f(x, y)$ while learning nothing more.

In this paper we introduce Distributed PSM protocols which generalize the standard two party PSM protocols to general graphs. To provide secure communication between neighboring nodes $u, v$, we need to find a third node $s$ in the graph so that both $u$ and $v$ can communicate with $s$ without seeing each other's messages. This requirement translates into covering the edge ( $u, v$ ) with a cycle containing the edge. The number of rounds of the protocol (e.g., sending messages on the cycle to $w$ ) is proportional to the length of the cycle. Furthermore, as we wish to run such a protocol for all edges of the graph simultaneously, we need to cover all edges in the graph while having each individual edge participating in a small number of cycles (this is desirable since the bandwidth is limited).
Low Congestion Covers. These requirements motivate the genuine definition of (d, c)-cover $\mathcal{C}$ which is a collection of cycles of length at most $d$ such that each edge appears at least once and at most c many times on each of the cycles (the congestion of the cover). Given a (d, c)-cycle cover, we can have all vertices of the graph compute a function $f\left(x_{u}, x_{v}\right)$ for every edge in the graph simultaneously in $\widetilde{O}\left(\mathrm{~d}+\mathrm{c}\right.$ ) rounds ${ }^{2}$ (in the CONGEST models). A-priori, it is not clear that cycle covers that enjoy both low congestion and short lengths even exist, nor if it is possible to efficiently find them. Perhaps quite surprisingly, we prove the following theorem regarding cycle cover. Throughout, we will have an $n$-vertex graph $G$ and use the notation $\widetilde{O}$ is the "Big O" notation that hides polylog $(n)$ factors.

Theorem 1 (Low Congestion Cycle Cover). Every 2-edge connected graph with diameter $D$ has a ( $\mathrm{d}, \mathrm{c})$-cycle cover where $\mathrm{d}=\widetilde{O}(D)$ and $\mathrm{c}=\widetilde{O}(1)$. That is, the edges of $G$ can be covered by cycles such that each cycle is of length at most d and each edge participates in at most c cycles.

We note that this theorem is existentially tight, in the sense that the $\Theta(D)$ factor is necessary (consider a cycle graph). We also obtain a construction that is nearly optimal with respect to the best possible for the input graph $G$. For each edge $e$, let $C_{e}$ be the shortest cycle in $G$ containing $e$ and let $\operatorname{OPT}(G)=\max _{e}\left|C_{e}\right|$. Clearly, every (d, c)-cycle cover must satisfy that d $\geq \mathrm{OPT}(G)$. Note that whereas it might be that $\operatorname{OPT}(G)=O(D)$, in general $\mathrm{OPT}(G)$ might be much smaller. Thus, we show an instance optimal version of our cycle cover theorem:

Theorem 2 (Instance Optional Low Congestion Cycle Cover). For every 2-edge connected graph $G$, one can construct $a(\mathrm{~d}, \mathrm{c})$ cycle cover $\mathcal{C}$ where $\mathrm{d}=\widetilde{O}(\operatorname{OPT}(G))$ and $\mathrm{c}=\widetilde{O}(1)$. In addition, for every edge $e$, there is a cycle in $\mathcal{C}$ that covers $e$ and is of length $\widetilde{O}\left(\left|C_{e}\right|\right)$.

Although several variants of cycle cover problems have been considered in the literature, none of them fit our objective of having both short cycles and small congestion. For instance the $k$ cycle cover problem aims to minimize only the congestion, i.e., restricting each edge to appear on at most $k$ cycles while allowing arbitrary long cycles. On the other hand, in the minimum cycle cover problem that objective is minimize the total length of all cycles in the cover (i.e., instead of minimizing the length of the longest cycle as in our setting) [Fan97].

In order to use these covers in our compiler, we build them once, in a prepossessing step, and show that the covers of Theorem 1 can be constructed distributively in $\widetilde{O}(n+D \cdot \Delta)$ rounds (see

[^2]Lemma 11 and corollary 7). Alternatively, in Section 7 we also provide a much more efficient distributed construction on the expense of having somewhat worse bounds in the quality of the output covers.

Theorem 3 (Distributed Low Congestion Cycle Cover). For every 2-edge connected graph $G$ with diameter $D$ and every $\epsilon \in[0,1]$, there is a distributed algorithm that constructs a ( $\widetilde{O}\left(4^{1 / \epsilon}\right.$. $\left.\mathrm{OPT}(G)), \widetilde{O}\left(n^{\epsilon}\right)\right)$ cycle cover in $\widetilde{O}\left(4^{1 / \epsilon} . \mathrm{OPT}(G)+n^{\epsilon}\right)$ rounds.

Our low congestion cycle covers turn out to be a basic building block for constructing two other, more complex, covers which we need in our final secure compiler. The first is private neighborhood cover which generalizes the standard notion of neighborhood covers introduced by Awerbuch et al. [ABCP98]. Roughly speaking, the private neighborhood cover of a graph $G=$ $(V, E)$ is a collection of $n$ trees, one per node $u_{i}$, where each tree $T\left(u_{i}\right) \subseteq G \backslash\left\{u_{i}\right\}$ contains all the neighbors of $u_{i}$ but do not contain $u_{i}$. Intuitively, the private neighborhood covers allow all neighbors $\Gamma\left(u_{i}\right)$ of all nodes $u_{i}$, to exchange a secret without $u_{i}$. Note that these covers exists if and only if the graph is 2 -vertex connected ${ }^{3}$. Similarly to low-congestion cycle covers, we define (d, c)-private neighborhood covers in which each tree $T\left(u_{i}\right)$ has depth at most d and each edge belongs to at most c many trees. This allows us to use all trees simultaneously in $\widetilde{O}(\mathrm{~d}+\mathrm{c})$ rounds.

Theorem 4 (Private Neighborhood Cover). Every 2-vertex connected graph with diameter $D$ and maximum degree $\Delta$, has a ( $\mathrm{d}, \mathrm{c}$ )-private neighborhood cover for $\mathrm{d}=\widetilde{O}(D \cdot \Delta)$ and $c=\widetilde{O}(D)$.

The second covering structure can be seen as a generalization of (edge) cycle covers. In $P_{2}$ cycle covers we wish to cover every $P_{2}$ path in the graph by a cycle. Intuitively, $P_{2}$ cycle covers provide private communication between every pair of nodes of distance 2 in the graph. We show:

Theorem 5 ( $P_{2}$ Cover). For every 2-vertex connected graph with diameter $D$ and maximum degree $\Delta$, there exist $a(\mathrm{~d}, \mathrm{c})-P_{2}$ cycle cover where $\mathrm{d}=\widetilde{O}(D \cdot \Delta)$ and $c=\widetilde{O}\left(D \cdot \Delta^{2}\right)$.

Similarly to Theorem 2, the dependency in $D$ in Theorems 4 and 5 can be replaced by the optimal dilation of the corresponding structure in the input graph $G$. Finally, both these covering structures can be constructed also distributively by extending the constructions of low-congestion cycle covers (see Lemmas 9 and 10 in Section 7).

### 1.1.2 Secure Simulation

The Security Notion. With this graph framework in mind, we return to our original of constructing secure distributed algorithms. We want to design an algorithm in which every node $u$ learns its desired output but does not learn anything else about the inputs and outputs of the rest of the nodes in the graph. In our framework, the topology of the graph is not considered private and is not protected by our security notion. While there are many possible ways to define this kind of privacy, we use the strongest possible notion of perfect privacy which is information theoretic and relies and no computation assumptions. This notion uses the existence of an (unbounded) simulator, with the following intuition: a node learns nothing, except its own output $y$, from the messages it receives throughout the execution of the algorithm, if a simulator can produce the same output while receiving only $y$ and the graph $G$.

[^3]We achieve security in what is known as the "semi-honest" model, where the adversary, acting as one of the nodes in the graph, is not allowed to deviate from the prescribed protocol, but can run arbitrary computation given all the messages it received. Moreover, we assume that the adversary does no collude with other nodes in the graph. It is possible to extend the security notion to include these notions, more on this is discussed at Section 8.

The General Compiler. In a distributed algorithm, every node has a state and in each round, it updates its state by applying a local function $f$ that depends on its immediate neighbors. Theorems 1, 4 and 5 provide the required graph framework for computing this function $f$ in a secure and efficient manner (using a distributed PSM) for all nodes in the graph simultaneously. The communication complexity of the secure protocol depends on the computational complexity of the function $f$. In almost all distributed algorithms, the local update function $f$ can be computed in polynomial time and as a result the communication overhead of the secure protocol is negligible. We call this family of functions natural and for simplicity of presentation we state our results for natural distributed algorithms. Finally, we can state our main result: any natural distributed algorithm can be compiled to an equivalent one (that is one that has the same output for each node) that is secure.

Theorem 6 (Secure Simulation). Let $G$ be a 2-vertex connected $n$-vertex graph with diameter $D$ and maximal degree $\Delta$. Let $\mathcal{A}$ be a natural distributed algorithm that runs on $G$ in $r$ rounds. Then, $\mathcal{A}$ can be transformed to an equivalent algorithm $\mathcal{A}^{\prime}$ with perfect privacy which runs in $\widetilde{O}(r D \cdot \operatorname{poly}(\Delta))$ rounds, using $\widetilde{O}(n+D \cdot \Delta)$ rounds of pre-processing.

By using Theorem 2, the diameter dependency in the bounds of Theorem 6 can be replaced by $\operatorname{OPT}(G)$. Hence, we obtain our simulation blows up the round complexity by factor $\widetilde{O}(\operatorname{OPT}(G)$. poly $(\Delta)$ ) where $\operatorname{OPT}(G)$ is the best possible round complexity for a secure computation over a single edge in $G$. We note that our compiler works for any distributed algorithm rather than only on natural ones. The number of rounds will be proportional to the space complexity of the algorithm (an explicit statement for any algorithm can be found in Remark 1). Moreover, to avoid the preprocessing step, one can use Theorem 3 with $\epsilon=1 / \sqrt{\log n}$, to get a secure simulation with $\widetilde{O}\left(r \cdot 2^{\sqrt{\log n}} \cdot D \cdot \operatorname{poly}(\Delta)\right)$ rounds (with no preprocessing). We observe that the barrier of [Kus89] can be extended to a cycle graph, which shows that the linear dependency in $D$ in the round complexity of our compiler is existentially unavoidable. Our results are summarized in Figure 1.

Applications for Known Distributed Algorithms. Theorem 6 enables us to compile almost all of the known distributed algorithms to a secure version of them, examples include computing majority of votes as well as many other examples which we briefly mention next. It is worth noting that deterministic algorithms for problems in which the nodes do not have any input cannot be made secure by our approach since these algorithms only depend on the graph topology which we do not try to hide. Our compiler is meaningful for algorithms where the nodes have input or for randomized algorithms which define a distribution over the output of the nodes. For instance, whereas the deterministic coloring algorithms cannot be made secure, the randomized coloring algorithms (see e.g., [BE13]) which sample a random legal coloring of the graph can be made secure. Specifically, we get a distributed algorithm that (legally) colors a graph (or computes a legal configuration, in general), while the information that each node learns at the end is as if a centralized entity ran the algorithm for the entire network, and revealed each node's output privately (i.e., revealing $v$ the final color of $v$ ).


Figure 1: An illustrated summary of our results.

MIS, Coloring, Matching and More. Our approach captures global (e.g., MST) as well as many local problems [NS95]. The MIS algorithm of Luby [Lub86] along with our compiler yields $\widetilde{O}(D \cdot \operatorname{poly}(\Delta))$ secure algorithm according to the notion described above. Slight variations of this algorithm also gives the $O(\log n)$-round $(\Delta+1)$-coloring algorithm (e.g., Algorithm 19 of [BE13]). Combining it with our compiler we get a secure $(\Delta+1)$ coloring ${ }^{4}$ algorithm with round complexity of $\widetilde{O}(D \cdot \operatorname{poly}(\Delta))$. Using the Matching algorithm of Israeli and Itai [II86] we get an $\widetilde{O}(D \cdot$ poly $(\Delta))$ secure maximal matching algorithm. Finally, another example comes from distributed algorithms for the Lovász local lemma (LLL) which receives a lot of attention recently [ $\left.\mathrm{BFH}^{+} 16, \mathrm{FG} 17, \mathrm{CP} 17\right]$ for the class of bounded degree graphs. Using [CPS17], most of these (non-secure) algorithms for defective coloring, frugal coloring, and list vertex-coloring can be made secure within $\widetilde{O}(D)$ rounds.

## 2 Our Techniques

### 2.1 Low Congestion Cycle Covers

We give an overview of our low congestion cycle cover of Theorem 1. Let $G=(V, E)$ be a 2-edge connected $n$-vertex graph with diameter $D$. We begin by observing that all but $2 n$ of the edges in the graph can be covered by edge-disjoint cycles of length at most $\log n$. Since the girth of any graph with at least $2 n$ edges is $\log n$, we can repeatedly add a short cycle (up to length $\log n$ ) to the collection and remove it from the graph. The main challenge is in covering these last $2 n$ edges, as these edges might be arbitrary with potentially a large diameter.

The remaining $2 n$ are covered as follows. We construct a BFS tree $T$ in the graph $G$. All the remaining uncovered edges in the graph are either tree edges or non-tree edges. Our cover consists of two procedures, where the first constructs a low congestion cycle cover for the non-tree edges and the second covers the tree edges.

[^4]Covering the Non-Tree Edges. Let $E^{\prime}$ be the set of uncovered non-tree edges. The main difficulty in covering $E^{\prime}$ stems from the fact that the diameter ${ }^{5}$ of $G \backslash T$ might be large (e.g., $\Omega(n)$ ). Hence, to cover the edges of $E^{\prime}$ by short cycles, one must use the edges of $T$. A naïve approach is to cover every edge $e=(u, v)$ in $E^{\prime}$ by taking its fundamental cycle in $T$ (i.e., using the $u-v$ path in $T)$. Although this yields cycles of optimal lengths, the congestion on the tree edges might become $\Omega(n)$. The key challenge is to use the edges of $T$ (as we indeed have to) in a way that cycles enjoy the small diameter of the graph while not overloading any tree edge more than $\widetilde{O}(1)$ times.

Our approach is based on using the tree $T$ edges only for the purpose of connecting nodes that are somewhat close to each other, and thus the path between them should not occupy too many edges of the tree. To realize this approach, we define a specific way of partitioning the nodes of the tree $T$ to blocks according to $E^{\prime}$. In a very rough manner, a block would consist of a set of nodes that have few incident edges in $E^{\prime}$. To define these blocks, we number the nodes based on postorder traversal in $T$ and partition them into blocks containing nodes with consecutive numbering. The density of a block $B$ is the number of edge in $E^{\prime}$ with one endpoint in $B$. The blocks are partitioned such that no block has large density (at most some constant), and moreover, the number of blocks is not too large (say, at most $n / 8$ ). The only exception is a block containing a single node, in such a case its density can be unbounded (i.e., even $\Omega(n)$ ). The end goal is to connect nodes by their tree path only if they reside in the same block.

We then consider the contracted graph obtained by contracting all nodes in a given block into one supernode and connecting two supernodes $B_{1}$ and $B_{2}$, if there is an edge in $E^{\prime}$ whose one endpoint is in $B_{1}$, and the other endpoint is in $B_{2}$. This graph is in fact a multigraph, that might contain self-loops or multi-edges. The key idea is that the contracted graph contains only $n^{\prime}=n / 8$ nodes, and hence we can reuse the girth approach from before, and repeatedly find short cycles (length $\log n^{\prime}$ ) in it until we are left with at most $2 n^{\prime}=n / 4$ edges. These cycles are translated to cycles in the origin graph $G$ by using the tree paths $\pi(u, v, T)$ between nodes $u, v$ belonging to the same supernode (block).

Our key insight is that eventhough paths between two nodes in a block might be long, we show that every tree edge is "used" by at most two blocks. That is, for each edge $e$ of the tree, there are at most 2 blocks such that the tree path $\pi(u, v)$ of nodes $u, v$ in the block passes through $e$. (If a block has only a single node, then it will use no tree edges.) Since the blocks have constant density, we are able to bound the congestion on the edge $e$. The translation of cycles in the contracted graph to cycles in the original graph yields $O(D \log n)$-length cycles in the original graph where every edge belongs to $O(1)$ cycles.

The above step already covered all but $2 n^{\prime}=n / 4$ edges. We continue this process $\log n$ times until all edges of $E^{\prime}$ are covered, and thus get a $\log n$ factor in the congestion. We note that while producing short cycles of small congestion that cover all the non-tree edges with low congestion, these cycles might be non-simple. To handle that, we add an additional "cleanup" step (procedure SimplifyCycles) which takes the output collection of non-simple cycles and produces a collection of simple ones. In this process, some of the edges in the non-simple cycles might be omitted, however, we prove that only tree edges might get omitted and all non-tree edges remain covered by the simple cycles. This concludes the high level idea of covering the non-tree edges. We note the our blocking definition is quite useful also for distributed implementations. The reason is that although the blocks are not independent, in the sense that the tree path connecting two nodes in a given

[^5]block pass through other blocks, this independence is very limited. The fact that each tree edge is used in the tree paths for only two blocks allows us also the work distributively on many blocks simultaneously (see Section 7).
Covering the Tree Edges. Covering the tree edges turns out to be the harder case where new ideas are required. Specifically, whereas in the non-tree edge our goal is to find cycles that use the tree edge as rarely as possible, here we aim to find cycles that cover all edges in the tree, but still avoiding a particular tree edge from participating in too many cycles.

The algorithm for covering the tree edges is recursive, where in each step we split the tree into two edge disjoint subtrees $T_{1}, T_{2}$ that are balanced in terms of number of edges. To perform a recursive step, we would like to break the problem into two independent subproblems, one that covers the edges of $T_{1}$ and the other that covers the edges of $T_{2}$. However, observe that there might be edges $(u, v) \in T_{1}$ where the only cycle that covers them ${ }^{6}$ passes through $T_{2}$ (and vice versa). For every such node $u \in T_{1}$, let $s(u)$ be the first node in $T_{2}$ that appears on the fundamental cycle of the edge $(u, v)$.

To cover these tree edges, we employ two procedures, one on $T_{1}$ and the other on $T_{2}$ that together form the desired cycles (for an illustration, see Figures 9 and 11). First, we mark all nodes $u \in T_{1}$ such that their $s(u)$ is in $T_{2}$. Then, we use an Algorithm called TreeEdgeDisjointPath (see Lemma 4.3.2 [Pel00]) which solves the following problem: given a rooted tree $T$ and a set of $2 k$ marked nodes $M \subseteq V(T)$ for $k \leq n / 2$, find a matching of these vertices $\left\langle u_{i}, u_{j}\right\rangle$ into pairs such that the tree paths $\pi\left(u_{i}, u_{j}, T\right)$ connecting the matched pairs are edge-disjoint.

We employ Algorithm TreeEdgeDisjointPath on $T_{1}$ with the marked nodes as described above. Then for every pair $u_{i}, u_{j} \in T_{1}$ that got matched by Algorithm TreeEdgeDisjointPath, we add a virtual edge between $s\left(u_{i}\right)$ and $s\left(u_{j}\right)$ in $T_{2}$. Since this virtual edge is a non-tree edge with both endpoints in $T_{2}$, we have translated the dependency between $T_{1}$ and $T_{2}$ to covering a non-tree edge. At that point, we can simply use Algorithm NonTreeCover on the tree $T_{2}$ and the non-virtual edges. This computes a cycle collection which covers all virtual edges $\left(s\left(u_{i}\right), s\left(u_{j}\right)\right)$. In the final step, we replace each virtual edge $\left(s\left(u_{i}\right), s\left(u_{j}\right)\right)$ with the edge disjoint tree path $\left\langle u_{i}, u_{j}\right\rangle$ and the paths between $u_{i}$ and $s\left(u_{i}\right)$ (as well as the path connecting $u_{j}$ and $s\left(u_{j}\right)$ ).

This above description is simplified and avoids many details and complications that we had to address in the full algorithm. For instance, in our algorithm, a given tree edge might be responsible for the covering of up to $\Theta(D)$ many tree edges. This prevents us from using the edge disjoint paths of Algorithm TreeEdgeDisjointPath in a naïve manner. In particular, our algorithm has to avoid the multiple appearance of a given tree edge on the same cycle as in such a case, when making the cycle simple that tree edge might get omitted and will no longer be covered. See Section 4.1 for the precise details of the proof, and see Figure 2 for a summary of our algorithm. The construction of nearly optimal covers is in Section 5. The high level approach is to first construct the neighborhood cover for the graph and then apply the above algorithm is each cluster of the neighborhood cover.

### 2.2 From Low Congestion to Secure Simulation

Consider a distributed algorithm $\mathcal{A}$ that runs in $r$ rounds. In a broad view, $\mathcal{A}$ can be considered as a collection of $r$ functions $f_{1}, \ldots, f_{r}$. At round $i$, a node $u$ holds a state $\sigma_{i}$ and needs to update its state according to a function $f_{i}$ that depends on $\sigma_{i}$ and the state of its immediate neighbors. We assume that the final state $\sigma_{r}$ is the final output of the algorithm for node $u$.

[^6]
## Algorithm CycleCover $(G=(V, E))$

1. While there is a cycle of uncovered edges of length at most $\log n$, add it to $\mathcal{C}$.
2. Construct a BFS tree $T$ of $G$.
3. Let $\widetilde{E} \subseteq E$ be the subset of all uncovered non-tree edges.
4. Repeat $\log n$ times:
(a) Partition the nodes of $T$ with block density b with respect to $\widetilde{E}$.
(b) While there are $t$ edges $\left(u_{1}, v_{1}\right), \ldots,\left(u_{t}, v_{t}\right) \in \widetilde{E}$ for $t \leq \log n$ such that for all $i \in[t-1]$, $v_{i}$ and $u_{i+1}$ are in the same block and $v_{t}$ and $u_{1}$ are in the same block (with respect to the partitioning $\mathcal{B}$ ): Add the cycle $u_{1} \rightarrow v_{1} \rightarrow \pi\left(v_{1}, u_{2}\right) \rightarrow v_{2} \rightarrow \pi\left(v_{2}, u_{3}\right) \rightarrow v_{3} \rightarrow$ $\ldots \rightarrow v_{t} \rightarrow u_{1}$ to $\mathcal{C}$.
5. $\mathcal{C} \leftarrow \mathcal{C} \cup \operatorname{TreeEdgeCover}(T)$ (see Figure 12).
6. Output SimplifyCycles $(\mathcal{C})$.

Figure 2: Procedure for covering non-tree edges.

Our goal is to simulate this process of computing $\sigma_{1}, \sigma_{2}, \ldots, \sigma_{r}$, however, in an oblivious way without knowing any $\sigma_{i}$ except the last one $\sigma_{r}$ which contains the final output. Thus, we will have the node $u$ hold an "encrypted" state, $\widehat{\sigma}_{i}$, instead of the actual state $\sigma_{i}$. This encryption uses a a random mask $R$ such that $\widehat{\sigma}_{i} \oplus R=\sigma_{i}$. The key $R$ will be secret shared among the neighbors of $u$. That is, each of them will hold a single random string (called a share) $R^{j}$, such that $R=\bigoplus R^{j}$.

The neighbors of $u$, in turn, hold an encrypted version of their own state as well. The key for their state is secret shared among their neighbors. Thus, the information required in order to compute the function $f_{i}$ is now spread out among the 2 -neighborhood of $u$. Define a related function $f_{i}^{\prime}$ to be a function that gets an encrypted state of $u$ and its neighbors and the corresponding keys. Then the function decrypts the states, computes $f_{i}$ and finally re-encrypts the new state using a new encryption key. Our goal is to enable $u$ to compute the function $f_{i}^{\prime}$ that depends on its 2-neighborhood while learning but the output. Actually, we need all nodes $u \in V$ to be able to compute $f_{i}^{\prime}$ simultaneously.

To achieve this, we use a PSM protocol where $u$ is the server and its 2-neighborhood are the parties. The first step of a PSM protocol, is for all parties to share private common randomness (not known to $u$ ). This is exactly the reason for introducing the notion of low congestion private neighborhood cover. Using this cover, the neighborhood can communicate privately, and moreover, this can be done for all nodes simultaneously (with independent randomness for each neighborhood), with low congestion on the edges. The number of rounds is proportional to the depth of the trees computed in the neighborhood cover.

The last component missing in order to run a PSM protocol by $u$ with its $2^{\text {nd }}$-neighborhood is a private channel between each party and $u$. The immediate neighbors of $u$ have this private channel by definition. However, 2-hop neighbors do not. To simulate a private channel, each 2-hop neighbor $w$ secret shares its message to two parts. Each individual part leaks no information about
the message. Then, $w$ sends the two parts via two vertex disjoint paths to $u$. This is indeed the motivation for introducing the notion of $P_{2}$ cycle covers. Indeed, covering the 2-length path between $w$ and $u$ by a cycle, provides two vertex disjoint paths between $w$ and $u$. Since only the node $u$ gets both parts and it is the only node in the graph that can learn the message of $w$, hence we have established a virtual private channel to $u$. Moreover, since the $P_{2}$ cycle cover has low congestion, all 2-hop neighbors in the graph can communicate simultaneously with their target with bounded congestion. Again, the number of rounds of this communication is proportional to the length of the cycles.

Our secure compiler works round by round, where all nodes in the graph apply the PSM protocol for every round of the original algorithm $\mathcal{A}$. After securely simulating all the rounds of $\mathcal{A}$, each node holds an encrypted version of the last state, which contains nothing but the desired output. The neighbors of each node $u$ send it the encryption key (keys used solely for the last encryption), this allows $u$ to decrypt and obtain the final output of the algorithm. As a result, each node $u$ sees only perfectly encrypted states along the whole duration of the algorithm, where it proceeds from one state to the other using the PSM protocol. The security of the PSM protocol ensures that it learns nothing but the next encrypted state. A summary of the algorithm for a single node $u$ is given in Figure 3.

## The description of the simulation of algorithm $\mathcal{A}$ with respect to $u$.

1. For each round $i=1 \ldots r$ do:
(a) $u$ holds the encrypted state $\widehat{\sigma_{i}}$.
(b) For each neighbor $v$ of $u$ samples new encryption keys.
(c) Run a distributed PSM protocol for with server $u$ to compute the next state of $u$ :
i. Using the private 2-neighborhood cover the distance 2 neighbors of $u$ share private randomness.
ii. Nodes with distance 2 from $u$ use the $P_{2}$ cycle cover to simulate private channels to $u$.
(d) $u$ learns its new encrypted state $\widehat{\sigma}_{i+1}$.
2. Each neighbor of $u$ sends the final encryption key to $u$.
3. Using these keys, $u$ computes its final output $\sigma_{r}$.

Figure 3: A schematic overview of the simulated algorithm.

## 3 Preliminaries and Model

Unless stated otherwise, the logarithms in this paper are base 2. For a distribution $\mathcal{D}$ we denote by $x \leftarrow \mathcal{D}$ an element chosen from $\mathcal{D}$ uniformly at random. For an integer $n \in \mathbb{N}$ we denote by $[n]$ the set $\{1, \ldots, n\}$. We denote by $U_{n}$ the uniform distribution over $n$-bit strings. For two distributions (or random variables) $X, Y$ we write $X \equiv Y$ if they are identical distributions. That is, for any $x$ it holds that $\operatorname{Pr}[X=x]=\operatorname{Pr}[Y=x]$.

Graph Notations. For a tree $T \subseteq G$, let $T(z)$ be the subtree of $T$ rooted at $z$, and let $\pi(u, v, T)$ be the tree path between $u$ and $v$, when $T$ is clear from the context, we may omit it and simply write $\pi(u, v)$. The fundamental cycle $C_{e, T}$ of an edge $e=(u, v) \notin T$ is the cycle formed by taking $e$ and the tree path between $u$ and $v$ in $T_{0}$, i.e., $C_{e, T}=e \circ \pi(u, v, T)$. For $u, v \in G$, let $\operatorname{dist}(u, v, G)$ be the length (in edges) of the shortest $u-v$ path in $G$. For every integer $i \geq 1$, let $\Gamma_{i}(u, G)=\left\{v \mid \operatorname{dist}_{G}(u, v) \leq i\right\}$. When $i=1$, we simply write $\Gamma(u, G)$. Let $\operatorname{deg}(u, G)=|\Gamma(u, G)|$ be the degree of $u$ in $G$. For a subset of edges $E^{\prime} \subseteq E(G)$, let $\operatorname{deg}\left(u, E^{\prime}\right)=\left|\left\{v:(u, v) \in E^{\prime}\right\}\right|$ be the number of edges incident to $u$ in $E^{\prime}$. For a subset of nodes $U$, let $\operatorname{deg}\left(U, E^{\prime}\right)=\sum_{u \in U} \operatorname{deg}\left(u, E^{\prime}\right)$. For a $u_{1}-u_{2}$ path $P_{1}$ and an $u_{2}-u_{3}$ path, the path $P_{1} \circ P_{2}$ is the concatenation of the two paths. For a subset of vertices $S_{i} \subseteq V(G)$, let $G\left[S_{i}\right]$ be the induced subgraph on $S_{i}$.
Fact 1. [Moore Bound, [Bol04]] Every n-vertex graph $G=(V, E)$ with at least $2 n^{1+1 / k}$ edges has a cycle of length at most $2 k$.

### 3.1 Distributed Algorithms

The Communication Model. We use a standard message passing model, the CONGEST model [Pel00], where the execution proceeds in synchronous rounds and in each round, each node can send a message of size $O(\log n)$ to each of its neighbors. In this model, local computation is done for free at each node and the primary complexity measure is the number of communication rounds. Each node holds a processor with a unique and arbitrary ID of $O(\log n)$ bits. Throughout, we make an extensive use of the following useful tool, which is based on the random delay approach of [LMR94].

Theorem 7 ([Gha15, Theorem 1.3]). Let $G$ be a graph and let $A_{1}, \ldots, A_{m}$ be $m$ distributed algorithms in the CONGEST model, where each algorithm takes at most d rounds, and where for each edge of $G$, at most c messages need to go through it, in total over all these algorithms. Then, there is a randomized distributed algorithm (using only private randomness) that, with high probability, produces a schedule that runs all the algorithms in $O(\mathrm{c}+\mathrm{d} \cdot \log n)$ rounds, after $O\left(\operatorname{d~}_{\log }{ }^{2} n\right)$ rounds of pre-computation.

A Distributed Algorithm. Consider an $n$-vertex graph $G$ with maximal degree $\Delta$. We model a distributed algorithm $A$ that works in $r$ rounds as describing $r$ functions $f_{1}, \ldots, f_{r}$ as follows. Let $u$ be a node in the graph with input $x_{u}$ and neighbors $v_{1}, \ldots, v_{\Delta}$. At any round $i$, the memory of a node $u$ consists of a state, denoted by $\sigma_{i}$ and $\Delta$ messages $m_{v_{1} \rightarrow u \ldots,} m_{v_{\Delta} \rightarrow u}$ that were received in the previous round.

Initially, we set $\sigma_{0}$ to contained only the input $x_{u}$ of $u$ and its ID and initialize all messages to $\perp$. At round $i$ the node $u$ updates its state to $\sigma_{i+1}$ according to its previous state $\sigma_{i}$ and the message from the previous round, and prepares $\Delta$ messages to send $m_{u \rightarrow v_{1}}, \ldots, m_{u \rightarrow v_{\Delta}}$. To ease notation (and without loss of generality) we assume that each state contains the ID of the node $u$, thus, we can focus on a single update function $f_{i}$ for every round that works for all nodes. The function $f_{i}$ gets the state $\sigma_{i}$ and messages $m_{v_{1} \rightarrow u} \ldots, m_{v_{\Delta} \rightarrow u}$, and randomness $s_{i}$ and outputs the next state and outgoing message:

$$
\sigma_{i}, m_{u \rightarrow v_{1}}, \ldots, m_{u \rightarrow v_{\Delta}} \leftarrow f\left(\sigma_{i-1}, m_{v_{1} \rightarrow u}, \ldots, m_{v_{\Delta} \rightarrow u}, s_{i}\right) .
$$

At the end of the $r$ rounds, each node $u$ has a state $\sigma_{r}$ and a final output of the algorithm. Without loss of generality, we assume that $\sigma_{r}$ is the final output of the algorithm (we can always modify $f_{r}$ accordingly).

Natural Distributed Algorithms. We define a family of distributed algorithms which we call natural, which captures almost all known distributed algorithms. A natural distributed algorithm has two restrictions for any round $i$ : (1) the size the state is bounded by $\left|\sigma_{i}\right| \leq \Delta \cdot \operatorname{polylog}(n)$, and (2) the function $f_{i}$ is computable in polynomial time. The input for $f_{i}$ is the state $\sigma_{i}$ and at most $\Delta$ message each of length $\log n$. Thus, the input length $m$ for $f_{i}$ is bounded by $m \leq \Delta \cdot \operatorname{polylog}(n)$, and the running time should be polynomial in this input length.

We introduce this family of algorithms mainly for simplifying the presentation of our main result. For these algorithms, our main statement can be described with minimal overhead. However, our results are general and work for any algorithm, with appropriate dependency on the size of the state and the running time the function $f_{i}$ (i.e., the internal computation time at each node $u$ in round $i$ ).
Notations. We introduce some notations: For an algorithm $\mathcal{A}$, graph $G$, input $X=\left\{x_{v}\right\}_{v \in G}$ we denote by $\mathcal{A}_{u}(G, X)$ the random variable of the output of node $u$ while performing algorithm $\mathcal{A}$ on the graph $G$ with inputs $X$ (recall that $\mathcal{A}$ might be randomized and thus the output is a random variable and not a value). Denote by $\mathcal{A}(G, X)=\left\{\mathcal{A}_{u}(G, X)\right\}_{u \in G}$ the collection of outputs (in some canonical ordering). Let $\operatorname{View}_{u}^{\mathcal{A}}(G, X)$ be a random variable of the viewpoint of $u$ in the running of the algorithm $\mathcal{A}$. This includes messages sent to $u$, its memory and random coin during all rounds of the algorithm.
Secure Distributed Computation. Let $\mathcal{A}$ be a distributed algorithm. Informally, we say that $\mathcal{A}^{\prime}$ computes $\mathcal{A}$ (or simulates $\mathcal{A}$ ) in a secure manner of $\mathcal{A}$ if when running the algorithm $\mathcal{A}^{\prime}$ every node $u$ learns the final output $\sigma_{r}$ of $\mathcal{A}$ but "nothing more". This notion is captured by the existence of a simulator and is defined below.

Definition 1 (Perfect Privacy). Let $\mathcal{A}$ be a distributed (possibly randomized) algorithm, that works in $r$ rounds. We say that an algorithm $\mathcal{A}^{\prime}$ computes $\mathcal{A}$ with perfect privacy if for every graph $G$, every $u \in G$ and it holds that:

1. Correctness: For every input $X=\left\{x_{v}\right\}_{v \in V}: \mathcal{A}(G, X) \equiv \mathcal{A}^{\prime}(G, X)$.
2. Perfect Privacy: There exists a randomized algorithm (simulator) Sim such that for every input $X=\left\{x_{v}\right\}_{v \in V}$ it holds that

$$
\operatorname{View}_{u}^{\mathcal{A}^{\prime}}(G, X) \equiv \operatorname{Sim}\left(G, x_{u}, \mathcal{A}_{u}(G, X)\right)
$$

This security definition is known as the "semi-honest" model, where the adversary, acting a one of the nodes in the graph, is not allowed to deviate from the prescribed protocol, but can run arbitrary computation given all the messages it received. Moreover, we assume that the adversary does no collude with other nodes in the graph. It is possible to extend the security notion to include these notions, more on this is discussed at Section 8.

### 3.2 Cryptography with Perfect Privacy

One of the main cryptographic tools we use is a specific protocol for secure multiparty computation that has perfect privacy. Feige Kilian and Naor [FKN94] suggested a model where two players having inputs $x$ and $y$ wish to compute a function $f(x, y)$ in a secure manner. They achieve this by each sending a single message to a third party that is able to compute the output of the function
$f$ from these messages, but learn nothing else about the inputs $x$ and $y$. For the protocol to work, the two parties need to share private randomness that is not known to the third party. This model was later generalized to multi-players and is called the Private Simultaneous Model [IK97], which we formally describe next.

Definition 2 (The PSM model). Let $f:\left(\{0,1\}^{m}\right)^{k} \rightarrow\{0,1\}^{m}$ be a $k$ variant function. A PSM protocol for $f$ consists of a pair of algorithms (PSM.Enc, PSM.Dec) where PSM.Enc: $\{0,1\}^{m} \times$ $\{0,1\}^{r} \rightarrow\{0,1\}^{t}$ and PSM.Dec: $\left(\{0,1\}^{t}\right)^{k} \rightarrow\{0,1\}^{m}$ such that

- For any $X=\left(x_{1}, \ldots, x_{k}\right)$ it holds that:

$$
\operatorname{Pr}_{R \in\{0,1\}^{r}}\left[\operatorname{PSM} . \operatorname{Dec}\left(\operatorname{PSM} \cdot \operatorname{Enc}\left(x_{1}, R\right), \ldots, \operatorname{PSM} \cdot \operatorname{Enc}\left(x_{k}, R\right)\right)=f\left(x_{1}, \ldots, x_{k}\right)\right]=1
$$

- There exists a randomized algorithm (simulator) Sim such that for $X=x_{1}, \ldots, x_{k}$ and for $R$ sampled from $\{0,1\}^{r}$, it holds that

$$
\left\{\operatorname{PSM} \cdot \operatorname{Enc}\left(x_{i}, R\right)\right\}_{i \in[k]} \equiv \operatorname{Sim}\left(f\left(x_{1}, \ldots, x_{k}\right)\right)
$$

The communication complexity of the PSM protocol is the encoding length $t$ and the randomness complexity of the protocol is defined to be $|R|=r$.

Theorem 8 (Follows from [IK97]). For every function $f:\left(\{0,1\}^{m}\right)^{k} \rightarrow\{0,1\}^{\ell}$ that is computable by an $s=s(m, k)$-space TM there is an efficient perfectly secure PSM protocol whose communication complexity and randomness complexity are $O\left(\mathrm{~km} \mathrm{\ell} \cdot 2^{2 s}\right)$.

We describe two additional tools that we will use, secret sharing and one-time-pad encryption.
Definition 3 (Secret Sharing). Let $x \in\{0,1\}^{n}$ be a message. We say $x$ is secret shared to $k$ shares by choosing $k$ random strings $x^{1}, \ldots, x^{k} \in\{0,1\}^{n}$ conditioned on $x=\bigoplus_{j=1}^{k} x^{j}$. Each $x^{j}$ is called a share, and notice that the joint distribution of any $k-1$ shares is uniform over $\left(\{0,1\}^{n}\right)^{k-1}$.

Definition 4 (One-Time-Pad Encryption). Let $x \in\{0,1\}^{n}$ be a message. A one-time pad is an extremely simple encryption scheme that has information theoretic security. For a random key $K \in\{0,1\}^{n}$ the "encryption" of $x$ according to $K$ is $\widehat{x}=x \oplus K$. It is easy to see that the encrypted message $\widehat{x}$ (without the key) is distributed as a uniform random string. To decrypt $\widehat{x}$ using the key $K$ we simply compute $x=\widehat{x} \oplus K$. The key $K$ might be references as the encryption key or decryption key.

Paper Organization. In Section 4 we describe the centralized constructions of our low-congestion covers. We start by showing the construction of cycle covers (in Section 4.1). We then use the cycle cover construction to compute private neighborhood cover and $P_{2}$ covers (in Sections 4.2 and 4.3 respectively). Section 6 describes the secure simulation which generalizes PSM to general graphs. Finally, Section 7 considers the distributed construction of our low congestion covers.

## 4 Low-Congestion Covers

### 4.1 Cycle Cover

We give the formal definition of a cycle cover and prove our main theorem regarding low-congestion cycle covers. Intuitively, a cycle cover is a collection is cycles in the graph such that each edge is covered by at least one cycle from the collection. We care about two main parameters regarding the cycle cover that we wish to minimize: (1) cycle length: the maximal length of a cycle and (2) edge congestion: the maximal number of cycles an edge participates in.

Definition 5 (Low-Congestion Cycle Cover). For a given graph $G=(V, E)$, a (d, c) low-congestion cycle cover $\mathcal{C}$ of $G$ is a collection of cycles that cover all edges of $G$ such that each cycle $C \in \mathcal{C}$ is of length at most $O(\mathrm{~d})$ and each edge appears in at most $O(\mathrm{c})$ cycles in $\mathcal{C}$. That is, for every $e \in E$ it holds that $1 \leq|\{C \in \mathcal{C}: e \in C\}| \leq O$ (c).

We also consider partial covers, that cover only a subset of edges $E^{\prime}$. We say that a cycle cover $\mathcal{C}$ is a ( $\mathrm{d}, \mathrm{c}$ ) cycle cover for $E^{\prime} \subseteq E$, if all cycles are of length at most $D$, each edge of $E^{\prime}$ appears in at least one of the cycles of $\mathcal{C}$, and no edge in $E(G)$ appears in more than $c$ cycles in $\mathcal{C}$. That is, in this restricted definition, the covering is with respect to the subset of edges $E^{\prime}$, however, the congestion limitation is with respect to all graph edges.

The main contribution of this section is an existential result regarding cycle covers with low congestion. Namely, we show that any graph that is 2-edge connected has a cycle cover where each cycle is at most the diameter of the graph (up to $\log n$ factors) and each edge is covered by $O(\log n)$ cycles. Moreover, the proof is actually constructive, and yields a polynomial time algorithm that computes such a cycle cover.

Theorem 1. For every n-vertex graph $G$ with diameter $D$ that is 2-edge connected, there exists a (d, c)-cycle cover with $\mathrm{d}=O(D \log n)$ and $\mathrm{c}=O\left(\log ^{3} n\right)$.

Our construction of a (d, c)-cycle cover $\mathcal{C}$ is composed of 3 different covers $\mathcal{C}_{1}, \mathcal{C}_{2}, \mathcal{C}_{3}$, each covering a different subset of edges $E_{1}, E_{2}, E_{3}$ respectively, such that $E_{1} \cup E_{2} \cup E_{3}=E$. In a high-level overview, $\mathcal{C}_{1}$ will be constructed by a greedy algorithm and will contain a all edges of the graph except $2 n$. Then, $\mathcal{C}_{2}$ and $\mathcal{C}_{3}$ will be more precisely crafted to cover the remaining uncovered edges. Specifically, we compute a BFS tree $T$ on the graph $G$, and define $E_{2}$ to be all the non-tree edges of $E$ that are not covered by $\mathcal{C}_{1}$, and similarly we define $E_{3}$ to be all the tree edges of $E$ not covered by $\mathcal{C}_{1}$. We describe each cover separately. The pseudo-code for the algorithm is given in Figure 4. The algorithm uses two procedures, NonTreeCover and TreeCover which are given in Section 4.1.1 and Section 4.1.2 respectively.

Reducing to Linear Number of Edges. The first part of the algorithm computes a cycle cover with no congestion (that is, each edge participates in at most one cycle) that covers all edges except at most a linear number of edges in the graph. In particular, Item 1 of the CycleCover algorithm, can be seen as performing the following: it starts with graph $G^{\prime} \leftarrow G$ and as long as that $G^{\prime}$ has at least $2 n$ edges it finds the shortest cycle $C$ in $G$, adds $C$ to the cycle collection $\mathcal{C}_{1}$ and remove $C$ from $G^{\prime}$. In the next lemma, we show that this step covers all edges expect at most $2 n$ edges.

Lemma 1. For every n-vertex graph which is 2-edge connected, all graph edges, except for at most $2 n$ edges, can be covered by a $(\log n, 1)$-cycle cover. In particular, the output cover $\mathcal{C}_{1}$ of Algorithm CycleCover is a $(\log n, 1)$ cycle cover for all but $2 n$ edges in $G$.

## Algorithm CycleCover $(G=(V, E))$

1. While there is a cycle of uncovered edges of length at most $\log n$, add it to $\mathcal{C}_{1}$.
2. Construct a BFS tree $T$ of $G$ (with respect to edge set $E$ ).
3. Let $E_{2}$ be all uncovered non-tree edges, and let $E_{3}$ be all uncovered tree edges.
4. $\mathcal{C}_{2} \leftarrow \operatorname{NonTreeCover}\left(T, E_{2}\right)$.
5. $\mathcal{C}_{3} \leftarrow \operatorname{TreeCover}\left(T, E_{3}\right)$
6. Output $\mathcal{C}_{1} \cup \mathcal{C}_{2} \cup \mathcal{C}_{3}$.

Figure 4: Centralized algorithm for finding a cycle cover of a graph $G$.
Proof. The proof mainly uses the fact that while the graph many edges, then the firth is small. Specifically, using Fact 1 , with $k=\log n$ we get that the girth of a graph with at least $2 n$ edges is at most $2 \log n$. Hence, as long as that the graph has at least $2 n$ edges, a cycle of length $2 \log n$ can be found. We get that all but $2 n$ edges in $G$ are covered by edge-disjoint cycles of length $2 \log n$, and therefore $\mathcal{C}_{1}$ is a $(2 \log n, 1)$-cover for these edges.

The collection of all these edge distinct cycles is $\mathcal{C}_{1}$ and the edges covered are $E_{1}$. The other cycle covers will cover the rest of the edges. To cover the remaining $2 n$ edge, we construct a BFS tree $T$ in $G$ and define $E_{2}=E(G) \backslash\left\{E(T) \cup E_{1}\right\}$ to be all edges $E$ that where not covered by $\mathcal{C}_{1}$ and are not tree edges. Finally, set $E_{3}=E(T) \backslash E_{1}$ and note that $E_{1} \cup E_{2} \cup E_{3}=E$. The covering of $E_{2}$ and $E_{3}$ is described in Section 4.1.1 and Section 4.1.2 respectively.

### 4.1.1 Covering Non-Tree Edges

In this subsection, we show that the set of edges $E_{2}$, i.e., the set of non-tree edges the are not covered in the previous phase, can be covered by a $(D \log n, O(1))$-cycle cover denoted $\mathcal{C}_{2}$. Actually, what we show is slightly more general: if the tree is of depth $D(T)$ the length of the cycles is at most $O(D(T) \log n)$. Lemma 2 will be useful for covering the tree-edges as well and is used again in see next subsection (Section 4.1.2).

Lemma 2. Let $G=(V, E)$ be a n-vertex graph, let $T \subseteq G$ be a tree of depth $D(T)$ and let $E^{\prime} \subseteq E$ be such that $\left|E^{\prime}\right| \leq 2 n$. Then, there exists a $(D(T) \log n, \log n)$-cycle cover $\mathcal{C}_{2}$ for the edges of $E^{\prime}$. In addition, each cycle in $\mathcal{C}_{2}$ is used to cover $O(\log n)$ edges in $E^{\prime}$.

The rest of this subsection is devoted to the proof of Lemma 2. A key component in the proof is a partitioning of the nodes of the tree $T$ into blocks. The partitioning is based on a numbering of the nodes from 1 to $n$ and grouping nodes with consecutive numbers into blocks under certain restrictions. We define a numbering of the nodes

$$
N: V(T) \rightarrow[|V(T)|]
$$

by traversing the nodes of the tree in post order. That is, we let $N(u)=i$ if $u$ is the $i^{\text {th }}$ node traversed. Using this mapping, we proceed to defining a partitioning of the nodes into blocks and show some of their useful properties.

For a block $B$ of nodes, the notation $\operatorname{deg}\left(B, E^{\prime}\right)$ is the number of edges in $E^{\prime}$ that have an endpoint in the set $B$. We call this the density of block $B$ with respect to $E^{\prime}$. For a subset of edges $E^{\prime}$, and a density bound b (which will be set to a constant), an ( $\left.E^{\prime}, \mathrm{b}\right)$-partitioning $\mathcal{B}$ is a partitioning of the nodes of the graph into blocks that satisfies the following properties:

1. Every block consists of a consecutive subset of nodes (w.r.t. their $N(\cdot)$ numbering).
2. If a block $B$ has density $\operatorname{deg}\left(B, E^{\prime}\right)>\mathrm{b}$ then $B$ consists of a single node.
3. The total number of blocks is at most $4|\widetilde{E}| / \mathrm{b}$.

Claim 1. For any b and $E^{\prime}$, there exists an $\left(E^{\prime}, \mathrm{b}\right)$-partitioning partitioning of the nodes of $T$ satisfying the above properties.

Proof. This partitioning can be constructed by a greedy algorithm that traverses nodes of $T$ in increasing order of their numbering $N(\cdot)$ and groups them into blocks while the density of the block does not exceed b (see Figure 5 for the precise procedure). Indeed, properties 1 and 2

## Algorithm Partition $(T, \widetilde{E})$

1. Let $\mathcal{B}$ be an empty partition, and let $B$ be an empty block.
2. Traverse the nodes of $T$ in post-order, and for each node $u$ do:
(a) If $\operatorname{deg}(B, \widetilde{E})+\operatorname{deg}(u, \widetilde{E}) \leq \mathrm{b}$ add $u$ to $B$.
(b) Otherwise, add the block $B$ to $\mathcal{B}$ and initialize a new block $B=\{u\}$.

## 3. Output $\mathcal{B}$.

Figure 5: Partitioning procedure.
are satisfied directly by the construction. For property 3 , let $t$ be the number of blocks $B$ with $\operatorname{deg}\left(B, E^{\prime}\right) \leq \mathrm{b} / 2$. By the construction, we know that for any such block $B$ the block $B^{\prime}$ that comes after $B$ satisfies $\operatorname{deg}\left(B, E^{\prime}\right)+\operatorname{deg}\left(B^{\prime}, E^{\prime}\right)>\mathrm{b}$. Let $B_{1}, \ldots, B_{\ell}$ be the final partitioning. Then, we have $t$ pairs of blocks that have density at least b and the rest of the $(\ell-t / 2)$ blocks that have density at least b/2. Formally, we have

$$
\sum_{i=1}^{\ell} \operatorname{deg}\left(B_{i}, E^{\prime}\right)>t \mathrm{~b}+(\ell-t / 2) \mathrm{b} / 2=\ell \mathrm{b} / 2
$$

On the other hand, since it is a partitioning of $E^{\prime}$ we have that $\sum_{i=1}^{\ell} \operatorname{deg}\left(B_{i}, E^{\prime}\right)=2\left|E^{\prime}\right|$. Thus, we get that $\ell \mathrm{b} / 2 \leq 2\left|E^{\prime}\right|$ and therefore $\ell \leq 4\left|E^{\prime}\right| / \mathrm{b}$ as required.

Our algorithm for covering the edges of $E^{\prime}$ makes use of this block partitioning with $\mathrm{b}=16$. For any two nodes $u, v \in V(T)$, we use the notation $\pi(u, v, T)$ (or simply $\pi(u, v)$ ) to denote the unique simple path in the tree $T$ from $u$ to $v$. The algorithm begins with an empty collection $\mathcal{C}$ and then performs $\log n$ iterations where each iteration works as follows: Let $\widetilde{E} \subseteq E^{\prime}$ be the set of uncovered edges (initially $\widetilde{E}=E^{\prime}$ ). Then, we partition the nodes of $T$ with respect to $\widetilde{E}$ and
density parameter b. Finally, we search for cycles of length at most $\log n$ between the blocks. If such a cycle exists, we map it to a cycle in $G$ by connecting nodes $u, v$ within a block by the path $\pi(u, v)$ in the tree $T$. This way a cycle of length $\log n$ between the blocks translates to a cycle of length $D(T) \log n$ in the original graph $G$. Denote the resulting collection by $\mathcal{C}$.

We note that the cycles $\mathcal{C}$ might not be simple. This might happen if and only if the tree paths $\pi\left(v_{i}, u_{i+1}\right)$ and $\pi\left(v_{j}, u_{j+1}\right)$ intersect for some $j \in[t]$. Notice that the if an edge appears more than once in a cycle, then it must be a tree edge. Thus, we can transform any non-simple cycle $C$ into a collection of simple cycles that cover all edges that appeared only once in $C$ (the formal procedure is given at Figure 7). Since these cycle are constructed to cover only non-tree edges, we get that this transformation did not hurt the cover of $E^{\prime}$. The formal description of the algorithm is given in Figure 6.

## Algorithm NonTreeEdgeCover $\left(T, E^{\prime}\right)$

1. Initialize a cover $\mathcal{C}$ as an empty set.
2. Repeat $\log \left|E^{\prime}\right|$ times:
(a) Let $\widetilde{E} \subseteq E^{\prime}$ be the subset of all uncovered edges.
(b) Construct an ( $\widetilde{E}, \mathrm{~b}$ )-partitioning $\mathcal{B}$ of the nodes of $T$.
(c) While there are $t$ edges $\left(u_{1}, v_{1}\right), \ldots,\left(u_{t}, v_{t}\right) \in \widetilde{E}$ for $t \leq \log n$ such that for all $i \in[t-1]$, $v_{i}$ and $u_{i+1}$ are in the same block and $v_{t}$ and $u_{1}$ are in the same block (with respect to the partitioning $\mathcal{B})$ : Add the cycle $u_{1} \rightarrow v_{1} \rightarrow \pi\left(v_{1}, u_{2}\right) \rightarrow v_{2} \rightarrow \pi\left(v_{2}, u_{3}\right) \rightarrow v_{3} \rightarrow$ $\ldots \rightarrow v_{t} \rightarrow u_{1}$ to $\mathcal{C}$.
3. Compute $\mathcal{C}^{\prime} \leftarrow$ SimplifyCycles $(\mathcal{C})$ and output $\mathcal{C}^{\prime}$.

Figure 6: Procedure for covering non-tree edges.

## Algorithm SimplifyCycles(C)

1. While there is a cycle $C \in \mathcal{C}$ with a vertex $w \in C$ that appears more than once:
(a) Remove $C$ from $\mathcal{C}$.
(b) Let $C=v_{1} \rightarrow \cdots \rightarrow v_{k}$ and define $v_{k+i}=v_{i}$.
(c) Let $i_{1}, \ldots, i_{\ell}$ be such that $v_{i_{j}}=w$ for all $j \in[\ell]$, and let $i_{\ell+1}=i_{1}$.
(d) For all $j \in[\ell]$ let $C_{j}=v_{i_{j}} \rightarrow v_{i_{j}+1} \cdots \rightarrow v_{i_{j+1}}$, and if $\left|C_{j}\right| \geq 3$, add $C_{j}$ to $\mathcal{C}$.
2. Output $\mathcal{C}$.

Figure 7: Procedure making all cycles in $\mathcal{C}$ simple.
We move to the analysis of the algorithm, and show that it yields the desired cycle cover. That is, we show three things: that every cycle has length at most $O(D(T) \log n)$, that each edge is covered by at most $O(\log n)$ cycles, and that each edge has at least one cycle covering it.

Cycle Length. The bound of the cycle length follows directly from the construction. The cycles added to the collection are of the form $u_{1} \rightarrow v_{1} \rightarrow \pi\left(v_{1}, u_{2}\right) \rightarrow v_{2} \rightarrow \pi\left(v_{2}, u_{3}\right) \rightarrow v_{3} \rightarrow \ldots \rightarrow v_{t} \rightarrow$ $u_{1}$, where each $\pi\left(v_{i}, u_{i+1}\right)$ are paths in the tree $T$ and thus are of length at most $2 D(T)$. Notice that the simplification process of the cycles can only make the cycles shorter. Since $t \leq \log n$ we get that the cycle lengths are bounded by $O(D(T) \log n)$.
Congestion. To bound the congestion of the cycle cover we exploit the structure of the partitioning, and the fact that each block in the partition has a low density. We begin by showing that by the post-order numbering, all nodes in a given subtree have a continuous range of numbers. For every $z \in V(T)$, let $\min _{N}(z)$ be the minimal number of a node in the subtree of $T$ rooted by $z$. That is, $\min _{N}(z)=\min _{u \in T_{z}} N(u)$ and similarly let $\max _{N}(z)=\max _{u \in T_{z}} N(u)$.

Claim 2. For every $z \in V(T)$ and for every $u \in G$ it holds that (1) $\max _{N}(z)=N(z)$ and (2) $N(u) \in\left[\min _{N}(z), \max _{N}(z)\right]$ iff $u \in T_{z}$.

Proof. The proof is by induction on the depth of $T_{z}$. For the base case, we consider the leaf nodes $z$, and hence $T_{z}$ with 0-depth, the claim holds vacuously. Assume that the claim holds for nodes in level $i+1$ and consider now a node $z$ in level $i$. Let $v_{i, 1}, \ldots, v_{i, \ell}$ be the children of $z$ ordered from left to right. By the post-order traversal, the root $v_{i, j}$ is the last vertex visited in $T_{v_{i, j}}$ and hence $N\left(v_{i, j}\right)=\max _{N}\left(v_{i, j}\right)$. Since the traversal of $T_{v_{i, j}}$ starts right after finishing the traversal of $T_{v_{i, j-1}}$ for every $j \geq 2$, it holds that $\min _{N}\left(v_{i, j}\right)=N\left(v_{i, j-1}\right)+1$. Using the induction assumption for $v_{i, j}$, we get that all the nodes in $T_{z} \backslash\{z\}$ have numbering in the range $\left[\min _{N}\left(v_{i, 1}\right), \max _{N}\left(v_{i, \ell}\right]\right.$ and any other node not in $T_{z}$ is not in this range. Finally, $N(z)=N\left(v_{i, \ell}\right)+1$ and so the claim holds.

The cycles in the cover we compute are composed of paths $\pi(u, v)$ for two nodes $u$ and $v$ in the same block. Thus, to bound the congestion on an edge $e \in T$ we need to bound the number of blocks such that there contains $u, v$ in that block such that $\pi(u, v)$ passes through $e$. The next claim shows that every edge in the tree is effected by at most 2 blocks.

Claim 3. Let $e \in T$ be a tree edge and define $\mathcal{B}(e)=\{B \in \mathcal{B} \mid \exists u, v \in B$ s.t. $e \in \pi(u, v)\}$. Then, $|\mathcal{B}(e)| \leq 2$ for every $e \in T$.

Proof. Let $e=(w, z)$ where $w$ is closer to the root in $T$, and let $u, v$ be two nodes in the same block $B$ such that $e \in \pi(u, v)$. Let $\ell$ be the LCA of $u$ and $v$ in $T$ (it might be that $\ell \in\{u, v\}$ ), then the tree path between $u$ and $v$ can be written as $\pi(u, v)=\pi(u, \ell) \circ \pi(\ell, v)$. Without loss of generality, assume that $e \in \pi(\ell, v)$. This implies that $v \in T_{z}$ but $u \notin T_{z}$. Hence, the block of $u$ and $v$ intersects the nodes of $T_{z}$. Each block consists of a consecutive set of nodes, and by Claim 2 also $T_{z}$ consists of a consecutive set of nodes with numbering in the range $\left[\min _{N}(z), \max _{N}(z)\right.$ ], thus there are at most two such blocks that intersect $e=(w, z)$, i.e., blocks $B$ that contains both a vertex $y$ with $N(y) \in\left[\min _{N}(z), \max _{N}(z)\right]$ and a vertex $y^{\prime}$ with $N\left(y^{\prime}\right) \notin\left[\min _{N}(z), \max _{N}(z)\right]$, and the claim follows.

Finally, we use the above claims to bound the congestion. Consider any tree edge $e=(w, z)$ where $w$ is closer to the root than $z$. Let $T_{z}$ be the subtree of $T$ rooted at $z$. Fix an iteration $i$ of the algorithm. We characterize all cycles in $\mathcal{C}$ that go through this edge.

For any cycle that passes through $e$ there must be a block $B$ and two nodes $u, v \in B$ such that $e \in \pi(u, v)$. By Claim 3, we know that there are that at each iteration of the algorithm, there are at most two such blocks $B$ that can affect the congestion of $e$. Moreover, we claim that each such
block has density at most b . Otherwise it would be a block containing a single node, say $u$, and thus the path $\pi(u, u)=u$ is empty and cannot contain the edge $e$. For each edge in $E^{\prime}$ we construct a single cycle in $\mathcal{C}$, and thus for each one of the two blocks that affect $e$ the number of pairs $u, v$ such that $e \in \pi(u, v)$ is bounded by $\mathrm{b} / 2$ (each pair $u, v$ has two edges in the block $B$ and we know that the total number of edges is bounded by b).

To summarize the above, we get that for each iteration, that are at most 2 blocks that can contribute to the congestion of an edge $e$ : one block that intersects $T_{z}$ but has also nodes smaller than $\min _{N}(z)$ and one block that intersects $T_{z}$ but has also nodes larger than $\max _{N}(z)$. Each of these two blocks can increase the congestion of $e$ by at most $\mathrm{b} / 2$. Since there are at most $\log n$ iterations, we can bound the total congestion by $b \log n$. Notice that if an edge appears $k$ times in a cycle, then this congestion bound counts all $k$ appearances. Thus, after the simplification of the cycles, the congestion remains unchanged.
Cover. We show that each edge in $E^{\prime}$ was covered by some cycle. We begin by showing cover for the cycles before the simplification procedure, and then show that the cover remains after this procedure. The idea is that at each iteration of the algorithm, the number of uncovered edges is reduced by half. Therefore, the $\log n$ iterations should suffice for covering all $2 n$ edges of $E^{\prime}$. In each iteration we partition the nodes into blocks, and we search for cycles between the blocks. The point is that if the number of edges is large, then we considering the blocks as nodes in a new virtual graph, this graph has a large number of edges and thus must have a short cycle. At each iteration, the blocks become larger which make the virtual have less nodes and the number of edges is larger relative to the small number of nodes.

In what follows, we formalize the intuition given above. Let $\widetilde{E}_{i}$ be the set $\widetilde{E}$ at the $i^{\text {th }}$ iteration of the algorithm. Consider the iteration $i$ with the set of uncovered edge set $\widetilde{E}_{i}$. Our goal is to show that $\widetilde{E}_{i+1} \leq 1 / 2 \widetilde{E}_{i}$. By having $\log \left|E^{\prime}\right|$ iterations, last set will be empty.

Let $\mathcal{B}_{i}$ be the partitioning performed at iteration $i$ with respect to the edge set $\widetilde{E}_{i}$. Define a super-graph $\widetilde{G}$ in which each block $B_{j} \in \mathcal{B}_{i}$ is represented by a node $\widetilde{v}_{j}$, and there is an edge $\left(\widetilde{v}_{j}, \widetilde{v}_{j^{\prime}}\right)$ in $\widetilde{G}$ if there is an edge in $\widetilde{E}_{i}$ between some node $u$ in $B_{j}$ and a node $u^{\prime}$ in $B_{j^{\prime}}$, i.e.,

$$
\left(\widetilde{v}_{j}, \widetilde{v}_{j^{\prime}}\right) \in E(\widetilde{G}) \Longleftrightarrow \widetilde{E}_{i} \cap\left(B_{j} \times B_{j^{\prime}}\right) \neq \emptyset
$$

See Figure 8 for an illustration. The number of nodes in $\widetilde{G}$, which we denote by $n_{i}$, is the number of blocks in the partition and is bounded by $n_{i} \leq 4\left|\widetilde{E}_{i}\right| / \mathrm{b}$. Let $B(u)$ be the block of the node $u$. The algorithm finds cycles of the form $u_{1} \rightarrow v_{1} \rightarrow \pi\left(v_{1}, u_{2}\right) \rightarrow v_{2} \rightarrow \pi\left(v_{2}, u_{3}\right) \rightarrow v_{3} \rightarrow \ldots \rightarrow v_{t} \rightarrow u_{1}$, which is equivalent to finding the cycle $B\left(u_{1}\right), \ldots, B\left(u_{t}\right)$ in the graph $\widetilde{G}$. In general, any cycle of length $t$ in $\widetilde{G}$ is mapped to a cycle in $G$ of length at most $t \cdot D(T)$. Then, the algorithm adds the cycle to $\mathcal{C}$ and removes the edges of the cycle (thus removing them also from $\widetilde{G}$ ). At the end of iteration $i$ the graph $\widetilde{G}$ has no cycles of length at most $\log n$. At this point, the next set of edges $\widetilde{E}_{i+1}$ is exactly the edges left in $\widetilde{G}$. By Fact 1 (and recalling that $\mathrm{b}=16$ ) we get that if $\widetilde{G}$ does not have any cycles of length at most $\log n$ then we get the following bound on the number of edges:

$$
\widetilde{E}_{i+1} \leq 2 n_{i}=8\left|\widetilde{E}_{i}\right| / \mathrm{b}=\left|\widetilde{E}_{i}\right| / 2
$$

Thus, all will be covered by a cycle $C$ before the simplification process. We show that the simplification procedure of the cycle maintains the cover requirement. This stems from the fact that any edge that appears more than once in a cycle, might be dropped but is not the edge for which this cycle was constructed to cover. That is, for each non-tree edge, we construct a cycle


Figure 8: Left: Schematic illustration of the block partitioning in the tree $T$. Dashed edges are those that remain to be covered after employing Alg. LocalCover, where each two blocks are connected by exactly one edge. Each dashed edge corresponds to super-edges in $\widetilde{G}$. Right: A triangle in the super-graph $\widetilde{G}$.
that covers it. By the construction of the cycles, if a cycle is not simple it must be because of a tree edge. It is left to show that this process only drops edges that appear more than once:
Claim 4. Let $C$ be a cycle and let $\mathcal{C}^{\prime} \leftarrow \operatorname{SimplifyCycles}(C)$. Then, for every edge $e \in C$ that appears at most once in $C$ there is a cycle $C^{\prime} \in \mathcal{C}^{\prime}$ such that $e \in C^{\prime}$.

Proof. The procedure SimplifyCycles works in iterations where in each iteration it chooses a vertex $w$ that appears more than once in $C$ and partitions the cycle $C$ to consecutive parts, $C_{1}, \ldots, C_{\ell}$. All edges in $C$ appear in some $C_{j}$. However, $C_{j}$ might not be a proper cycle since it might be the case that $\left|C_{j}\right| \leq 2$. Thus, we show that in an edge $e \in C$ appeared at most once in a cycle $C$ then it will appear in $C_{j}$ for some $j$ where $\left|C_{j}\right| \geq 3$. We show that this holds for any iteration and thus will hold at the end of the process.

We assume without loss of generality that no vertex has two consecutive appearances. Denote $e=\left(v_{2}, v_{3}\right)$ and let $C=v_{1} \rightarrow v_{2} \rightarrow v_{3} \rightarrow \ldots v_{k}$ for $k \geq 3$. Since $e$ does not appear again in $C$ we know that $v_{1}, v_{2}, v_{3}$ are distinct. Thus, if $k=3$ then $C$ will not be split again and the claim follows.

Therefore, assume that $k \geq 4$. Since $e$ does not appear again in $C$ we know that $v_{2}, v_{3}, v_{4}$ are distinct (it might be the case that $v_{1}=v_{4}$ ). Thus, we know that $\left|\left\{v_{1}, v_{2}, v_{3}, v_{4}\right\}\right| \geq 3$. Any subsequence begins and ends at the same vertex and thus the subsequence $C_{j}$ that contains $e$ must contains all of $v_{1}, v_{2}, v_{3}, v_{3}$ and thus $\left|C_{j}\right| \geq 3$, and the claim follows.

### 4.1.2 Covering Tree Edges

Finally, we present Algorithm TreeCover that computes a cycle cover for the tree edges. The algorithm is recursive and uses Algorithm NonTreeCover as a black-box. Formally, we show:
Lemma 3. For every n-vertex graph $G$ and a tree $T \subseteq G$ of depth $D$, there exists a $\left(D \log n, \log ^{3} n\right)$ cycle cover $\mathcal{C}_{3}$ for the edges of $T$.

We begin with some notation. Throughout, when referring to a tree edge $(u, v) \in T$, the node $u$ is closer to the root of $T$ than $v$. Let $E(T)=\left\{e_{1}, \ldots, e_{n-1}\right\}$ be an ordering of the edges of $T$ in non-decreasing distance from the root. For every tree edge $e \in T$, define the swap edge of $e$ by $e^{\prime}=\operatorname{Swap}(e)$ to be an arbitrary edge in $G$ that restores the connectivity of $T \backslash\{e\}$. Since the graph $G$ is 2-edge connected such an edge $\operatorname{Swap}(e)$ is guaranteed to exist for every $e \in T$. Let $e=(u, v)$ and $\left(u^{\prime}, v^{\prime}\right)=\operatorname{Swap}(e)$, we denote $u$ by $p(v)$ (since $u$ is the parent of $v$ in the tree) and $v^{\prime}$ by $s(v)$, where $s(v)$ is the endpoint of $\operatorname{Swap}(e)$ that do not belong to $T\left(u^{\prime}\right)$ (i.e., the subtree $T$ rooted at $u$ ). Define the $v-s(v)$ path

$$
P_{e}=\pi\left(v, u^{\prime}\right) \circ \operatorname{Swap}(e) .
$$

For an illustration see Figure 9.
For the tree $T$, we construct a subset of tree edges denoted by $I(T)$ that we are able to cover. These edges are independent in the sense that their $P_{e}$ paths are "almost" edge disjoint (as will be shown next). The subset $I(T)$ is constructed by going through the edges of $T$ in non-decreasing distance from the root. At any point, we add $e$ to $I(T)$ only if it is not covered by the $P_{e^{\prime}}$ paths of the $e^{\prime}$ edges already added.

Claim 5. The subset $I(T)$ satisfies the following properties:

- For every $e \in E(T)$, there exists $e^{\prime} \in I(T)$ such that $e \in e^{\prime} \circ P_{e^{\prime}}$.
- For every $e, e^{\prime} \in I(T)$ such that $e \neq e^{\prime}$ it holds that $P_{e}$ and $P_{e^{\prime}}$ have no tree edge in common (no edge of $T$ is in both paths).
- For every swap edge $(z, w)$, there exists at most two paths $P_{e}, P_{e^{\prime}}$ for $e, e^{\prime} \in I(T)$ such that one passes through $(z, w)$ and the other through $(w, z)$. That is, each swap edge appears at most twice on the $P_{e}$ paths, once in each direction.

Proof. The first property follows directly from the construction. Next, we show that they share no tree edge in common. Assume that there is a common edge $(z, w) \in P_{e} \cap P_{e}^{\prime} \cap E(T)$. Then, both $e, e^{\prime}$ must be on the path from root to $z$ on the tree and hence $e^{\prime} \in P_{e}$, leading to contradiction. For the third property, assume towards constriction that both $P_{e}$ and $P_{e^{\prime}}$ use the same swap edge in the same direction. Again it implies that both $e, e^{\prime}$ are on the path from the root to $z$ on $T$.

Our cycle cover for the $I(T)$ edges will be shown to cover all the edges of the tree $T$. This is because the cycle that we construct to cover an edge $e \in I(T)$ necessarily contains $P_{e}$.

Algorithm TreeCover uses the following procedure TreeEdgeDisjointPath, usually used in the context of distributed routing.
Key Tool: Route Disjoint Matching. Algorithm TreeEdgeDisjointPath solves the following problem defined by Peleg (see Lemma 4.3.2 [Pel00]): given a rooted tree $T$ and a set of $2 k$ marked nodes $M \subseteq V(T)$ for $k \leq n / 2$, the goal is to find (by a distributed algorithm) a matching of these vertices $\left\langle w_{i}, w_{j}\right\rangle$ into pairs such that the tree paths $\pi\left(w_{i}, w_{j}, T\right)$ connecting the matched pairs are edge-disjoint. This matching can be computed distributively in $O(\operatorname{Diam}(T))$ rounds by working from the leaf nodes towards the root. In each round a node $u$ that received information on more $\ell \geq 2$ unmarked nodes in its subtree, match all but at most one into pairs and upcast to its parent the ID of at most one unmarked node in its subtree. It is easy to see that all tree paths between matched nodes are indeed edge disjoint.

We are now ready to explain the cycle cover construction of the tree edges $E(T)$.

Description of Algorithm TreeCover. We restrict attention for covering the edges of $I(T)$. The tree edges $I(T)$ will be covered in a specific manner that covers also the edges of $E(T) \backslash I(T)$. The key idea is to define a collection of (virtual) non-tree edges $\widetilde{E}=\{(v, s(v)):(p(v), v) \in I(T)\}$ and covering these non-tree edges by enforcing the cycle that covers the non-tree edge $(v, s(v))$ to covers the edges $e=(p(v), v)$ as well as the path $P_{e}$. Since every edge $e^{\prime} \in T$ appears on one of the $e \circ P_{e}$ paths, this will guarantee that all tree edges are covered.

Algorithm TreeCover is recursive and has $O(\log n)$ levels of recursion. In each independent level of the recursion we need to solve the following sub-problem: Given a tree $T^{\prime}$, cover by cycles the edges of $I\left(T^{\prime}\right)$ along with their $P_{e}$ paths. The key idea is to subdivide this problem into two independent and balanced subproblems. To do that, the tree $T^{\prime}$ gets partitioned ${ }^{7}$ into two balanced edge disjoint subtrees $T_{1}^{\prime}$ and $T_{2}^{\prime}$, where $\left|T_{1}^{\prime}\right|,\left|T_{2}^{\prime}\right| \leq 2 / 3 \cdot\left|T^{\prime}\right|$ and $E\left(T_{1}^{\prime}\right) \cup E\left(T_{2}^{\prime}\right)=E\left(T^{\prime}\right)$. Some of the tree edges in $T^{\prime}$ are covered by applying a procedure that computes cycles using the edges of $T^{\prime}$, and the remaining ones will be covered recursively in either $T_{1}^{\prime}$ or $T_{2}^{\prime}$. Specifically, the edges of $I\left(T^{\prime}\right)$ are partitioned into 4 types depending on the position of their swap edges. Let

$$
E_{x, y}^{\prime}=\left\{(u, v) \in E\left(T_{x}^{\prime}\right) \cap I\left(T^{\prime}\right) \mid v \in V\left(T_{x}^{\prime}\right) \text { and } s(v) \in V\left(T_{y}^{\prime}\right) \backslash V\left(T_{x}^{\prime}\right)\right\}, \text { for every } x, y \in\{1,2\} .
$$

The algorithm computes a cycle cover $\mathcal{C}_{1,2}$ (resp., $\mathcal{C}_{2,1}$ ) for covering the edges of $E_{1,2}^{\prime}, E_{2,1}^{\prime}$ respectively. The remaining edges $E_{11}^{\prime}$ and $E_{22}^{\prime}$ are covered recursively by applying the algorithm on $T_{1}^{\prime}$ and $T_{2}^{\prime}$ respectively. See Fig. 9 for an illustration.

We now describe how to compute the cycle cover $\mathcal{C}_{1,2}$ for the edges of $E_{1,2}^{\prime}$. The edges $E_{2,1}^{\prime}$ are covered analogously (i.e., by switching the roles of $T_{1}^{\prime}$ and $T_{2}^{\prime}$ ). Recall that the tree edges $E_{1,2}^{\prime}$ are those edges $(p(v), v)$ such $v \in T_{1}^{\prime}$ and $s(v) \in T_{2}^{\prime}$. The procedure works in $O(\log n)$ phases, each phase $i$ computes three cycle collections $\mathcal{C}_{i, 1}^{\prime}, \mathcal{C}_{i, 2}^{\prime}$ and $\mathcal{C}_{i, 3}^{\prime}$ which together covers at least half of the yet uncovered edges of $E_{1,2}^{\prime}$ (as will be shown in analysis).

Consider the $i^{\text {th }}$ phase where we are given the set of yet uncovered edges $X_{i} \subseteq E_{1,2}^{\prime}$. We first mark all the vertices $v$ with $(p(v), v) \in X_{i}$. Let $M_{i}$ be this set of marked nodes. For ease of description, assume that $M_{i}$ is even, otherwise, we omit one of the marked vertices $w$ (from $M_{i}$ ) and take care of its edge $(p(w), w)$ in later phases. We apply Algorithm TreeEdgeDisjointPath $\left(T_{1}^{\prime}, M_{i}\right)$ (see Lemma 4.3.2 [Pel00]) which matches the marked vertices $M_{i}$ into pairs $\Sigma=\left\{\left\langle v_{1}, v_{2}\right\rangle \mid v_{1}, v_{2} \in\right.$ $\left.M_{i}\right\}$ such that for each pair $\sigma=\left\langle v_{1}, v_{2}\right\rangle$ there is a tree path $\pi(\sigma)=\pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right)$ and all the tree paths $\pi(\sigma), \pi\left(\sigma^{\prime}\right)$ are edge disjoint for every $\sigma, \sigma^{\prime} \in \Sigma$.

Let $X_{i}^{\prime \prime}=\left\{e=(p(v), v) \in X_{i}: \exists v^{\prime}\right.$ and $\left\langle v, v^{\prime}\right\rangle \in \Sigma$, s.t. $\left.e \in \pi\left(v, v^{\prime}, T_{1}^{\prime}\right)\right\}$ be the set of edges in $X_{i}$ that appear on the collection of edge disjoint paths $\{\pi(\sigma), \sigma \in \Sigma\}$. Our goal is to cover all edges in $E_{i}^{\prime \prime}=X_{i}^{\prime \prime} \cup\left\{P_{e} \mid e \in X_{i}^{\prime \prime}\right\}$ by cycles $\mathcal{C}_{i}$. To make sure that all edges $E_{i}^{\prime \prime}$ are covered, we have to be careful that each such edge appears on a given cycle exactly once. Towards this end, we define a directed conflict graph $G_{\Sigma}$ whose vertex set are the pairs of $\Sigma$, and there is an $\operatorname{arc}\left(\sigma^{\prime}, \sigma\right) \in A\left(G_{\Sigma}\right)$ where $\sigma=\left\langle v_{1}, v_{2}\right\rangle, \sigma=\left\langle v_{1}^{\prime}, v_{2}^{\prime}\right\rangle$, if at least one of the following cases holds: Case (I) $e=\left(p\left(v_{1}\right), v_{1}\right)$ on $\pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right)$ and the path $\pi^{\prime}=\pi\left(v_{1}^{\prime}, v_{2}^{\prime}, T_{1}^{\prime}\right)$ intersects the edges of $P_{e}$; Case (II) $e^{\prime}=\left(p\left(v_{2}\right), v_{2}\right)$ on $\pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right)$ and the path $\pi^{\prime}$ intersects the edges of $P_{e^{\prime}}$. Intuitively, a cycle that contains both $\pi^{\prime}$ and $P_{e}$ is not simple and in particular might not cover all edges on $P_{e}$. Since the goal of the pair $\sigma=\left\langle v_{1}, v_{2}\right\rangle$ is to cover all edges on $P_{e}$ (for $e \in \pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right)$ ), the pair $\sigma^{\prime}$ "interferes" with $\sigma$.

In the analysis section (Claim 6), we show that the outdegree in the graph $G_{\Sigma}$ is bounded by 1 and hence we can color $G_{\Sigma}$ with 3 colors. This allows us to partition $\Sigma$ into three color classes

[^7]$\Sigma_{1}, \Sigma_{2}$ and $\Sigma_{3}$. Each color class $\Sigma_{j}$ is an independent set in $G_{\Sigma}$ and thus it is "safe" to cover all these pairs by cycles together. We then compute a cycle cover $\mathcal{C}_{i, j}$, for each $j \in\{1,2,3\}$. The collection of all these cycles will be shown to cover the edges $E_{i}^{\prime \prime}$.

To compute $\mathcal{C}_{i, j}$ for $j=\{1,2,3\}$, for each matched pair $\left\langle v_{1}, v_{2}\right\rangle \in \Sigma_{j}$, we add to $T_{2}^{\prime}$ a virtual edge $\widehat{e}$ between $s\left(v_{1}\right)$ and $s\left(v_{2}\right)$. Let

$$
\widehat{E}_{i, j}=\left\{\left(s\left(v_{1}\right), s\left(v_{2}\right)\right) \mid\left\langle v_{1}, v_{2}\right\rangle \in \Sigma_{j}\right\} .
$$

We cover these virtual non-tree edges by cycles using Algorithm NonTreeCover on the tree $T_{2}^{\prime}$ with the non-tree edges $\widehat{E}_{i, j}$. Let $\mathcal{C}_{i, j}^{\prime \prime}$ be the output $O(D \log n, \log n)$ cycle cover of Algorithm NonTreeCover $\left(T_{2}^{\prime}, E_{i, j}^{\prime}\right)$. The output cycles of $\mathcal{C}_{i, j}^{\prime \prime}$ are not yet cycles in $G$ as they consists of two types of virtual edges: the edges in $\widehat{E}_{i, j}$ and the edges $\widetilde{E}=\{(v, s(v)) \mid(p(v), v) \in I(T)\}$. First, we translate each cycle $C^{\prime \prime} \in \mathcal{C}_{i, j}^{\prime \prime}$ into a cycle $C^{\prime}$ in $G \cup \widetilde{E}$ by replacing each of the virtual edges $\widehat{e}=\left(s\left(v_{1}\right), s\left(v_{2}\right)\right) \in \widehat{E}_{i, j}$ in $C^{\prime \prime}$ with the path $P(\widehat{e})=\left(s\left(v_{1}\right), v_{1}\right) \circ \pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right) \circ\left(v_{2}, s\left(v_{2}\right)\right)$. Then, we replace each virtual edge $(v, s(v)) \in \widetilde{E}$ in $C^{\prime}$ by the $v-s(v)$ path $P_{e}$ for $e=(p(v), v)$. This results in cycles $\mathcal{C}_{i, j}$ in $G$.

Finally, let $\mathcal{C}_{i}=\mathcal{C}_{i, 1} \cup \mathcal{C}_{i, 2} \cup \mathcal{C}_{i, 3}$ and define $X_{i+1}=X_{i} \backslash X_{i}^{\prime \prime}$ to be the set of edges $e \in X_{i}$ that are not covered by the paths of $\Sigma$. If in the last phase $\ell=O(\log n)$, the set of marked nodes $M_{\ell}$ is odd, we omit one of the marked nodes $w \in M_{\ell}$, and cover its tree edge $e=(p(w), w)$ by taking the fundamental cycle of the swap edge $\operatorname{Swap}(e)$ into the cycle collection. The final cycle collection for $E_{1,2}^{\prime}$ is given by $\mathcal{C}_{1,2}=\bigcup_{i=1}^{\ell} \mathcal{C}_{i}$. The same is done for the edges $E_{2,1}^{\prime}$. This completes the description of the algorithm. The final collection of cycles is denoted by $\mathcal{C}_{3}$. See Figure 12 for the full description of the algorithm. See Figures 9 to 11 and for illustration.

We analyze the TreeCover algorithm and show that it finds short cycles, with low congestion and that every edge of $T$ is covered.
Short Cycles. By construction, each cycle that we compute using Algorithm NonTreeCover consists of at most $O(\log n)$ non-tree edges $\widehat{E}_{i, j}$. The algorithm replaces each non-tree edge $\widehat{e}=\left(v_{1}, v_{2}\right)$ by an $v_{1}-v_{2}$ path in $G$ of length $O(D)$. This is done in two steps. First, $\widehat{e}=\left(v_{1}, v_{2}\right)$ is replaced by a path $P_{\widehat{e}}=\left(v_{1}, s\left(v_{1}\right)\right) \circ \pi\left(v_{1}, v_{2}, T\right) \circ\left(v_{2}, s\left(v_{2}\right)\right)$ in $G \cup \widetilde{E}$. Then, each $(v, s(v))$ edge is replaced by the path $P_{(p(v), v)}$ in $G$, which is also of length $O(D)$. Hence, overall the translated path $v_{1}-v_{2}$ path in $G$ has length $O(D)$. Since there are $O(\log n)$ virtual edges that are replaced on a given cycle, the cycles of $G$ has length $O(D \log n)$.
Cover. We start with some auxiliary property used in our algorithm.
Claim 6. Consider the graph $G_{\Sigma}$ constructed when considering the edges in $E_{1,2}^{\prime}$. The outdegree of each pair $\sigma^{\prime}=\left\langle v_{1}^{\prime}, v_{2}^{\prime}\right\rangle \in \Sigma$ in $G_{\Sigma}$ is at most 1 . Therefore, $G_{\Sigma}$ can be colored in 3 colors.

Proof. Let $\sigma=\left\langle v_{1}, v_{2}\right\rangle$ be such that $\sigma^{\prime}$ interferes with $\sigma$ (i.e., $\left.\left(\sigma^{\prime}, \sigma\right) \in A\left(G_{\Sigma}\right)\right)$. Without loss of generality, let $e=\left(p\left(v_{1}\right), v_{1}\right)$ be such that $e \in \pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right)$ and $\pi\left(v_{1}^{\prime}, v_{2}^{\prime}, T_{1}^{\prime}\right)$ intersects the edges of $P_{e}$.

We first claim that this implies that $e$ appears above the LCA of $v_{1}^{\prime}$ and $v_{2}^{\prime}$ in $T_{1}^{\prime}$, and hence by the properties of our partitioning, also in $T$. Assume towards contradiction otherwise, since $e \circ P_{e}$ is a path on $T$ (where $e$ is closer to the root) and since $P_{e}$ intersects $\pi\left(v_{1}^{\prime}, v_{2}^{\prime}, T_{1}^{\prime}\right)$, it implies that $e \in \pi\left(v_{1}^{\prime}, v_{2}^{\prime}, T_{1}^{\prime}\right)$. Since the vertex $v_{1}$ is marked, we get a contradiction that $v_{1}^{\prime}$ got matched with $v_{2}^{\prime}$ as the algorithm would have matched $v_{1}$ with one of them. In particular, we would get that the


Figure 9: Left: Illustration the swap edge $e^{\prime}=\operatorname{Swap}(e)$ and the path $P_{e}$ for an edge $e \in T$. For each tree edge $e=(u, v) \in T$, we add the auxiliary edge $(v, s(v))$. Right: The tree $T^{\prime}$ is partitioned into two balanced trees $T_{1}^{\prime}$ and $T_{2}^{\prime}$. The root vertex in this example belongs to both trees. The edges $\widetilde{E}$ are partitioned into four sets: $E_{1,1}^{\prime}$ (e.g., the edge $\left(p\left(v_{3}\right), v_{3}\right)$ ), $E_{2,2}^{\prime}$ (e.g., the edge $\left.\left(p\left(v_{4}\right), v_{4}\right)\right), E_{1,2}^{\prime}$ (e.g., the edge $\left.\left(p\left(v_{1}\right), v_{1}\right)\right), E_{2,1}^{\prime}$ (e.g., the edge $\left.\left(p\left(v_{5}\right), v_{5}\right)\right)$. The algorithm covers the edges of $E_{1,2}^{\prime}$ by using Algorithm TreeEdgeDisjointPath to compute a matching and edge disjoint paths in $T_{1}^{\prime}$. See the tree paths between $v_{1}$ and $v_{2}$ and $v_{1}^{\prime}$ and $v_{2}^{\prime}$. Based on this matching, we add virtual edges between vertices of $T_{2}^{\prime}$, for example the edges $\left(s\left(v_{1}\right), s\left(v_{2}\right)\right)$ and $\left(s\left(v_{1}^{\prime}\right), s\left(v_{2}^{\prime}\right)\right)$ shown in dashed. The algorithm then applies Algorithm NonTreeCover to cover these non-tree edges in $T_{2}^{\prime}$.
paths $\pi(\sigma)$ and $\pi\left(\sigma^{\prime}\right)$ are not edge disjoint, as both contain $e$. Hence, we prove that $e$ is above the LCA of $v_{1}^{\prime}$ and $v_{2}^{\prime}$.

Next, assume towards contradiction that there is another pair $\sigma^{\prime \prime}=\left\langle v_{1}^{\prime \prime}, v_{2}^{\prime \prime}\right\rangle \in \Sigma$ such that $\sigma^{\prime}$ interferes with $\sigma^{\prime \prime}$. Without loss of generality, let $v_{1}^{\prime \prime}$ be such that $e^{\prime \prime}=\left(p\left(v_{1}^{\prime \prime}\right), v_{1}^{\prime \prime}\right)$ in on $\pi\left(v_{1}^{\prime \prime}, v_{2}^{\prime \prime}, T_{1}^{\prime}\right)$ and $P_{e^{\prime \prime}}$ intersects with $\pi\left(v_{1}^{\prime}, v_{2}^{\prime}, T_{1}^{\prime}\right)$. This implies that $e^{\prime \prime}$ is also above the LCA of $v_{1}^{\prime}$ and $v_{2}^{\prime}$ in $T_{1}^{\prime}$. Since one of the edges of $P_{e^{\prime \prime}}$ is on $\pi\left(v_{1}^{\prime}, v_{2}^{\prime}, T_{1}^{\prime}\right)$ it must be that either $e^{\prime \prime}$ on $P_{e}$ or vice verca, in contradiction that $e, e^{\prime \prime} \in I(T)$.

We now claim that each edge $e \in T$ is covered. By the definition of $I(T) \subseteq E(T)$, it is sufficient to show that:

Claim 7. For every edge $e \in I(T)$, there exists a cycle $C \in \mathcal{C}_{3}$ such that $e \circ P_{e} \subseteq C$.
We now consider a specific tree edge $e=(p(v), v)$. First, note that since $(v, s(v))$ is a non-tree edge, there must be some recursive call with the tree $T^{\prime}$ such that $v \in T_{1}^{\prime}$ and $s(v) \in T_{2}^{\prime}$ where $T_{1}^{\prime}$ and $T_{2}^{\prime}$ are the balanced partitioning of $T^{\prime}$. At that point, $(v, s(v))$ is an edge in $E_{1,2}^{\prime}$. We show that in the $\ell=O(\log n)$ phases of the algorithm for covering the $E_{1,2}^{\prime}$, there is a phase in which $e=(p(v), v)$ is covered.


Figure 10: Illustration of replacing a single virtual edge $\left(s\left(v_{1}\right), s\left(v_{2}\right)\right)$ in a cycle $C^{\prime \prime} \in \mathcal{C}_{i, j}^{\prime \prime}$ by an $s\left(v_{1}\right)-s\left(v_{2}\right)$ path in $G$.


Figure 11: Translating virtual cycles into cycles in $G$. Each cycle contains $O(\log n)$ virtual edges which are are replaced by (almost) edge disjoint paths in $G$. Note that each edge on $e_{i} \circ P_{e_{i}}$ appears exactly once since the $P_{e}$ paths are tree-edge disjoint and the pairs $\sigma=\left\langle u_{j}, v_{j}\right\rangle$ in $\Sigma_{i, j}$ do not interfere with each other.

Claim 8. (I) For every $e=(p(v), v) \in E_{1,2}^{\prime}$ except at most one edge $e^{*}$, there is a phase $i_{e}$ where Algorithm TreeEdgeDisjointPath matched $v$ with some $v^{\prime}$ such that $e \in \pi\left(v, v^{\prime}\right)$.

## Algorithm TreeCover $\left(T^{\prime}\right)$

1. If $\left|T^{\prime}\right|=1$ then output empty collection.
2. Let $\mathcal{C}$ be an empty collection.
3. Partition $T^{\prime}$ into balanced $T_{1}^{\prime} \cup T_{2}^{\prime}$.
4. Let $E^{\prime}$ be an empty set.
5. For every $(u, v) \in T$ let $\left(u^{\prime}, v^{\prime}\right)=\operatorname{Swap}((u, v))$ and add a virtual edge $\left(v, v^{\prime}\right)$ to $E^{\prime}$.

6 . For $i=1, \ldots, O(\log n)$ :
(a) Let $M_{i}$ be all active nodes $v \in V\left(T_{1}^{\prime}\right)$ s.t. $\left.\operatorname{Swap}(v) \in V\left(T_{2}\right)\right\}$.
(b) Apply TreeEdgeDisjointPath $\left(T_{1}^{\prime}, M_{i}\right)$ and let $\Sigma=\left\{\left\langle v_{1}, v_{2}\right\rangle\right\}$ be the collection of matched pairs.
(c) Partition $\Sigma$ into 3 independent sets $\Sigma_{1}, \Sigma_{2}$ and $\Sigma_{3}$.
(d) For every $j \in\{1,2,3\}$ compute a cycle cover $\mathcal{C}_{i, j}$ as follows:
i. For every pair $\left\langle v_{1}, v_{2}\right\rangle$ in $\Sigma_{j}$ add a virtual edge $\left(s\left(v_{1}\right), s\left(v_{2}\right)\right)$ to $\widehat{E}_{i, j}$.
ii. $\mathcal{C}_{i, j}^{\prime \prime} \leftarrow \mathcal{C}_{i, j}^{\prime \prime} \cup \operatorname{NonTreeCover}\left(T_{2}, \widehat{E}_{i, j}\right)$.
iii. Translate $\mathcal{C}_{i, j}^{\prime \prime}$ to cycles $\mathcal{C}_{i, j}$ in $G$.
(e) Let $\mathcal{C}_{i}=\mathcal{C}_{i, 1} \cup \mathcal{C}_{i, 2} \cup \mathcal{C}_{i, 3}$.
7. $\mathcal{C}_{1}=\bigcup_{i} \mathcal{C}_{i}$.
8. Repeat where $T_{1}^{\prime}$ and $T_{2}^{\prime}$ are switched.
9. Add to $\mathcal{C}$ the output of $\operatorname{Tree} \operatorname{Cover}\left(T_{1}^{\prime}\right) \cup \operatorname{Tree} \operatorname{Cover}\left(T_{2}^{\prime}\right)$.
10. Output SimplifyCycles $\left(\mathcal{C} \cup \mathcal{C}_{1}\right)$.

Figure 12: Procedure for covering tree edges.
(II) Each edge $e \neq e^{*}$ is covered by the cycles computed in phase $i_{e}$.

Proof. Consider phase $i$ where we need to cover the edges of $X_{i}$. Recall, that the algorithm marks the set of nodes $v$ with $(p(v), v) \in X_{i}$, resulting in the set $M_{i}$. Let $\Sigma$ be the output pairs of Algorithm TreeEdgeDisjointPath $\left(T_{1}^{\prime}, M_{i}\right)$. We first show that at least half of the edges in $X_{i}$ are covered by the paths of $\Sigma$.

If $M_{i}$ is odd, we omit one of the marked nodes and then apply Algorithm TreeEdgeDisjointPath to match the pairs in the even-sized set $M_{i}$. The key observation is that for every matched pair $\left\langle v_{1}, v_{2}\right\rangle$, it holds that either $\left(p\left(v_{1}\right), v_{1}\right)$ or $\left(p\left(v_{2}\right), v_{2}\right)$ is on $\pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right)$ (or both). Hence, at least half of the edges of $X_{i}$ are on the edge disjoint paths $\pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right)$.

We therefore get that after $\ell=c \log n$ phases, we are left with $\left|M_{\ell}\right|=O(1)$ at that point if $\left|M_{\ell}\right|$ is odd, we omit one vertex $v^{*}$ such that $e^{*}=\left(p\left(v^{*}\right), v^{*}\right)$. Claim (I) follows.

We now consider (II), let $e=(p(v), v)$ and consider phase $i=i_{e}$ in which $e \in \pi\left(v, v^{\prime}, T_{1}^{\prime}\right)$ where $v^{\prime}$ is the matched pair of $v$. We show that all the edges of $e \circ P_{e}$ are covered by the cycles $\mathcal{C}_{i}$ computed in that phase. By definition, $\left\langle v, v^{\prime}\right\rangle$ belongs to $\Sigma$. By Claim $6, G_{\Sigma}$ can be colored by 3 colors, let $\Sigma_{j} \subseteq \Sigma$ be the color class that contains $\left\langle v, v^{\prime}\right\rangle$.

We will show that there exists a cycle $C$ in $\mathcal{C}_{i, j}$ that covers each edge $e^{\prime \prime} \in e \circ P_{e}$ exactly once. Recall that the algorithm applies Algorithm NonTreeCover which computes a cycle cover $\mathcal{C}_{i, j}^{\prime \prime}$ to cover all the virtual edges $\widehat{E}_{i, j}$ in $T_{2}^{\prime}$. Also, $\left(s(v), s\left(v^{\prime}\right)\right) \in \widehat{E}_{i, j}$.

Let $C^{\prime \prime}$ be the (simple) cycle in $\mathcal{C}_{i, j}^{\prime \prime}$ that covers the virtual edge $\left(s(v), s\left(v^{\prime}\right)\right)$. In this cycle $C^{\prime \prime}$ we have two types of edges: edges in $T_{2}^{\prime}$ and virtual edges $\left(s\left(v_{1}\right), s\left(v_{2}\right)\right)$. First, we transform $C^{\prime \prime}$ into a cycle $C^{\prime}$ in which each virtual edge $\widehat{e}=\left(s\left(v_{1}\right), s\left(v_{2}\right)\right)$ is replaced by a path $P(\widehat{e})=$ $\left(s\left(v_{1}\right), v_{1}\right) \circ \pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right) \circ\left(v_{2}, s\left(v_{2}\right)\right)$. Next, we transform $C^{\prime}$ into $C \subseteq G$ by replacing each edge $\left(v_{1}, s\left(v_{1}\right)\right) \in \widetilde{E}$ in $C^{\prime \prime}$ by the $v_{1}-s\left(v_{1}\right)$ path $P_{e_{1}}$ for $e_{1}=\left(p\left(v_{1}\right), v_{1}\right)$.

We now claim that the final cycle $C \subseteq G$, contains each of the edges $e \circ P_{e}$ exactly once, hence even if $C$ is not simple, making it simple still guarantees that $e \circ P_{e}$ remain covered. Since $T_{1}^{\prime}$ and $T_{2}^{\prime}$ are edge disjoint, we need to restrict attention only two types of $T_{1}^{\prime}$ paths that got inserted to $C$ : (I) the edge disjoint paths $\Pi_{i, j}=\left\{\pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right) \mid\left\langle v_{1}, v_{2}\right\rangle \in \Sigma_{j}\right\}$ and (II) the $v^{\prime}-s\left(v^{\prime}\right)$ paths $P_{e^{\prime}}$ for every edge $e^{\prime}=\left(p\left(v^{\prime}\right), v^{\prime}\right)$ (appears on $C^{\prime}$ ).

We first claim that there is exactly one path $\pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right) \in \Pi_{i, j}$ that contains the edge $e=$ $(p(v), v)$. By the selection of phase $i, e \in \pi\left(v, v^{\prime}, T_{1}^{\prime}\right)$ where $v^{\prime}$ is the pair of $v$. Since all paths $\Pi_{i, j}$ are edge disjoint, no other path contains $e$. Next, we claim that there is no path $\pi \in \Pi_{i, j}$ that passes through an edge $e^{\prime} \in P_{e}$. Since $e=(p(v), v) \in \pi\left(v, v^{\prime}, T_{1}^{\prime}\right)$ and all edges on $P_{e}$ are below $e$ on $T^{8}$, the path $\pi\left(v, v^{\prime}, T_{1}^{\prime}\right)$ does not contain any $e^{\prime} \in P_{e}$. In addition, since all pairs in $\Sigma_{j}$ are independent in $G_{\Sigma}$, there is no path in $\pi\left(\sigma^{\prime}\right) \in \Pi_{i, j}$ that intersects $P_{e}$ (as in such a case, $\sigma$ interferes with $\left\langle v, v^{\prime}\right\rangle$ ). We get that $e$ appears exactly once on $\Pi_{i, j}$ and no edge from $P_{e}$ appears on $\Pi_{i, j}$. Finally, we consider the second type of paths in $T_{1}^{\prime}$, namely, the $P_{e^{\prime}}$ paths. By construction, every $e^{\prime} \in X_{i}$ is in $I(T)$ and hence that $P_{e^{\prime}}$ and $P_{e}$ share no tree edge. We get that when replacing the edge $(v, s(v))$ with $P_{e}$ all edges $e^{\prime} \in P_{e}$ appears and non of the tree edges on $P_{e}$ co-appear on some other $P_{e^{\prime \prime}}$. All together, each edge on $e \circ P_{e}$ appears on the cycle $C$ exactly once. This completes the cover property.

Since the edge $e^{*}$ is covered by taking the fundamental cycle of its swap edge, we get that all edges of $E_{1,2}^{\prime}$ are covered. Since each edge $(v, s(v))$ belongs to one of these $E_{1,2}^{\prime}$ sets, the cover property is satisfied.

Congestion. A very convenient property of our partitioning of $T^{\prime}$ into two trees $T_{1}^{\prime}$ and $T_{2}^{\prime}$ is that this partitioning is closed for LCAs. In particular, for $j \in\{1,2\}$ then if $u, v \in T_{j}^{\prime}$, the LCA of $u, v$ in $T^{\prime}$ is also in $T_{j}^{\prime}$. Note that this is in contrast to blocks of Section 4.1.1 that are not closed to LCAs.

We begin by proving by induction on $i=\{1, \ldots, O(\log n)\}$ that all the trees $T^{\prime}, T^{\prime \prime} \ldots$ considered in the same recursion level $i$ are edge disjoint. In the first level, the claim holds vacuously as there is only the initial tree $T$. Assume it holds up to level $i$ and consider level $i+1$. As each tree $T_{j}$ in level $i-1$ is partitioned into two edge disjoint trees in level $i+1$, the claim holds.

Note that each edge $e=(v, s(v))$ is considered exactly once, i.e., in one recursion call on $T^{\prime}=T_{1}^{\prime} \cup T_{2}^{\prime}$ where without loss of generality, $v \in T_{1}^{\prime}$ and $s(v) \in T_{2}^{\prime} \backslash T_{1}^{\prime}$. By Claim 8, there is at most one edge $e^{*} \in E_{1,2}^{\prime}$, which we cover by taking the fundamental cycle of $\operatorname{Swap}\left(e^{*}\right)$ in $T$.

[^8]We first show that the congestion in the collection of all the cycles added in this way is bounded by $O(\log n)$. To see this, we consider one level $i$ of the recursion and show that each edge appears on at most 2 of the fundamental cycles $\mathcal{F}_{i}$ added in that level. Consider an edge $e^{*}$ that is covered in this way in level $i$ of the recursion. That is the fundamental cycle of $\operatorname{Swap}\left(e^{*}\right)$ given by $\pi\left(v^{*}, s\left(v^{*}\right)\right) \cup P_{e^{*}}$ was added to $\mathcal{F}_{i}$. Let $T^{\prime}$ be such that $T^{\prime}=T_{1}^{\prime} \cup T_{2}^{\prime}$ and $e^{*}=\left(p\left(v^{*}\right), v^{*}\right)$ is such that $v^{*} \in T_{1}^{\prime}$ and $s\left(v^{*}\right) \in T_{2}^{\prime}$. Since both $v$ and $s(v)$ are in $T^{\prime}$, the tree path $\pi\left(v^{*}, s\left(v^{*}\right)\right) \subseteq T^{\prime}$. As all other trees $T^{\prime \prime} \neq T^{\prime}$ in level $i$ of the recursion are edge disjoint, they do not have any edge in common with $\pi\left(v^{*}, s\left(v^{*}\right)\right)$. For the tree $T^{\prime}$, there are at most two fundamental cycles that we add. One for covering an edge in $E_{1,2}^{\prime}$ and one for covering an edge in $E_{2,1}^{\prime}$. Since $e^{*} \in I(T)$, and each edge appears on at most two paths $P_{e}, P_{e^{\prime}}$ for $e, e^{\prime} \in I(T)$, overall each edge appears at most twice on each of the cycles in $\mathcal{F}_{i}$ (once in each direction of the edge) and over all the $O(\log n)$ of the recursion, the congestion due to these cycles is $O(\log n)$.

It remains to bound the congestion of all cycles obtained by translating the cycles computed using Algorithm TreeCover. We do that by showing that the cycle collection $\mathcal{C}_{i}$ computed in phase $i$ to cover the edges of $E_{1,2}^{\prime}$ is an $O(D \log n, \log n)$ cover. Since there are $O(\log n)$ phases and $O(\log n)$ levels of recursion, overall it gives an $O\left(D \log n, \log ^{3} n\right)$ cover.

Since all trees considered in a given recursion level are edge disjoint, we consider one of them: $T^{\prime}$. We now focus on phase $i$ of $\operatorname{Algorithm} \operatorname{Tree} \operatorname{Cover}\left(T^{\prime}\right)$. In particular, we consider the output cycles $\mathcal{C}_{i, j}^{\prime \prime}$ for $j \in\{1,2,3\}$ computed by Algorithm NonTreeCover for the edges $\widehat{E}_{1,2}$ and $T_{2}^{\prime}$. Each edge $e \in T_{2}^{\prime}$ appears on $O(\log n)$ cycles of $\mathcal{C}_{i, j}^{\prime \prime}$. Each virtual edge $\widehat{e}=\left(s\left(v_{1}\right), s\left(v_{2}\right)\right)$ is replaced by an $s\left(v_{1}\right)-s\left(v_{2}\right)$ path $P(\widehat{e})=\left(s\left(v_{1}\right), v_{1}\right) \circ \pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right) \circ\left(s\left(v_{1}\right), v_{1}\right)$ in $G \cup \widetilde{E}$. Let $\mathcal{C}_{i, j}^{\prime}$ be the cycles in $G \cup \widetilde{E}$ obtained from $\mathcal{C}_{i, j}^{\prime \prime}$ by replacing the edges of $\widehat{e} \in \widehat{E}_{1,2}$ with the paths $P(\widehat{e})$ in $G \cup \widetilde{E}$. Note that every two paths $P(\widehat{e})$ and $P\left(\hat{e}^{\prime}\right)$ are edge disjoint for every $\widehat{e}, \widehat{e}^{\prime} \in \widehat{E}_{1,2}$. The edges $\left(s\left(v_{1}\right), v_{1}\right)$ of $\widetilde{E}$ gets used only in tree $T^{\prime}$ in that recursion level. Hence, each edge $\left(v_{1}, s\left(v_{1}\right)\right)$ appears on $O(\log n)$ cycles $C^{\prime}$ in $G \cup \widetilde{E}$. Since the paths $\pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right)$ are edge disjoint, each edge $e^{\prime} \in \pi\left(v_{1}, v_{2}, T_{1}^{\prime}\right)$ appears on at most $O(\log n)$ cycles $C^{\prime}$ in $G \cup \widetilde{E}$ (i.e., on the cycles translated from $C^{\prime \prime} \in \mathcal{C}_{i, j}^{\prime \prime}$ that contains the edge $\left.\widehat{e}=\left(s\left(v_{1}\right), s\left(v_{2}\right)\right)\right)$. Up to this point we get that each virtual edge $(v, s(v)) \in \widetilde{E}$ appears on $O(\log n)$ cycles of $\mathcal{C}_{i, j}^{\prime}$. Finally, when replacing $(v, s(v))$ with the paths $P_{(p(v), v)}$, the congestion in $G$ is increased by factor of at most 2 as every two path $P_{e}$ and $P_{e^{\prime}}$ for $e, e^{\prime} \in I(T)$, are nearly edge disjoint (each edge ( $z, w$ ) appears on at most twice of these paths, one time in each direction). We get that the cycle collection $\mathcal{C}_{i}$ is an $O(D \log n, \log n)$ cover, as desired.

### 4.2 Private Neighborhood Cover

An neighborhood cover, introduced by Awerbuch et al. [ABCP98] is a set of overlapping sets of nodes in the graph (called clusters) with the property that every node has its entire $r$-neighborhood contained in one of these clusters. We introduce the notion of Private Neighborhood Cover in which the graph $G$ is decomposed into overlapping trees $T\left(u_{1}\right), \ldots, T\left(u_{n}\right)$ such that each tree $T\left(u_{i}\right)$ contains the $r$-neighborhood of $u_{i}$ in $G$ but does not contain $u_{i}$. Hence, $T\left(u_{i}\right)$ provides the $d$-neighborhood of $u_{i}$ a way to communicate privately without their root $u_{i}$. As in the setting of classical neighborhood cover, also in this private variant, the goal is to compute a collection of trees (or clusters) with small overlap and small diameter. Thus, we are interested in the existence of a low-congestion private neighborhood cover. Note that the standard notion of neighborhood cover of Awerbuch et al. [ABCP98] plays a role in computing instance-optimal covers (see Section 5). This
is independent of the current section which studies private neighborhood cover to provide private channels between the neighbors of a given vertex.

Definition 6 (Private Neighborhood Cover). Let $G=\left(V=\left\{u_{1}, \ldots, u_{n}\right\}, E\right)$ be an 2-vertex connected graph and let $d \in\{1, \ldots, \operatorname{Diam}(G)\}$. The private $d$-neighborhood cover $\mathcal{N}$ of $G$ is a collection of $n$ subtrees $T\left(u_{1}\right), \ldots, T\left(u_{n}\right)$ in $G$ such that for every $i \in\{1, \ldots, n\}$ it holds that $\Gamma_{d}\left(u_{i}\right) \backslash\left\{u_{i}\right\} \subseteq T\left(u_{i}\right)$, but $u_{i} \notin T\left(u_{i}\right)$. An (d, c) private d-neighborhood cover $\mathcal{N}$ is a private d-neighborhood cover that satisfies:

1. $\operatorname{Diam}\left(T\left(u_{i}\right)\right)=O(\mathrm{~d})$ for every $i \in\{1, \ldots, n\}$,

## 2. Every edge $e \in E$ appears in at most $O$ (c) trees.

Note that since the graph is 2-vertex connected, all nodes of $\Gamma(u)$ are indeed connected in $G \backslash\{u\}$ for every node $u$. The main challenge is in showing that all $n$ neighborhood spanning trees (i.e., for each $\Gamma(u, G)$ ) can be both of small diameter and of low-congestion.

Our main contribution is for the most interesting case where $d=1$. Later on in the section, we show how private 1-neighborhood cover can be used to obtain $P_{2}$ cover, which is what we need for the distributed secure simulations.

Theorem 4 (Nearly-Optimal Neighborhood-Cover). For every 2-vertex connected graph $G$ with maximum degree $\Delta$ and diameter $D$, there exists $a(\mathrm{~d}, \mathrm{c})$ private 1-neighborhood cover with $\mathrm{d}=$ $O(D \cdot \Delta \cdot \log n)$ and $\mathrm{c}=O\left(D \cdot \log \Delta \cdot \log ^{3} n\right)$.

In fact we show that given a construction of $(\mathrm{d}, \mathrm{c})$ cycle cover $\mathcal{C}$, we can construct an $(\mathrm{d} \cdot \Delta, \mathrm{c}$. $D \cdot \log \Delta)$ private neighborhood cover $\mathcal{N}$. In particular, using our construction of $\left(D \log n, \log ^{3} n\right)$ cycle cover $\mathcal{C}$ yeilds the desired bound.

The construction of the private 1-neighborhood cover $\mathcal{N}$ consists of $\ell=O(\log \Delta)$ phases. In each phase, we compute an $\left(D \log n, \log ^{3} n\right)$ cycle cover in some auxiliary graph using Theorem 1 . We start by having for each node $u$ an empty forest $F_{0}(u)=(\Gamma(u, G), \emptyset)$ consisting only of $u$ 's neighbors. Then in each phase, we add edges to these forests so that the number of connected components (containing the neighbors $\Gamma(u, G)$ ) is reduced by factor 2 . After $O(\log \Delta)$ phases, we will have for every $u \in V$, a tree $T(u)$ in $G \backslash\{u\}$ that spans all neighbors $\Gamma(u, G)$. We need some notation. Let $\mathcal{C}_{0}$ be a cycle cover of $G$. For every $i \in\{0, \ldots, \ell\}$, let $C C_{i}(u)$ be the number of connected components in the forest $F_{i}(u)$. In particular, $C C_{0}(u)=\operatorname{deg}(u)$.

In each phase $i \geq 1$, we are given a collection of forests $\mathcal{N}_{i-1}=\left\{F_{i-1}\left(u_{1}\right), \ldots, F_{i-1}\left(u_{n}\right)\right\}$ such that (I) $F_{i-1}\left(u_{j}\right) \subseteq G \backslash\left\{u_{j}\right\}$, (II) $\Gamma\left(u_{j}\right) \subseteq V\left(F_{i-1}\left(u_{j}\right)\right)$ and (III) $F_{i-1}\left(u_{j}\right)$ has $C C_{i-1}\left(u_{j}\right) \leq$ $\operatorname{deg}\left(u_{j}\right) / 2^{i-1}$ connected components. The goal of phase $i$ is to add edges to each $F_{i-1}\left(u_{j}\right)$ in order to reduce the number of connected components by factor 2 . The algorithm uses the current collection of forests $\mathcal{N}_{i-1}$ to define an auxiliary graph $\widetilde{G}_{i}$ which contains the edges of $G$ and some additional virtual nodes and edges. For every $u \in V$, we add to $\widetilde{G}_{i}$ a set of $k=C C_{i-1}(u)$ virtual nodes $\widetilde{u}_{1}, \ldots, \widetilde{u}_{k}$. We connect $u$ to each of its virtual copies $\widetilde{u}_{j}$ and in addition connect the $j^{\text {th }}$ virtual copy $\widetilde{u}_{j}$ is connected to all the neighbors of $u$ (in $\Gamma(u, G)$ ) that belong to the $j^{\text {th }}$ connected component in the forest $F_{i-1}(u) \in \mathcal{N}_{i-1}$. This is done for every $u \in G$. The final auxiliary graph $\widetilde{G}_{i}$ has $O(m)$ nodes, $O(m)$ edges and diameter at most $2 D$. Note that each edge of $\widetilde{G}_{i}$ has one real endpoint (i.e., node of $G$ ) and one virtual endpoint. Next, the algorithm uses Algorithm CycleCover to compute an $\left(D \log n, \log ^{3} n\right)$ cycle cover $\widetilde{\mathcal{C}_{i}}$ for the edges of $\widetilde{G}_{i}$. To map these virtual cycles to real cycles $\mathcal{C}_{i}$ in $G$, we simply replace a virtual node $\widetilde{u}_{j}$ with the real node $u$. As the virtual neighbors
of $u$, namely, $\widetilde{u}_{j}$, are connected to the neighbors of $u$, this indeed defines legal cycles in $G$. Define $G_{i}(u)=F_{i-1}(u) \cup\left\{C \mid C \in \mathcal{C}_{i}\right.$ and $\left.(u, v) \in C\right\} \backslash\{u\}$ and let $F_{i}(u) \subseteq G_{i}(u)$ be a forest that spans all the neighbors of $u$. This forest can be computed, for instance, by running a BFS from a neighbor $u$ in each connected component of $G_{i}(u)$. This completes the description of phase $i$. The final neighborhood cover is given by $\mathcal{N}=\left\{F_{\ell}\left(u_{1}\right), \ldots, F_{\ell}\left(u_{n}\right)\right\}$. We now turn to analyze this construction and prove Theorem 4.
Small Diameter Trees. We begin by showing that the diameter of each tree $T\left(u_{i}\right)$ is bounded by $O(\Delta D \cdot \log n)$. Note that this bound is existentially tight (up to logarithmic factors) as there are graphs $G$ with diameter $D$ and there is a node $u$ with degree $\Delta$ such that the diameter of $G \backslash\{u\}$ is $O(\Delta D)$.

Claim 9. For every $i \in\{0, \ldots, \log \Delta\}$ and for every $u \in V$ the number of connected components satisfies $C C_{i}(u) \leq \Delta / 2^{i}$.

Proof. The lemma is shown by induction on $i$. The case of $i=0$ holds vacuously. Assume that the claim holds up to $i-1$ and consider phase $i$. By construction, for each $u$, the auxiliary graph $\widetilde{G}_{i}$ contains $C C_{i-1}(u)$ virtual nodes $\widetilde{u}_{j}$ that are connected to $u$.

The cycle cover $\widetilde{\mathcal{C}}_{i}$ for $\widetilde{G}_{i}$ covers all these virtual $\left(u, \widetilde{u}_{j}\right)$ by virtual cycles, each such cycle connects two virtual nodes. Since every two virtual nodes of $u$ in $\widetilde{G}_{i}$ are connected to neighbors of $u$ that belong to different components in $G_{i-1}(u)$, every cycle that connects two virtual neighbors is mapped into a cycle that connects two of $u$ 's neighbors that belong to a different connected component in phase $G_{i-1}(u)$. Hence, the number of connected components in the forest $F_{i}(u)$ has been decreased by factor at least 2 compared to that of $F_{i-1}(u)$.

Claim 10. The diameter of each tree $T\left(u_{i}\right) \in \mathcal{N}$ is $O(\Delta \cdot D \cdot \log n)$.
Proof. We first claim that the diameter of each component in the forest $F_{i}(u)$ is bounded by $O(\Delta \cdot D \cdot \log n)$ for every $u \in V$ and every $i \in\{1, \ldots, \ell\}$. To see this, note that the forest $F_{i}(u)$ is formed by collection of $O(D \log n)$-length cycles that connect $u$ 's neighbors. Hence, when removing $u$, we get paths of length $O(D \log n)$. Consider the process where in each phase $i$, every two $u$ 'sneighbors that are connected by a cycle in $\mathcal{C}_{i}$ are connected by a single "edge". By the Proof of Claim 9, after $\ell=O(\log \Delta)$ phases, we get a connected tree with $\operatorname{deg}(u)$ nodes, and hence of "diameter" $\operatorname{deg}(u)$. Since each edge corresponds to a path of length $O(D \log n)$ in $G$, we get that the final diameter of $F_{\ell}(u)$ is $O(\operatorname{deg}(u) \cdot D \cdot \log n)$.

## Congestion.

Claim 11. Each edge e appears on $O\left(D \log ^{3} n\right)$ different subgraphs $T\left(u_{i}\right) \in \mathcal{N}$.
We first show that the cycles $\mathcal{C}_{i}$ computed in $G$ have congestion $O\left(\log ^{3} n\right)$ for every $i \in\{1, \ldots, \ell\}$. Clearly, the cycles $\widetilde{\mathcal{C}}_{i}$ computed in $\widetilde{G}_{i}$ have congestion of $O\left(\log ^{3} n\right)$. Consider the mapping of cycles $\widetilde{\mathcal{C}}_{i}$ in $\widetilde{G}_{i}$ to a cycles $\mathcal{C}_{i}$ in $G$. Edges of the type $\left(u, \widetilde{u}_{j}\right)$ are replaced by $(u, u)$ and hence there is no real edge in the cycle. Edges of the type $\left(\widetilde{u}_{j}, w\right)$ are replaced by $(u, w)$. Since there is only one virtual node of $u$ that connects to $w$, and since $\left(\widetilde{u}_{j}, w\right)$ appears in $O\left(\log ^{3} n\right)$ many cycles, also ( $u, w$ ) appears in $O\left(\log ^{3} n\right)$ many cycles (i.e., this conversion does not increase the congestion).

Note that the cycle $C$ of each edge ( $u, v$ ) joins the $G_{i}$ subgraphs of at most $D$ nodes since in our construction a cycle $C$ might cover up to $D$ edges. (Recall that in Algorithm TreeCover a cycle
that covers an a tree edge $e$, also covers up to $D$ edges on the path $P_{e}$ ). In addition, each edge $e^{\prime}$ appears on different cycles in $\mathcal{C}_{i}$.

We now claim that each edge $e$ appears on $O\left(i \log ^{3} n \cdot D\right)$ graphs $G_{i}(u)$. For $i=1$, this holds as the cycle $C$ of an edge $(u, v)$ joins the subgraphs $G_{1}(x)$ and $G_{1}(y)$ for every edge $(x, y)$ that is covered by $C$. Assume it holds up to $i-1$ and consider phase $i$. In phase $i$, we add to the $G_{i}(u)$ graphs the edges of $\mathcal{C}_{i}$. Again, each cycle $C^{\prime}$ of an edge $(u, v)$ joins $D$ graphs $G_{i}(x), G_{i}(y)$ for every $(x, y)$ that is covered by $C^{\prime}$. Hence each edge $e$ appears on $O\left(D \cdot \log ^{3} n\right)$ of the subgraphs $G_{i}\left(u_{j}\right) \backslash G_{i-1}\left(u_{j}\right)$. By induction assumption, each $e$ appears on $(i-1) \log ^{3} n \cdot D$ graphs $G_{i-1}\left(u_{j}\right)$ and hence overall each edge $e$ appears on $O\left(i \log ^{3} n\right)$ graphs $G_{i}\left(u_{j}\right)$. Therefore we get that each edge appears on $O\left(\log \Delta \cdot \log ^{3} n \cdot D\right)$ trees in $\mathcal{N}$. Finally, we show that we can construct low-congestion 2-neighborhood cover using the construction for 1-neighborhood.

Corollary 1 (Private 2-Neighborhood Cover). Given an (d, c) 1-neighborhood cover $\mathcal{N}$, there exists an $(\mathrm{d}+1, \mathrm{c}+\Delta) 2$-neighborhood cover $\mathcal{N}^{\prime}$.

Proof. For each $y$, and for every $w \in \Gamma_{2}(y, G)$ that is not in $T(y)$, we add $w$ to $T(y) \in \mathcal{N}$ by connecting $w$ to some neighbor $z \in \Gamma(y, G) \cap \Gamma(w, G)$. Let $T^{\prime}(y)$ be the resulting tree and let $\mathcal{N}^{\prime}=\left\{T^{\prime}(y), y \in V\right\}$. Clearly, $\Gamma_{2}(y, G) \subseteq V\left(T^{\prime}(y)\right)$. Since each edge $(w, z)$ that we add to $T(y)$ might be added to $\Delta$ many trees $y^{\prime}$ for $y^{\prime} \in \Gamma(z)$, each edge appears in $\mathrm{c}+\Delta$ many trees in $\mathcal{N}^{\prime}$.

In particular, plugging in our 1-neighborhood cover we get the following corollary:
Corollary 2. For every 2 -node connected $n$-node graph with maximum degree $\Delta$ and diameter $D$, there exists a ( $\mathrm{d}, \mathrm{c}$ ) private 2-neighborhood cover with $\mathrm{d}=\widetilde{O}(D \Delta)$ and $\mathrm{c}=\widetilde{O}(\Delta+D)$.

We observe that our bounds are nearly optimal. To see this, consider a node $u$ that has $\Delta$ neighbors and $\Theta\left(\Delta^{2}\right)$ nodes at distance 2, that is $\left|\Gamma_{2}(u)\right|=\Theta\left(\Delta^{2}\right)$. The node $u$ must participate in the neighborhood cover of each node $v \in \Gamma_{2}(u)$, where for each such cover, at least one edge connected to $u$ is covered. That is, there are $\Theta\left(\Delta^{2}\right)$ covers where each contains at least one of the $\Delta$ edges of $u$, and therefore there must be an edge with congestion at least $\Theta(\Delta)$.

## 4.3 $\quad P_{2}$ Cycle Cover

In section, we consider cycles that cover not only a single edge, but rather paths of length 2 . In particular, such cycles provide two-node disjoint paths between each node $u$ and its second hop neighbor $v \in \Gamma_{2}(u, G)$. We begin by providing a general definition of $P_{k}$ cycle cover which can be viewed as a natural extension to (edge) cycle cover which is in fact a $P_{1}$ cycle cover.

Definition 7 ( $P_{k}$ Cycle Cover). Let $G=(V, E)$ be an n-node graph and let $k$ be an integer. $A P_{k}$ Cycle Cover $\mathcal{C}$ of $G$ is a collection of simple cycles that cover all paths of length $k$ in $G$. That is, for every length-k path $P^{\prime}$ in $G$, there exists a cycle $C \in \mathcal{C}$ such that $P^{\prime} \subseteq C$.

An (d, c) $P_{k}$ cycle cover $\mathcal{C}$ satisfies in addition that each cycle $C \in \mathcal{C}$ is of length at most $O(\mathrm{~d})$ and each edge appears in at most $O(\mathrm{c})$ cycles in $\mathcal{C}$.

It is easy to see that $P_{k}$ cycle cover exist (i.e., each $P_{k}$ path in the graph is covered by a cycle) iff the graph is $(k-1)$-node connected. For our algorithmic purposes, it is sufficient to consider the case of $P_{2}$ cycle cover. An alternative viewpoint on $P_{2}$ cycle cover is that it computes low-congestion collection consisting of two node disjoint $u-v$ paths for every $u \in V$ and $v \in \Gamma_{2}(u)$. We note
that our construction of $P_{2}$ cycle cover can be extended to general $P_{k}$ covers, with the expense of stricter bounds and connectivity constraints.

We prove the existence of a low-congestion $P_{2}$ cover. Namely, we show that every $u-v$ path of length 2 is contained in at least one simple cycle of $\mathcal{C}$ (i.e., this is the covering property of $\mathcal{C})$. In addition, all cycles in $\mathcal{C}$ are of length $O\left(\Delta \cdot D \cdot \log ^{2} n\right)$ and each edge appears on at most $O\left(\Delta^{2} \log \Delta \cdot \log ^{3} n \cdot D\right)$ different cycles.

Theorem 5 ( $P_{2}$ cycle cover). For every two-node connected $n$-node graph $G=(V, E)$ with diameter $D$ and maximum degree $\Delta$, there exists a $\left(\Delta \cdot D \cdot \log n, \Delta^{2} \cdot D \cdot \log \Delta \log ^{3} n\right) P_{2}$ cycle cover $\mathcal{C}$. Moreover, these parameters are existentially optimal (up to poly-logarithmic factors).

Proof. In fact we show that given a construction of ( $\mathrm{d}, \mathrm{c}$ ) private neighborhood cover $\mathcal{N}$, one can have also a ( $\mathrm{d}, \mathrm{c} \cdot \Delta^{2}$ ) $P_{2}$ cover $\mathcal{C}$.

We will use the construction of $\left(D \cdot \Delta \cdot \log n, \log ^{3} n \cdot \log \Delta \cdot D\right)$ private neighborhood cover $\mathcal{N}$ to define a simple cycle $C_{x, y, z}$ for every $y \in V$ and every $x, z \in \Gamma(y, G)$. To define cycles $C_{x, y, z}$ for a particular node $y$, we use the tree $T(y) \in \mathcal{N}$. This tree contains an $x-z$ path $\pi(x, z, T(y))$ in $G \backslash\{y\}$ for every $x, z \in \Gamma(y, G)$. Now, define

$$
C_{x, y, z}=\pi(x, z, T(y)) \circ(x-y-z), \text { for every } y \in V \text { and } x, z \in \Gamma(y), \text { and } \mathcal{C}=\bigcup_{y \in V} \bigcup_{x, z \in \Gamma(y)} C_{x, y, z} .
$$

Clearly, every cycle in $\mathcal{C}$ is simple and has length $O(D \cdot \Delta \cdot \log n)$ (i.e., the bound on the diameter of each tree $T(y) \in \mathcal{N}$ ). In addition, every two-edge path $x-y-z$ in $G$ is covered by one of the cycles in $\mathcal{C}$. It remains to bound the congestion.

By the properties of $\mathcal{N}$, each edge $e$ appears on $\ell=O\left(\log \Delta \log ^{3} n \cdot D\right)$ different trees $T\left(y_{1}\right), \ldots$, $T\left(y_{\ell}\right)$. For a fixed $y_{i}$, each edge $e \in T\left(y_{i}\right)$ appears on $O\left(\Delta^{2}\right)$ cycles $C_{x, y_{i}, z}$ as there are $\Delta^{2}$ pairs $x, z$ in $\Gamma\left(y_{i}, G\right)$. Hence, $e$ appears on $O\left(\Delta^{2} \cdot D \cdot \log \Delta \cdot \log ^{2} n\right)$ segments of $\pi(x, z, T(y)) \subseteq C_{x, y, z}$. Note that the cycle $C_{x, y, z}$ also contains the edges $(x, y)$ and ( $y, z$ ) which are not in the neighborhood cover of $y$ (since $y \notin T(y)$ ). Each edge $(x, y)$ appears on at most $2 \Delta P_{2}$ cycles in this role (i.e., not as part of a tree $T(y))$. Hence, overall each edge appears on $O\left(\Delta^{2} \cdot D \cdot \log \Delta \cdot \log ^{3} n\right) P_{2}$ cycles in $\mathcal{C}$.

We next claim that these parameters are existentially tight up to logarithmic factors. Since the degree of each node is $\Delta$, the diameter of $G \backslash\{u\}$ might be $\Omega(D \cdot \Delta)$. Since we compute two node disjoint $y-z$ paths for every $y \in V$ and $z \in \Gamma_{2}(y)$, one of the paths might be of length $\Omega(D \cdot \Delta)$ and so is the length of the cycle covering a given two-edge path. Turning to congestion, we show that there are graphs in which in any $P_{2}$ cycle cover $\mathcal{C}$, there must bean edge $e$ that appears on $\Omega\left(\Delta^{2}\right)$ cycles. In particular, this holds for the edge $e$ in the figure. The node $y$ has two sets of $\Omega(\Delta)$ neighbors, $L$ and $R$. Any cycle covering the path $x-y-z$ for $x \in L$ and $z \in R$ must go through $e$. Hence, overall $\Omega\left(\Delta^{2}\right)$ cycles go through $e$.

## 5 Universally (Nearly) Optimal Covers

The goal of this section is to construct covers that are competitive with the best possible cover for a given graph (rather than being competitive with respect to the worst possible graph).


Figure 13: Illustration of $P_{2}$ cycle cover with congestion $\Omega\left(\Delta^{2}\right)$. The node $y$ in the above graph is such that all simple cycles covering the paths $x-y-z$ for $x \in L$ and $z \in R$ must through the same edge $e$. Thus, $e$ has congestion at least $\Omega\left(\Delta^{2}\right)$.

### 5.1 Cycle Covers

For a given graph $G$, let $\mathrm{OPT}_{G}$ be the optimal dilation of any cycle cover (even with unbounded congestion) for $G$. In this section, we show the construction of $\left(\widetilde{O}\left(\mathrm{OPT}_{G}\right), \widetilde{O}(1)\right)$ cycle cover $\mathcal{C}$. Our nearly optimal cycle cover can also be tuned to be nearly optimal with respect to each edge $e$ in the sense that each edge $e$ is covered in the cycle cover by a cycle of length $\mathrm{OPT}_{G, e}$, where $\mathrm{OPT}_{G, e}$ is the length of the shortest cycle in $G$ that contains $e$.

Theorem 9. (I) For any graph $G$, one can construct an $\left.\widetilde{O}\left(\mathrm{OPT}_{G}\right), \widetilde{O}(1)\right)$ cycle cover $\mathcal{C}$. (II) In addition, for each edge $e \in G$, there is a cycle $C_{e}$ in $\mathcal{C}$ containing e such that $\left|C_{e}\right|=\widetilde{O}\left(\mathrm{OPT}_{G, e}\right)$.

We say that an algorithm $\mathcal{A}$ for constructing a ( $\mathrm{d}, \mathrm{c}$ ) cycle cover $\mathcal{C}$ is nice if it does not require that $G$ is bridgeless (i.e., 2 edge connected) but rather covers by cycles all edges that participate in some cycle. That is, for every graph $G$ and for every edge $e \in G$ such that $G \backslash\{e\}$ is connected, it holds that $e$ is covered by a cycle in $\mathcal{C}$. We first claim that our main algorithm (Algorithm CycleCover) is nice.

Claim 12. Algorithm CycleCover is nice, that is it covers all edge that participate in some cycle and does not require that the input graph is 2-edge connected.

Proof. Let $e$ be an edge for which there is cycle in $G$. If $e$ is covered by the first part of the algorithm (where a collection of edge disjoint cycles is computed), then we are done. It remains to consider two cases. Case (1) is where $e$ is a non-tree edge covered by Algorithm NonTreeCover. In such a case, by Lemma 2, the edge $e$ is covered by the output cycle collection of Algorithm NonTreeCover. Case (2) is where $e$ is a tree edge and hence handled by Algorithm TreeCover. Recall that in this algorithm, for each tree edge $e^{\prime}=\left(u^{\prime}, v^{\prime}\right)$, we consider its swap edge $\operatorname{Swap}\left(e^{\prime}\right)$. Since the edge $e$ belongs to some cycle in $G$, it implies that $G \backslash\{e\}$ is connected and hence there is a swap edge Swap ( $e$ ) that restores the connectivity of $T \backslash\{e\}$. In Algorithm TreeCover, we make sure that tree edge that has a swap edge is covered by the output cycles. In particular, each swap edge $\operatorname{Swap}\left(e^{\prime}\right)$ is replaced by a virtual non-tree edge that is incident to $v^{\prime}$, and there is some cycle that covers
both the virtual edge along with its corresponding tree edge $e^{\prime}$. These cycles are later converted to cycles in $G$ which cover all tree edges that have swap edges. The claim follows.

Now, to show Theorem 9 (I), we in fact prove the following transformation lemma that holds for every nice algorithm.

Lemma 4. For any graph $G$ with diameter $D$, every nice algorithm $\mathcal{A}$ that constructs an $(f(D), c)$ cover $\mathcal{C}$ for $G$ can be transformed into an algorithm $\mathcal{A}^{\prime}$ that constructs an $\left(f\left(\widetilde{O}\left(\mathrm{OPT}_{G}\right)\right), \widetilde{O}(\mathrm{c})\right)$ cover for $G$.

Applying the above transformation to Algorithm CycleCover we get that Theorem 9 (I) follows. The key tool of this transformation Lemma is a t-neighborhood cover, a set of clusters in the graph with small overlap, such that for every vertex there exists a cluster which contains its entire $t$-neighborhood.

Definition 8 (Neighborhood Cover, [ABCP96]). $A(k, t, q)$ neighborhood cover of an n-vertex graph $G=(V, E)$ is a collection of subsets of $V$ (denoted as clusters), $\mathcal{S}=\left\{S_{1}, \ldots, S_{r}\right\}$ with the following properties:

- For every vertex $v$, there exists a cluster $S_{i}$ such that $\Gamma_{t}(V) \subseteq S_{i}$.
- The strong diameter ${ }^{9}$ of each cluster $S_{i}$ is at most $O(k \cdot t)$.
- Each vertex belongs to at most q clusters.

Awerbuch et al. showed how to construct such a neighborhood cover in the LOCAL model.
Lemma 5 ([ABCP96]). For every integers $k$, , and every $n$-vertex graph $G=(V, E)$, there is a $(k, t, q)$ neighborhood cover with $q=O\left(k \cdot n^{1 / k}\right)$. In particular, for $k=O(\log n)$ we get that $q=\widetilde{O}(1)$ and the diameter of each cluster is $\widetilde{O}(t)$.

For each edge $e=(u, v)$, let $C_{e}$ be the shortest cycle in $G$ that contains $e$. Define $\alpha_{G}=$ $\max _{e \in G}\left|C_{e}\right|$ to be the length of the longest $C_{e}$ cycle among all $e \in G$. It is easy to see that every (d, c) cycle cover $\mathcal{C}$ of $G$ must satisfy that $\mathrm{d} \geq \alpha_{G}$ (regardless of the congestion c). Clearly, $\alpha_{G}$ is always upper bounded by $O(D)$ but it can be much smaller. Since $\mathrm{OPT}_{G}$ is the optimal cover regardless of congestion, we have that $\mathrm{OPT}_{G}=\alpha_{G}$. We now prove Lemma 4, and construct a cycle cover of dilation $\widetilde{O}\left(\mathrm{OPT}_{G}\right)$ while preserving a low congestion of $\widetilde{O}(1)$.

Proof. Let $\mathcal{A}$ be a nice algorithm that constructs an $(f(D), c)$ cycle cover. We now describe algorithm $\mathcal{A}^{\prime}$ that uses $\mathcal{A}$ to construct an $\left(f\left(\widetilde{O}\left(\alpha_{G}\right)\right), \widetilde{O}(c)\right)$ cycle cover. Algorithm $\mathcal{A}^{\prime}$ starts by constructing an $(k, t, q)$ neighborhood cover $\mathcal{S}=\left\{S_{1}, \ldots, S_{r}\right\}$ for $k=O(\log n)$ and $t=\alpha_{G}$. Hence, the diameter $D_{i}$ of each graph $G\left(S_{i}\right)$ is $O\left(\log n \cdot \alpha_{G}\right)$. Then, it applies Algorithm $\mathcal{A}$ in each subgraph $G\left[S_{i}\right]$ and computes an $\left(f\left(D_{i}\right), \mathrm{c}\right)$ cycle cover $\mathcal{C}_{i}$ for every $i \in\{1, \ldots, r\}$. The final cover is given by $\mathcal{C}^{*}=\bigcup_{i=1}^{r} \mathcal{C}_{i}$. This completes the description of the algorithm.

We first show that $\mathcal{C}^{*}$ is a cover. Consider an edge $e=(u, v)$. Let $S_{i}$ be such that $\Gamma_{\alpha_{G}}(u) \subseteq S_{i}$. By the definition of neighborhood cover such $S_{i}$ exists. Since $e$ belongs to a cycle $C_{e}$ in $G$ of length at most $\alpha_{G}$, it holds that $C_{e} \subseteq G\left[S_{i}\right]$. Since $\mathcal{A}$ is nice, each edge in $G^{\prime}$ that belongs to some cycle in $G^{\prime}$ is covered by the output cycles of Algorithm $\mathcal{A}$. Hence, $e$ is covered by the cycles

[^9]of $\mathcal{C}_{\widetilde{O}}$. By construction, the diameter of $G\left[S_{i}\right]$ is $\widetilde{O}\left(\alpha_{G}\right)$ and hence each cycle in $\mathcal{C}^{*}$ has length $f\left(\widetilde{O}\left(\alpha_{G}\right)\right)$. It remains to bound the congestion. Each edge appears on at most c cycles in $\mathcal{C}_{i}$ for every $i \in\{1, \ldots, r\}$. Since each vertex $v$ appears on $O(\log n)$ different clusters $S_{j} \in \mathcal{S}$, overall, we get that each edge appears on $\widetilde{O}(c)$ cycles in $\mathcal{C}^{*}$.

We next show that by slightly adopting the construction algorithm of Lemma 4, we can obtain also local (near) optimality - i.e., optimality (up to logarithmic factors) for each of the edges. Recall that for each edge $e, C_{e}$ is the shortest cycle that contains $e$.

Lemma 6. For every graph $G=(V, E)$, one can construct a cycle cover $\mathcal{C}$ such that each edge $e$ is covered in $\mathcal{C}$ be a cycle of length $\widetilde{O}\left(\left|C_{e}\right|\right)$ and each edge appears on $\widetilde{O}(1)$ cycles.

Proof. The idea is to apply the algorithm of Lemma 4 for $O\left(\log \alpha_{G}\right)$ many times. In step $i=$ $\left\{1, \ldots,\left\lceil\log \alpha_{G}\right\rceil\right\}$, we construct a $\left(k=O(\log n), t=2^{i}, q\right)$ neighborhood cover $\mathcal{S}_{i}=\left\{S_{i, 1}, \ldots, S_{i, r_{i}}\right\}$, such that each subgraph $G\left[S_{i, j}\right]$ has diameter $D_{i, j}=O\left(\log n \cdot 2^{i}\right)$ and each vertex belongs to $O(\log n)$ subsets $S_{i, j}$. We then apply Algorithm CycleCover in each subgraph $G\left[S_{i, j}\right]$, resulting in an $\left(\widetilde{O}\left(\log n \cdot 2^{i}\right), \widetilde{O}(1)\right)$ cycle cover $\mathcal{C}_{i, j}$. The final cover is $\mathcal{C}^{*}=\bigcup_{i, j} \mathcal{C}_{i, j}$.

We first claim that each edge $e=(u, v)$ belongs to a cycle in $\mathcal{C}^{*}$ of length $\widetilde{O}\left(\left|C_{e}\right|\right)$. Let $i$ be such that $\left|C_{e}\right| \in\left[2^{i-1}, 2^{i}\right]$. Then by the properties of the neighborhood cover, there exists a subset $S_{i, j}$ such that $V\left(C_{e}\right) \subseteq S_{i, j}$. By the same argument as in Lemma $4, e$ is covered by a cycle of length $\widetilde{O}\left(2^{i}\right)=\widetilde{O}\left(\left|C_{e}\right|\right)$ in $\mathcal{C}_{i, j}$. Finally, by the proof of Lemma 4, each edge appears on $\widetilde{O}(1)$ cycles in $\bigcup_{j} \mathcal{C}_{i, j}$ for every $i \in\left\{1, \ldots,\left\lceil\log \alpha_{G}\right\rceil\right\}$. As there are $O(\log n)$ steps, overall, each edge appears on $\widetilde{O}(1)$ cycles in $\mathcal{C}^{*}$.

### 5.2 Private Neighborhood Cover and $P_{2}$ Covers

We now turn to consider constructions of private neighborhood covers that are competitive with the best possible one. Let $O P T_{G}$ be the optimal depth of private neighborhood cover (regardless of the congestion). Let $d_{v}=\operatorname{Diam}(G \backslash\{v\})$. Clearly, there are bad graph examples $G$ with maximum degree $\Delta$ and diameter $D$, for which there exists $v \in G$ such that $d_{v}=\Theta(\Delta \cdot D)$. However, in general $d_{v}$ might be much small then the universal upper bound of $O(\Delta \cdot D)$. The goal of this subsection is to construct a private neighborhood cover in which the depth of each tree of $v$ is competitive with $d_{v}$.

Corollary 3. For every 2 -vertex connected $n$-vertex graph $G$ with maximum degree $\Delta$ and diameter $D$, there exists a (d, c) private 1-neighborhood cover $\mathcal{N}=\left\{T\left(u_{1}\right), \ldots, T\left(u_{n}\right)\right\}$ with $\mathrm{d}=\widetilde{O}\left(O P T_{G} \cdot \Delta\right)$ and $\mathrm{c}=\widetilde{O}\left(O P T_{G}\right)$. In addition, each tree $T\left(u_{i}\right)$ has diameter of $\widetilde{O}\left(\sqrt{\Delta} \cdot d_{v}\right)$.

Proof. The construction is similar to that of Theorem 4 only that we apply Lemma 4with Algorithm CycleCover. Since for each edge $e=(u, v)$, it holds that $\left|C_{e}\right|=O\left(d_{v}\right)$, the computed cycles when constructing $T(v)$ are of length $\widetilde{O}\left(d_{v}\right)$.

Next we consider $P_{2}$-Covers. For each $P_{2}$ path $P$, let $C_{P}$ be the shortest cycle in $G$ that contains $P$. Clearly the dilation of every $P_{2}$-cover is at least $\beta_{G}=\max \left\{C_{P}|P \subseteq G,|P|=2\}\right.$. In addition, it is easy to see that $\beta_{G} \geq \alpha_{G}$. By combining Corollary 3 and Corollary 1, we get:

Corollary 4. For every 2 -node connected $n$-node graph with maximum degree $\Delta$, there exists a ( $\mathrm{d}, \mathrm{c}$ ) private 2-neighborhood cover with $\mathrm{d}=\widetilde{O}\left(\alpha_{G} \cdot \Delta\right)$ and $\mathrm{c}=\widetilde{O}\left(\Delta+\alpha_{G}\right)$.

## 6 Secure Simulation via Low-Congestion Covers

In this section we describe how to transform any distributed algorithm $\mathcal{A}$ to a new algorithm $\mathcal{A}^{\prime}$ which has the same functionality as $\mathcal{A}$ (i.e., the output for every node $u$ in $\mathcal{A}$ is the same as in $\mathcal{A}^{\prime}$ ) but has perfect privacy (as is defined in Definition 1). Towards this end, we assume that the combinatorial structures required are already computed (in a preprocessing stage), namely, a 2 -path cover and a neighborhood cover for vertices in the graph. The output of the preprocessing stage is given in a distributed manner. The (distributed) output of the cycle cover is such that each vertex $v$ knows for each of its incident edges $(u, v)$ all edge IDs $e^{\prime}$ such that $(u, v)$ appears on the cycle that covers $e^{\prime}$. The output of the private neighborhood cover for each node $u$, is such that each vertex $v$ knows its parent in the private neighborhood cover of $u$ (if such exists). Finally, in $P_{2}$ cover, each vertex $v$ knows for each of its incident edges $e=(u, v)$, the IDs of all $P_{2}$ paths ${ }^{10}$ that are covered by cycles that go through $e$.

Theorem 6. Let $G$ be an n-vertex graph with diameter $D$ and maximal degree $\Delta$. Let $\mathcal{A}$ be a natural distributed algorithm that works on $G$ in $r$ rounds. Then, $\mathcal{A}$ can be transformed to an equivalent algorithm $\mathcal{A}^{\prime}$ with perfect privacy which runs in $\widetilde{O}(r D \cdot \operatorname{poly}(\Delta))$ rounds (after a preprocessing stage).

As a preparation for our secure simulation, we provide the following convenient view of distributed algorithm.

### 6.1 Our Framework

We treat the distributed $r$-round algorithm $\mathcal{A}$ from the view point of some fixed node $u$, as a collection of $r$ functions $f_{1}, \ldots, f_{r}$ as follows. Let $\Gamma(u)=\left\{v_{1}, \ldots, v_{k}\right\}$. At any round $i$, the memory of $u$ consists of a state, denoted by $\sigma_{i}$ and $\Delta$ messages $m_{v_{1} \rightarrow u} \ldots, m_{v_{\Delta} \rightarrow u}$ that were received in the previous round (in the degree of the node is less than $\Delta$ the rest of the messages are empty). Initially, we set $\sigma_{0}$ to be a fixed string and initialize all messages to NULL. At round $i$ the node $u$ updates its state to $\sigma_{i+1}$ according to its previous state $\sigma_{i}$ and the messages that it got in the previous round. It then prepares $k$ messages to send $m_{u \rightarrow v_{1}}, \ldots, m_{u \rightarrow v_{\Delta}}$. To ease notation (and without loss of generality) we assume that each state contains the ID of the node $u$. Thus, we can focus on a single update function $f_{i}$ for every round that works for all nodes. The function $f_{i}$ gets the state $\sigma_{i}$, the messages $m_{v_{1} \rightarrow u \ldots, m_{v_{\Delta} \rightarrow u}}$, and the randomness $s$. The output of $f_{i}$ is the next state $\sigma_{i+1}$, and at most $k$ outgoing messages:

$$
\left(\sigma_{i}, m_{u \rightarrow v_{1}}, \ldots, m_{u \rightarrow v_{\Delta}}\right) \leftarrow f_{i}\left(\sigma_{i-1}, m_{v_{1} \rightarrow u}, \ldots, m_{v_{\Delta} \rightarrow u}, s\right)
$$

Our compiler works round-by-round where each round $i$ is replaced by a collection of rounds that "securely" compute $f_{i}$, in a manner that will be explained next. The complexity of our algorithm depends exponentially on the space complexity of the functions $f_{i}$. Thus, we proceed by transforming the original algorithm $\mathcal{A}$ to one in which each $f_{i}$ can be computed in logarithmic space, while slightly increasing the number of rounds.

Claim 13. Any natural distributed algorithm $\mathcal{A}$ the runs in $r$ rounds can be transformed to a new algorithm $\widehat{\mathcal{A}}$ with the same output such that $\widehat{\mathcal{A}}$ is computable in logarithmic space using $r^{\prime}=$ $r \cdot \operatorname{poly}(\Delta+\log n)$ rounds.

[^10]Proof. Let $t$ be the running time of the function $f_{i}$. Then, $f_{i}$ can be computed with a circuit of at most $t$ gates. Note that since $\mathcal{A}$ is natural, it holds that $t \leq \operatorname{poly}(\Delta, \log n)$.

Instead of letting $u$ computing $f_{i}$ in round $i$, we replace the $i$ th round by $t$ rounds where each round computes only a single gate of the function $f_{i}$. These new rounds will have no communication at all, but are used merely for computing $f_{i}$ with a small amount of memory.

Let $g_{1}, \ldots, g_{t}$ be the gates of the function $f_{i}$ in a computable order where $g_{t}$ is the output of the function. We define a new state $\sigma_{i}^{\prime}$ of the form $\sigma^{\prime}=\left(\sigma_{i}, g_{1}, \ldots, g_{t}\right)$, where $\sigma_{i}$ is the original state, and $g_{j}$ is the value of the $j$ th gate. Initially, $g_{1}, \ldots, g_{t}$ are set to $\perp$. Then, for all $j \in[t]$ we define the function

$$
f_{i}^{j}\left(\sigma_{i}, g_{1}, \ldots, g_{j-1}, \perp, \ldots, \perp\right)=\left(\sigma_{i}, g_{1}, \ldots, g_{j-1}, g_{j}, \perp, \ldots, \perp\right)
$$

In the $j$ th round we compute $f_{i}^{j}$, until the final $g_{t}$ is computed. Note that $f_{i}^{j}$ can be computed with $\log$ arithmic space, and since $t \leq \operatorname{poly}(\Delta, \log n)$ we can compute $f_{i}^{j}$ with space $O(\log \Delta+$ $\log \log n)$. As a result, the $r$-round algorithm $\mathcal{A}$ is replaced by an $r t$-round algorithm $\widehat{\mathcal{A}}$, where $t \leq \operatorname{poly}(\Delta, \log n)$. That is, we have that $r^{\prime} \leq \operatorname{poly}(\Delta, \log n)$.

As we will see, our compiler will have an overhead of $\operatorname{poly}(\Delta, \log n)$ in the round complexity and hence the overhead of Claim 13 is insignificant. Thus, we will assume that the distributed algorithm $\mathcal{A}$ satisfies that all its functions $f_{i}$ are computable in logarithmic space (i.e., we assume that the algorithm is already after the above transformation).

### 6.2 Secure Simulation of a Single Round

In the algorithm $\mathcal{A}$ each node $u$ computes the function $f_{i}$ in each round $i$. In our secure algorithm $\mathcal{A}^{\prime}$ we want to simulate this computation, however, on encrypted data, such that $u$ does not get to learn the true output of $f_{i}$ in any of the rounds except for the last one. When we say "encrypted" data, we mean a "one-time-pad" (see Definition 4). That is, we merely refer to a process where we the data is masked by XORing it with a random string $K$. Then, $K$ is called the encryption (and also decryption) key. Using this notion, we define a related function $f_{i}^{\prime}$ that, intuitively, simulates $f_{i}$ on encrypted data, by getting an encrypted state and messages as input, decrypting them, then computing $f_{i}$ and finally encrypting back the output. We simulate every round of the original algorithm $\mathcal{A}$ by a PSM protocol for the function $f_{i}^{\prime}$.
The Secure Function $f_{i}^{\prime}$. The function $f_{i}^{\prime}$ gets the following inputs (encrypted elements will be denoted by the $\widehat{\text { notation): }}$

1. An encrypted state $\widehat{\sigma}_{i-1}$ and encrypted messages $\left\{\widehat{m}_{v_{j} \rightarrow u}\right\}_{j=1}^{\Delta}$.
2. Shares for the decryption key of the state $\left\{R_{\sigma_{i-1}}^{j}\right\}_{j=1}^{\Delta}$ and messages $\left\{R_{v_{j} \rightarrow u}^{k}\right\}_{j, k=1}^{\Delta}$.
3. Shares for randomness $\left\{R_{s}^{j}\right\}_{j=1}^{\Delta}$ for the function $f_{i}$.
4. Secret shared keys for encrypting the new state $\left\{R_{\sigma_{i}}^{j}\right\}_{j=1}^{\Delta}$ and messages $\left\{R_{u \rightarrow v_{j}}^{k}\right\}_{j, k=1}^{\Delta}$.

The function $f_{i}^{\prime}$ decrypts the state and messages and runs the function $f_{i}$ (using randomness $s$ ) to get the new state $\sigma_{i}$ and the outgoing messages $m_{u \rightarrow v_{1}}, \ldots, m_{u \rightarrow v_{\Delta}}$. Then, it encrypts the new state and messages using the encryption keys. In total, the function $f_{i}^{\prime}$ has $O\left(\Delta^{2}\right)$ inputs. The precise description of $f_{i}^{\prime}$ is given in Figure 14.

## The description of the function $f_{i}^{\prime}$.

Input: A secret shared encrypted state $\left\{\widehat{\sigma}_{i-1}^{j}\right\}_{j=1}^{\Delta}$, encrypted messages $\left\{\widehat{m}_{v_{j} \rightarrow u}\right\}_{j=1}^{\Delta}$, secret shared keys for decrypting the input $\left\{R_{\sigma_{i-1}}^{j}\right\}_{j=1}^{\Delta},\left\{R_{v_{j} \rightarrow u}^{k}\right\}_{j, k=1}^{\Delta}$, secret shared randomness $\left\{R_{s}^{j}\right\}_{j=1}^{\Delta}$ and secret shared keys for encrypting the output $\left\{R_{\sigma_{i}}^{j}\right\}_{j=1}^{\Delta},\left\{R_{u \rightarrow v_{j}}^{k}\right\}_{j, k=1}^{\Delta}$.

## Run:

1. Compute $\sigma_{i-1} \leftarrow\left(\bigoplus_{j=1}^{\Delta} \widehat{\sigma}_{i-1}^{j}\right) \oplus\left(\bigoplus_{j=1}^{\Delta} R_{\sigma_{i-1}}^{j}\right)$ and $s \leftarrow\left(\bigoplus_{j=1}^{\Delta} R_{s}^{j}\right)$.
2. For $j=1 \ldots \Delta$ : compute $m_{v_{j} \rightarrow u} \leftarrow \widehat{m}_{v_{j} \rightarrow u} \oplus\left(\bigoplus_{k=1}^{\Delta} R_{v_{j} \rightarrow u}^{k}\right)$.
3. Run $\sigma_{i}, m_{u \rightarrow v_{1}}, \ldots, m_{u \rightarrow v_{\Delta}} \leftarrow f\left(\sigma_{i-1}, m_{v_{1} \rightarrow u}, \ldots, m_{v_{\Delta} \rightarrow u}, s\right)$.
4. Compute $\widehat{\sigma_{i}} \leftarrow \sigma_{i} \oplus\left(\bigoplus_{j=1}^{\Delta} R_{\sigma_{i}}^{j}\right)$.
5. For $j=1 \ldots \Delta$ : compute $\widehat{m}_{u \rightarrow v_{j}} \leftarrow m_{u \rightarrow v_{j}} \oplus\left(\bigoplus_{k=1}^{\Delta} R_{u \rightarrow v_{j}}^{k}\right)$.
6. Output $\widehat{\sigma}_{i}, \widehat{m}_{u \rightarrow v_{1}}, \ldots, \widehat{m}_{u \rightarrow v_{\Delta}}$.

Figure 14: The function $f_{i}^{\prime}$.

Recall that in the PSM model, we have $k$ parties $p_{1}, \ldots, p_{k}$ and a server $s$, where it was assumed that (PI) all parties have private shared randomness (not known to $s$ ) and (PII) that each of party has a private communication channel to the server. Our goal is to compute $f_{i}^{\prime}$ securely by implementing a PSM protocol for all nodes in the graph simultaneously. We call this implementation a distributed PSM algorithm.

The distributed PSM algorithm securely computes $f_{i}^{\prime}$ by simulating the PSM protocol for $f_{i}^{\prime}$, treating $u$ as the server and its $2^{\text {nd }}$-neighborhood as the parties. The private neighborhood cover and the $P_{2}$ cycle cover provide the graph theoretical and algorithmic framework for implementing this distributed PSM protocol in general graph topologies by allowing the simultaneous computation for all nodes. For property (PI) we need to provide for every node $u$ in the graph a tree $T_{u}$ that spans all the vertices of $\Gamma_{2}(u)$ (i.e., the parties) in $G \backslash\{u\}$. Using this tree, all the parties can share private randomness that is not known to $u$. This can be done in $O\left(\operatorname{Diam}\left(T_{u}\right)+R\right)$ rounds, where $\operatorname{Diam}\left(T_{u}\right)$ is the diameter of the tree and $R$ is the number of random bits. To allow the simultaneous communication on these trees, we want each edge to appear in small number of trees and in addition, we would like the trees to be of small depth. These two requirements are encapsulated in our notion of low-congestion private-neighborhood-cover.

For property (PII) we need to provide a way for each party $p_{i}$ to communicate privately with the node $u$. If $p_{i}$ is an immediate neighbor of $u$ then this is achieved trivially. For a 2-hop neighbor $p_{i}$, we simulate a private channel between $u$ and $p_{i}$ by computing two vertex-disjoint paths between $p_{i}$ and $u$. To have a private channel between $u$ and $p_{i}$, we use the idea of secret sharing: $p_{i}$ sends a private message $m$ to $u$ by secret sharing the message to two parts (i.e., sampling random $m^{1}, m^{2}$ conditioned on $m^{1} \oplus m^{2}=m$ ) and sending $m^{1}$ on one $u-p_{i}$ path and $m^{2}$ on the other. In this way, only the node $u$ gets both shares of the secret and thus it is the only node that can reconstruct
the message $m$. The two disjoint paths from $p_{i}$ to $u$ can be obtained by computing a simple cycle that covers the (length 2) path from $p_{i}$ to $u$. The number of rounds required for communicating over this $u-p_{i}$ channel is proportional to the length of the cycle. Since we would like all nodes to communicate this way simultaneously with all their 2-hop neighbors, it is desired that each of these cycle be short and that each edge appears in a small number of different cycles. These two requirements are encapsulated in our low-congestion $P_{2}$ cycle cover. The final algorithm $\mathcal{A}_{i}^{\prime}(u)$ for securely computing $f_{i}^{\prime}$ is described in Figure 15.

The algorithm $\mathcal{A}_{i}^{\prime}(u)$ for securely computing $f_{i}^{\prime}$.
Input: Each node $v \in \Gamma_{2}(u)$ has input $x_{v} \in\{0,1\}^{m}$.

1. Let $T$ be the tree spanning $\Gamma_{2}(u)$ in $G \backslash\{u\}$ and let $w$ be the root.
2. $w$ chooses a random string $R$ and sends it to $\Gamma_{2}(u)$ using the tree $T$.
3. Each node $v \in \Gamma_{2}(u)$ computes $M_{v}=\operatorname{PSM} . \operatorname{Enc}\left(f_{i}^{\prime}, x_{v}, R\right)$.
4. Each immediate neighbor $v \in \Gamma_{1}(u)$ sends $M_{v}$ to $u$ directly.
5. Each 2-hop neighbor $v \in \Gamma_{2}(u) \backslash \Gamma_{1}(u)$ do as follows:
(a) Secret share the message $M_{v}$ to $M_{v}=M_{v}^{1} \oplus M_{v}^{2}$.
(b) Let $P_{v \rightarrow u}^{1}$ and $P_{v \rightarrow u}^{2}$ be two vertex disjoint paths from $v$ to $u$ (as obtained from the $P_{2}$ cover).
(c) Send $M_{v}^{1}$ to $u$ via $P_{v \rightarrow u}^{1}$ and $M_{v}^{2}$ via $P_{v \rightarrow u}^{2}$.
6. $u$ computes $M_{v}=M_{v}^{1} \oplus M_{v}^{2}$ for each 2-hop neighbor $v \in \Gamma_{2}(u) \backslash \Gamma_{1}(u)$.
7. $u$ computes $y=\operatorname{PSM} . \operatorname{Dec}\left(f_{i}^{\prime},\left\{M_{v}\right\}_{v \in \Gamma_{2}(u)}\right)$.

Figure 15: The description of the distributed PSM algorithm of node $u$ for securely computing the function $f_{i}^{\prime}$.

In what follows analyze the security and round complexity of Algorithm $\mathcal{A}_{i}^{\prime}$.
Round Complexity. Let $f:\{0,1\}^{m \cdot\left|\Gamma_{2}(u)\right|} \rightarrow\{0,1\}^{\ell}$ be a function with $\left|\Gamma_{2}(u)\right| \leq \Delta^{2}$ inputs, where each input is of length $m$ bits. The communication complexity of the PSM protocol depends on the input and output length of the function and also on the memory required to compute $f$. Suppose that $f$ is computable by an $s$-space TM. Then, by Theorem 8 the communication complexity (and randomness complexity) of the protocol is at most $O\left(\Delta^{2} m \ell \cdot 2^{2 s}\right)$.

In the first phase of the protocol, the root $w$ sends a collection of random bits $R$ to $\Gamma_{2}(u)$ using the private 2-neighborhood cover, where $|R|=O\left(\Delta^{2} \cdot m \cdot \ell \cdot 2^{2 s}\right)$. By Corollary 2, the diameter of the tree is at most $\widetilde{O}(D \Delta)$ and each edge belongs to $\widetilde{O}(\Delta)$ different trees. Therefore, there are total of $\widetilde{O}(\Delta \cdot|R|)$ many bits that need to go through a single edge when sending the information on all trees simultaneously. Using the random delay approach of Theorem 7, this can be done in $\widetilde{O}\left(D \Delta+\Delta^{3} \cdot m \cdot \ell \cdot 2^{2 s}\right)$ rounds. For the rest of the protocol, we need to send the PSM messages of length $|R|$ on the edges of the $P_{2}$-cycle cover. By Theorem 5 , the length of these cycles is at most $\widetilde{O}(D \cdot \Delta)$ and each edge appears on $\widetilde{O}\left(D \cdot \Delta^{2}\right)$ cycles. Hence, on each edge we need to
send $\widetilde{O}\left(D \cdot \Delta^{2} \cdot|R|\right)$ bits. Using the random delay approach again, this can be done on all cycles simultaneously using $\widetilde{O}\left(D \Delta^{4} \cdot m \cdot \ell \cdot 2^{2 s}\right)$ rounds. This is summarized by the following Lemma:

Lemma 7. Let $f:\left(\{0,1\}^{m}\right)^{\Delta^{2}} \rightarrow\{0,1\}^{\ell}$ be a function over $\Delta^{2}$ inputs where each is of length at most $m$ and that is computable by a s-space TM. Then, there is a distributed algorithm $\mathcal{A}_{i}^{\prime}(u)$ (in the CONGEST model) with perfect privacy where each node $u$ outputs $f$ evaluated on $\Gamma_{2}(u)$. The round complexity of $\mathcal{A}_{i}^{\prime}(u)$ is $\widetilde{O}\left(D \Delta^{4} \cdot m \ell \cdot 2^{2 s}\right)$.

### 6.3 The Final Secure Algorithm

Using the function $f_{i}^{\prime}$, we define the algorithm $\mathcal{A}_{u}^{\prime}$ for computing the next state and messages of the node $u$. We describe the algorithm for any $u$ in the graph and at the end we show that all the algorithms $\left\{\mathcal{A}_{u}^{\prime}\right\}_{u \in G}$ can be run simultaneously with low congestion.

The algorithm $\mathcal{A}_{u}^{\prime}$ involves running the distributed algorithm $\mathcal{A}_{i}^{\prime}(u)$ for all rounds $i$ on the 2neighborhood of $u$. First, each immediate neighbor $v_{j}$ of $u$ samples a share of an encryption key $R_{\sigma_{i}}^{j}$ for encrypting the next state of $u$. Moreover, $v_{j}$ sample a share of the randomness $R_{s}^{j}$ used to evaluate the function $f_{i}$. Each neighbor $w_{k}$ of $v_{j}\left(w_{k}\right.$ is of distance 2 from $\left.u\right)$ samples a share of an encryption key $R_{u \rightarrow v_{j}}^{k}$ for encrypting the message $m_{u \rightarrow v_{j}}$ from $u$ to $v_{j}$.

Then they run $\mathcal{A}_{i}^{\prime}(u)$ algorithm with $u$ as the server and $\Gamma_{2}(u)$ as the parties for computing the function $f_{i}^{\prime}$ (see Figure 15). The node $u$ has the encrypted state and message, the neighbors of $u$ have the (encryption and decryption) keys for the state (and randomness). For each neighbor $v_{j}$ of $u$ the neighbors of $v_{j}$ (not including $u$ ) have the (encryption and decryption) keys for the messages sent from $u$ to $v_{j}$. At the end of the protocol, $u$ computes the output of $f_{i}^{\prime}$ which is the encrypted output of the function $f_{i}$ and sends the encrypted message $\widehat{m}_{u \rightarrow v_{j}}$ to $v_{j}$.

After the final round, $u$ holds an encryption of the final state $\widehat{\sigma_{r}}$ which contains only the output of the original algorithm $A$. At this point, the neighbors of $u$ send it the decryption key for this last state, $u$ decrypts its state and outputs the decrypted state. Initially, the state $\sigma_{0}$ is a fixed string which is not encrypted, and the all encryption keys for this round are assumed to be 0 . This description is summarized in Figure 16.

Finally, we show that the protocol is correct and secure.
Correctness. The correctness follows directly from the construction. Consider a node $u$ in the graph. Originally, $u$ computes the sequence of states $\sigma_{0}, \ldots, \sigma_{r}$ where $\sigma_{r}$ contained the final output of the algorithm. In the compiled algorithm $\mathcal{A}^{\prime}$, for each round $i$ of $\mathcal{A}$ and every node $u$ the sub-algorithm $\mathcal{A}_{i}^{\prime}(u)$ computes $\widehat{\sigma_{i}}$, where $\widehat{\sigma}_{i}=\sigma_{i} \oplus\left(\bigoplus_{j=1}^{\Delta} R_{\sigma_{i}}^{j}\right)$, and each neighbor $v_{j} \in \Gamma_{1}(u)$ has $R_{\sigma_{i}}^{j}$. Thus, after the last round, $u$ has $\widehat{\sigma_{r}}$ and each of its neighbor $v_{j}$ has $R_{\sigma_{r}}^{j}$. Thus, $u$ computes $\widehat{\sigma}_{r} \oplus\left(\bigoplus_{j=1}^{\Delta} R_{\sigma_{r}}^{j}\right)=\sigma_{r}$ and outputs $\sigma_{r}$ as required.
Round Complexity. We compute the number of rounds of the algorithm for any natural algorithm $\mathcal{A}$. The algorithm consists of $r^{\prime}=r \cdot \operatorname{poly}(\Delta+\log n)$ iterations. In each iteration, every vertex $u$ implements algorithm $\mathcal{A}_{i}^{\prime}$ for the function $f_{i}^{\prime}$ (there are other operations in the iteration but they are negligible). We know that $f_{i}$ can be computed in $s$-space where $s=O(\log \Delta+\log \log n)$, and thus we can bound the size of each input to $f_{i}^{\prime}$ by poly $(\Delta) \cdot$ polylog $(n)$. Indeed, the state has this bound by the definition of a natural algorithm, and thus also the encrypted state (which has the exact same size), the messages and encryption keys for the messages have length at most $\log n$, and the randomness shares are of size at most the running time of $f_{i}$ which is at most $2^{s}$ where $s$ is the space of $f_{i}$ and thus the bound holds. The output length shares the same bound as well.

## The description of the algorithm $\mathcal{A}_{u}^{\prime}$.

1. For each round $i=1 \ldots r$ do:
(a) $u$ secret shares $\widehat{\sigma}=\bigoplus_{j=1}^{\Delta} \widehat{\sigma}_{i}^{j}$ and sends $\widehat{\sigma}_{i}^{j}$ to neighbor $v_{j}$.
(b) For each neighbor $v_{j}$ of $u$ :
i. $v_{j}$ samples $R_{\sigma_{i}}^{j}$ and $R_{s}^{j}$ at random (and stores it).
ii. Each neighbor $w_{k}$ of $v_{j}$ samples $R_{u \rightarrow v_{j}}^{k}$ at random (and stores it).
(c) Run the $\mathcal{A}_{i}(u)$ algorithm for $f_{i}^{\prime}$ with server $u$ and parties $\Gamma_{2}(u)$ where:
i. Each neighbor $v_{j}$ of $u$ has input $\widehat{\sigma}_{i}^{j}, \widehat{m}_{v_{j} \rightarrow u}, R_{\sigma_{i-1}}^{j}, R_{\sigma_{i}}^{j}, R_{s}^{j}$.
ii. Each neighbor $w_{k}$ of $v_{j}$ has input $R_{v_{j} \rightarrow u}^{k}, R_{u \rightarrow v_{j}}^{k}$.
iii. $u$ learns the final output of the algorithm $\left(\widehat{\sigma}_{i}, \widehat{m}_{u \rightarrow v_{1}}, \ldots, \widehat{m}_{u \rightarrow v_{\Delta}}\right)$.
(d) $u$ sends $\widehat{m}_{u \rightarrow v_{j}}$ to $v_{j}$ for all $j \in[\Delta]$.
2. Each neighbor $v_{j}$ of $u$ sends $R_{\sigma_{r}}^{j}$ to $u$.
3. $u$ computes $\sigma_{r}=\widehat{\sigma}_{r} \oplus\left(\bigoplus_{j=1}^{\Delta} R_{\sigma_{r}}^{j}\right)$ and outputs $\sigma_{r}$.

Figure 16: The description of the Algorithm $\mathcal{A}_{u}^{\prime}$. In the description above, we assume that in "round 0 " all keys are initialized to 0 . That is, we let $R_{\sigma_{0}}^{j}=0$ for all $j \in[\Delta]$, and initially set $R_{v_{j} \rightarrow u}^{k}=0$ for all $j, k \in[\Delta]$.

Since $f_{i}$ can be computed in $s$-space where $s=O(\log \Delta+\log \log n)$, we observe that $f_{i}^{\prime}$ can be computed in $s$-space as well. This includes running $f_{i}$ is a "lazy" manner. That is, whenever the TM for computing $f_{i}$ asks to read a the $i^{\text {th }}$ bit of the input, we generate the this bit by performing the appropriate XOR operations for the $i^{\text {th }}$ bit of the input elements. The memory required for this is only storing indexes of the input which is $\log (\Delta \cdot \operatorname{poly}(\log n))$ bits and thus $s$ bits suffice.

Then, by Lemma 7 we get that the $\mathrm{PSM}_{f_{i}^{\prime}}$ algorithm runs in $\widetilde{O}(D \cdot$ poly $(\Delta))$ rounds, and the total number of rounds of our algorithm is $\widetilde{O}(r D \cdot \operatorname{poly}(\Delta))$. In particular, if the degree $\Delta$ is bounded by polylog $(n)$ then we get $\widetilde{O}(r D)$ number of rounds.

Remark 1 (Round complexity for non-natural algorithms). If $\mathcal{A}$ is not a "natural" algorithm then we can bound the number of rounds with dependency on the time complexity of the algorithm. If each function $f_{i}$ (the local computation of the nodes) can be computed by a circuit of size $t$ then the number of rounds of the compiled algorithm is bounded by $\widetilde{O}(r D t \cdot \operatorname{poly}(\Delta))$.

Security. We begin by describing the security of a single sub-protocol $\mathcal{A}_{u}^{\prime}$ for any node $u$ in the graph. The algorithm $\mathcal{A}_{u}^{\prime}$ has many nodes involved, and we begin by showing how to simulate the messages of $u$. Fix an iteration $i$, and consider the all the messages sent to $u$ by the PSM protocol in $\mathcal{A}_{i}^{\prime}(u)$ denoted by $\left\{M_{v}\right\}_{v \in G}$, and let $\widehat{\sigma}_{i}, \widehat{m}_{u \rightarrow v_{1}}, \ldots, \widehat{m}_{u \rightarrow v_{\Delta}}$ be the output of the protocol. By the security of the PSM protocol, there is a simulator Sim such that the following two distributions are equal:

$$
\left\{M_{v}\right\}_{v \in G} \equiv \operatorname{Sim}\left(\widehat{\sigma}_{i}, \widehat{m}_{u \rightarrow v_{1}}, \ldots, \widehat{m}_{u \rightarrow v_{\Delta}}\right)
$$

Since $\widehat{\sigma}_{i}$ and $\widehat{m}_{u \rightarrow v_{1}}, \ldots, \widehat{m}_{u \rightarrow v_{\Delta}}$ are encrypted by keys that are never sent to $u$ we have that from the viewpoint of $u$ the distribution of $\widehat{\sigma}_{i}$ and of $\widehat{m}_{u \rightarrow v_{1}}, \ldots, \widehat{m}_{u \rightarrow v_{\Delta}}$ are uniformly random. Thus, we can run the simulator with a random string $R$ of the same length and have

$$
\operatorname{Sim}\left(\widehat{\sigma}_{i}, \widehat{m}_{u \rightarrow v_{1}}, \ldots, \widehat{m}_{u \rightarrow v_{\Delta}}\right) \equiv \operatorname{Sim}(R) .
$$

While this concludes the simulator for $u$, we need to show a simulator for other nodes that participate in the protocol. Consider the neighbors of $u$. Each neighbor $v_{j}$ of $u$ has the elements $\widehat{\sigma}_{i}^{j}, \widehat{m}_{v_{j} \rightarrow u}, R_{\sigma_{i-1}}^{j}, R_{\sigma_{i}}^{j}, R_{s}^{j}$ in its viewpoint. All these elements are uniform random strings which can be simulated by a simulator Sim by sampling a random string of the same length. Indeed, $\widehat{\sigma}_{i}^{j}$ is a share of $\widehat{\sigma}_{i}$ where the other shares are shared among the rest of the neighbors of $u$, and each share individually is completely uniform. The messages $\widehat{m}_{v_{j} \rightarrow u}$ are encrypted by a key that is shared among the neighbors of $v_{j}$ and thus it is again uniform. Finally, $R_{\sigma_{i}}^{j}, R_{s}^{j}$ are chosen to be random and thus can be simulated. In additional, $v_{j}$ receives messages of the PSM protocol. It gets the shared random string, which can be trivially simulated, and in addition it might receive the share $M_{w}^{b}$ for $b \in\{1,2\}$ for some node $w$. However, the other share $M_{w}^{3-b}$ is sent on a vertex-disjoint path to $u$ and thus $v$ does not see it. Therefore, this share alone is a uniformly random string. A similar reasoning holds for distance 2 neighbors of $u$ and also for all other nodes in the graph that participate in the protocol, as they only receive a subset of the messages that the neighbours receive.

To conclude, the privacy of $\mathcal{A}_{i}^{\prime}(u)$ follows from the perfect privacy of PSM protocol we use. The PSM security guarantees a perfect simulator for the server's viewpoint, and it is easy to construct a simulator for all other parties in the protocol as they only receive random messages. While the PSM was proven secure in a stand alone setting, in our protocol we have a composition of many instances of the protocol. Fortunately, it was shown in [KLR10] that any protocol that is perfectly secure and has a black-box non-rewinding simulator, is also secure under universal composability, that is, security is guaranteed to hold when many arbitrary protocols are performed concurrently with the secure protocol. We observe that the PSM has a simple simulator that is black-box and non-rewinding, and thus we can apply the result of [KLR10]. This is since the simulator of the PSM protocol is an algorithm that runs the protocol on an arbitrary message that agrees with the output of the function.

## 7 Distributed Computation of Low Congestion Covers

In this section, we describe the distributed algorithms for constructing the low-congestion covering structures. We start by describing the construction of low-congestion cycle cover.

### 7.1 Cycle Cover

We describe Algorithm DistCycleCover which constructs a low congestion cycle cover in the distributed setting. The output of the algorithm is defined as follows: the endpoints of every edge $e=(u, v)$ know all the identifiers of all the edges that are covered by the cycles that go through $e$.

Theorem 3. For every 2-edge connected n-node graph $G=(V, E)$ with diameter $D$ and maximum degree $\Delta$, Algorithm DistCycleCover computes a $\left(4^{1 / \epsilon} \cdot D \cdot \log \Delta, n^{\epsilon} \log ^{2} n\right)$ cycle cover $\mathcal{C}$ in $\widetilde{O}\left(4^{1 / \epsilon}\right.$. $D+n^{\epsilon}$ ) rounds for every $\epsilon \in(0,1)$.

We start by claiming that up to logarithmic factors in the round complexity and in the quality of our cycle cover, it is sufficient to restrict attention to graphs with maximum degree 5 .

Lemma 8. Let $\mathcal{A}^{\prime}$ be an algorithm that computes a (d, c) cycle cover for every $N$-node graph $G^{\prime}$ with maximum degree 5 within $f(N, \mathrm{~d}, \mathrm{c}, \operatorname{Diam}(G))$ rounds. Then, there exists an algorithm $\mathcal{A}$ that computes $a(\mathrm{~d} \log \Delta, \mathrm{c})$ cycle cover for every $n$-node graph $G=(V, E)$ with maximum degree $\Delta$, within $f(|E|, \mathrm{d} \log \Delta, \mathrm{c}, \operatorname{Diam}(G) \cdot \log \Delta)$ rounds.

Proof. The 2-edge connected graph $G$ can be transformed to a graph $G^{\prime}$ with bounded degree 5 in the following manner: For every node $u$ with degree $\ell=\operatorname{deg}(u)$, we add in $G^{\prime}, 2 \ell-1$ many copies $\widehat{u}_{1}, \ldots, \widehat{u}_{2 \ell-1}$ that are connected as follows. The node $u$ is the root of a binary tree $\widehat{T}_{u}$ of depth $\log (\operatorname{deg}(u))$ and with the leaf nodes $\widehat{u}_{1}, \ldots, \widehat{u}_{\ell}$ which correspond to the $\Gamma(u, G)$ nodes. That is, each leaf node $\widehat{u}_{j}$ corresponds to the $j^{\text {th }}$ neighbor of $u$. In addition, all these leaf nodes are connected via a path. The leaf nodes of different trees are connected as follows: the leaf node $\widehat{u}_{j}$ in $\widehat{T}_{u}$ that corresponds to the $j^{\text {th }}$ neighbor $v \in \Gamma(u, G)$ is connected to the corresponding copy of $u$ in $\widehat{T}_{v}$, namely, $\widehat{v}_{i}$. Overall we get a 2-edge connected graph with $\Theta(|E(G)|)$ nodes and $\Theta(|E(G)|)$ edges, has maximum degree 5 and is of diameter $\Theta(\log \Delta \cdot \operatorname{Diam}(G))$. Any round of an algorithm for $G^{\prime}$ can be simulated by a single round in the graph $G$. In addition, any cycle computed in $G^{\prime}$ can be easily translated into a cycle in $G$ without increasing the congestion by replacing each virtual copy $\widehat{u}_{j}$ with the node $u$.

Given Lemma 8, we can assume without loss of generality, that $G$ has maximum degree $O(1)$. As in the centralized construction, we start by constructing a BFS tree $T$. We first cover all the non-tree edges $E^{\prime}$ using Algorithm DistNonTreeCover, and then cover the tree edges $E(T)$, using Algorithm DistTreeCover. Since we already assume that we work on bounded degree graphs (by paying an extra factor of $\log \Delta$ in the length of the cycles), there is no need for the preliminary step that computes short edge disjoint cycles of length $O(\log n)$.

### 7.1.1 Distributed Algorithm for Covering Non-Tree Edges

In this subsection we describe Algorithm DistNonTreeCover that constructs an ( $2^{1 / \epsilon} \cdot D, n^{2 \epsilon}$ ) cycle cover $\mathcal{C}_{1}$ for the non tree edges. We next describe two basic procedures used by algorithm and then describe how to used them to compute the cover.
Tool (I): Computing an Hierarchy of Block Partitioning. We are given a BFS tree $T$, a subset of non-tree edges $E^{\prime}$ and an input parameter $\epsilon \in[0,1]$. Computing the cycle cover is based upon constructing a hierarchy of ( $E^{\prime}, \mathrm{b}_{i}$ ) block partitioning where $\mathrm{b}_{i}=m / n^{i \cdot \epsilon}$ for every $i \in\{0, \ldots, \ell\}$ where $\ell=c \cdot\lceil 1 / \epsilon\rceil$. As in Section 4.1, every block consists of a subset of nodes with consecutive postorder traversal numbering. To see how we can compute the postorder numbering distributively, see Appendix B. We now define the hierarchical block partitioning and then explain how to compute it in a distributed manner. The 0-level of the hierarchy consists of one block containing all nodes. In the $i^{\text {th }}$-level of the hierarchy, each $(i-1)$-level block is subdivided into $n^{\epsilon}$ sub-blocks, each with bounded density of $\mathrm{b}_{i}=m / n^{i \cdot \epsilon}$. This subdivision is done greedily, going through the nodes in the block from the lowest numbered node to the highest and grouping them into blocks of capacity $b_{i}$. Just like in the centralized setting, each block either consists of a single node with at least $\mathrm{b}_{i}$ edges in $E^{\prime}$ or with a block of bounded density $O\left(\mathrm{~b}_{i}\right)$.

Turning to the distributed implementation, we consider the $i$-phase for $i \geq 1$, where we are given the ( $i-1$ )-level of the partitioning $B_{i-1,1}, \ldots, B_{i-1, \ell}$ such that each node know the ID of the ( $i-1$ )level block $B_{i, j}$ to which it belongs and the LCA of its block in $T$ (knowing the LCA is important for communicating in the tree $T_{B_{i-1, j}}$ ). Equipped with that knowledge, we now describe how to compute the $i^{\text {th }}$-level of the partitioning. We will work on each block separately but simultaneously. That is, in phase $i$, we run (simultaneously) a collection of $\ell$ algorithms, one per ( $i-1$ )-level block $B_{i-1, j}$. From this point on, we focus on one such ( $i-1$ )-block $B=B_{i-1, j}$ and describe Algorithm PartitionSingleBlock which partitions $B$ into $O\left(n^{\epsilon}\right)$ blocks, each with bounded density $\mathrm{b}_{i}$.

The output of this algorithm is that each node $u$ knows the block IDs and the LCAs of all its blocks in each of $O(1 / \epsilon)$ levels of the hierarchy. Throughout, the density of a subset $S \subseteq V$ is the number of edges in $E^{\prime}$ that have at least one endpoint in $S$. We say that a block in level $i$ is full if its density is at least $\mathrm{b}_{i}$. We need some notation. Almost all our computation for a block $B$ are restricted to the subtree in $T$ that connects all $B$ nodes, namely, the subtree $T_{B}=\bigcup_{u \in B} \pi(u, \ell, T)$ where $\ell$ is the LCA of $B$ in $T$. Note that $V\left(T_{B}\right)$ is not necessarily $B$, that is, $T_{B}$ might contain nodes that belong to different blocks than $B$. For $u \in B, b_{i}(u)$ is the ID of the $i$-level block of $u$. For $u \in T_{B}$ (which might not necessarily be in $B$ ), let $\max _{B}(u)$ be the node of maximum ID in $T_{B}(u) \cap B$ (if $u \in B$, then $\max _{B}(u)=u$ ). Finally, let $b_{i}^{*}(u)$ be the $i$-level block ID of $\max _{B}(u)$. We work from the leaf nodes of $T_{B}$ up the root (i.e., the LCA of $B$-nodes). Letting $D_{B}$ be the diameter of the tree $T_{B}$, at step $j \geq 0$, we assume that all nodes $u$ in layers $D_{B}-j$ of $T_{B}$ hold the following information:

- the block ID $b_{i}^{*}(u)$,
- the LCA of the block $b_{i}^{*}(u)$,
- the density of the block $b_{i}^{*}(u)$,
- the number of $i$-level blocks in $T_{B}(u) \cap B^{11}$.

These values are trivially computed for all leaf nodes in level $D_{B}$ of $T_{B}$. We now show how to compute this information in level $D_{B}-(j+1)$. All nodes in layer $D-j$ send their block information to their parents in $T_{B}$. Consider now a parent $u$ (in layer $D_{B}-(j+1)$ ) and its children in $T_{B} \cap B$ : $v_{1}, \ldots, v_{k}$, ordered from left to right. The node $u$ looks for a maximal sequence of children (from left to right) each with only one $i$-level block in their subtree $T_{B}\left(v_{p}\right) \cap B$. For any such maximal sequence, it merges these $i$-level blocks greedily, by filling up blocks as long as the density of the block is below $\mathrm{b}_{i}$. Once all mergings of $i$-level blocks of $u$ 's children is done, $u$ can obtain all desired information. If $u \in B$, and the last $i$-level block of its rightmost child is not completely full (i.e., its density is below $\mathrm{b}_{i}$ ), then $u$ joins that block, and otherwise, it opens a new block and the ID of the block is simply the ID of $u$.

Hence, it is easy to see that $u$ has all the required information as it can easily get the density of the block to which it belongs and compute the number of block in $T_{B}(u) \cap B$. This completes the description of phase $i$. After $D_{B}$ rounds of working bottom up on $T_{B}$, we start working top-down. Here, there is no need to work layer by layer, but rather each node $u$ that just merged the blocks of its children into a full block (with density at least $b_{i}$ ), sends all the information about that block (ID, LCA which is node $u$ ) to all its relevant children (whose block got merged) and this information propagates down their tree $T_{B}\left(v_{p}\right)$. Since we only merge subtrees that contain a single

[^11]$i$-level block, there is no congestion in propagating the information of the full block down the tree. This completes the description of Algorithm PartitionSingleBlock.

We analyze the partitioning algorithm and show:
Claim 14. Algorithm PartitionSingleBlock partitions an (i-1)-block $B_{i-1, j}$ into $O\left(n^{\epsilon}\right)$ i-level blocks.
Proof. Let $B^{\prime}=B_{i-1, j}$ and $T^{\prime}=T_{B^{\prime}}$ be the subtree of $T$ connecting all nodes of $B^{\prime}$. Let $D^{\prime}$ be the depth of $T^{\prime}$. For node $u \in T^{\prime}$, let $d\left(u, B^{\prime}\right)=\sum_{v \in\left(T^{\prime}(u) \cap B^{\prime}\right)} \operatorname{deg}\left(v, E^{\prime}\right)$.

The blocks are defined bottom up on $T^{\prime}$. We will claim by induction on the number of round $r \geq 0$, that for each node $u \in V\left(T^{\prime}\right)$ in layer $D^{\prime}-r$ (in $T^{\prime}$ ) the following holds: If $T^{\prime}(u) \cap B^{\prime}$ (the subtree of $u$ in $T^{\prime}$ ) has more than one block of level $i$, then for any $i$-level block in $T^{\prime}(u)$ with density less than $\mathrm{b}_{i}$, either the $i$-level block before it or after it in $T^{\prime}(u) \cap B^{\prime}$ has density at least $\mathrm{b}_{i}$. As a result, we will get that that the number of $i$-level blocks in $T(u) \cap B^{\prime}$ is at most $d\left(u, B^{\prime}\right) / \mathrm{b}_{i}$. Applying this to the root of $T^{\prime}$ (the LCA of the nodes of $B^{\prime}$ ), we get that there are at most $\operatorname{deg}\left(B^{\prime}, E^{\prime}\right) / \mathrm{b}_{i}$. Since $B^{\prime}$ is a block in level $i-1$, we get that $\operatorname{deg}\left(B^{\prime}, E^{\prime}\right) \leq m / n^{\epsilon \cdot(i-1)}$ and hence, there are at most $n^{\epsilon} i$-level blocks.

For $r=0$, the claim holds vacuously. Assume that it holds up to round $r$ and consider round $r+1$. Let $u$ be a node in layer $D^{\prime}-r+1$ and let $v_{1}, \ldots, v_{\ell}$ be its children that belongs to $B^{\prime}$ in layer $D-r$. First, consider all $v_{j}$ with at least two $i$-level blocks and such that their last block has density less than $\mathrm{b}_{i}$. By induction, we know that each such low-density block has a block of density at least $\mathrm{b}_{i}$ before it. All $v_{j}$ with a single $i$-level block in $T^{\prime}\left(v_{j}\right) \cap B^{\prime}$ are merged until there is at most one low-density block. We note that if in $T^{\prime}\left(v_{j}\right)$, there are at least two $i$-level blocks, it implies that either the block right before the block of $v_{j}$ has density of at least b , or that $v_{j} \in B^{\prime}$ and $\operatorname{deg}\left(v_{j}, E^{\prime}\right) \geq \mathrm{b}$. Finally, if the $i$-level block of $v_{\ell}$ has low-density, $u \in B^{\prime}$ and $\operatorname{deg}\left(u, E^{\prime}\right) \leq \mathrm{b}_{i}$, then the number of low-density blocks is not increased and hence the claim holds by induction assumption.

We turn to bound the round complexity of the hierarchical partitioning.
Claim 15. The hierarchical block partitioning can be computed in $O(D)$ rounds.
The algorithm for computing the hierarchical block partitioning has $O(1 / \epsilon)$ steps, where in each step $i$, we run Algorithm PartitionSingleBlock for each ( $i-1$ )-level block $B_{i-1, j}$ simultaneously. It is easy to a single application of Algorithm PartitionSingleBlock for one block $B_{i-1, j}$ in level $i-1$ takes $O(D)$ rounds. The next lemma proves Claim 15 by showing that since Algorithm PartitionSingleBlock only communicates on the tree edges $T_{B_{i-1, j}}$, we can run this algorithm for all the blocks in level $i-1$ simultaneously (at the cost of increasing the number of rounds by factor 2 ). The next lemma is quite general and will be useful later, since most of the computation that we do for blocks $B$ uses only the tree edges $T_{B}$.

Claim 16. Consider the collection of blocks in the $i$ level of the partitioning: $B_{1}, \ldots, B_{k}$ and $\mathcal{A}_{1}, \ldots, \mathcal{A}_{k}$ be a set of algorithms, each takes $O(R)$ rounds, and sends messages only on the edges of $T_{B_{j}}$ respectively. Then, all algorithms $\mathcal{A}_{1}, \ldots, \mathcal{A}_{k}$ for each of the blocks can run simultaneously in $2 R$ rounds, in the CONGEST model.

Proof. By Claim 3, each edge $e=(x, y)$ can appear in at most 2 different blocks $B_{i}, B_{j}$. Hence, the congestion when working on all blocks simultaneously is increased by factor of at most 2 when compared to working on a single block.

Tool (II): Edge Disjoint Matching inside a Block. Our next tool uses again Algorithm TreeEdgeDisjointPath from [Pel00] (see Lemma 4.3.2 within). Recall that $T_{B_{i}}=\bigcup_{u \in B_{i}} \pi\left(\ell_{i}, u\right)$ is the subtree of $T$ that connects all nodes of $B_{i}$ where $\ell$ is the LCA of the nodes in $B_{i}$. In the Block Matching Problem we are given a block $B_{i}$ in the partitioning, a subset of marked nodes $M_{i} \subseteq B_{i}$ and a spanning tree $T$. The goal is to match nodes in $M_{i}$ (but at most one) into pairs, such that the (unique) tree path in $T_{B_{i}}$ between each pair are all edge disjoint. The distributed output of the problem is that each edge on the path $P_{x, y}$ between a matching pair $\langle x, y\rangle$ for $x, y \in M_{i}$ knows the endpoints of the path, namely, $x$ and $y$, and in addition, each node on the path knows the next hops towards $x$ and $y$ respectively.

Algorithm DistBlockMatching works in $O(D)$ rounds. In the first phase which works in $D$ rounds from the leafs up, it computes the matching and in the second phase (which takes $O(D)$ rounds as well), the edges of the paths between the matched nodes are getting marked. We start by explaining the first phase. We keep the invariant that in round $D-i$, every node $u$ at layer $i$ in $T_{B_{i}}$, keeps at most one ID of a node in $M_{i}$ which is not yet matched. In every round $i \geq 0$, each node $v$ sends the ID of at most one unmatched node in $T_{B_{i}}(v)$ to its parent $u$. A node $u$ in layer $D-i+1$ that received more than one ID of marked nodes in $M_{i}$, arbitrary match all but at most one of the received nodes into pairs. In $D$ rounds, all (but at most one) nodes of $M_{i}$ get matched.

In the second phase, for every matched pair $\langle x, y\rangle$, we mark the edges on its path $P_{x, y}$. Let $v$ be the node that matched $x$ and $y$ in the first phase. We now let $v$ send this matching outcome back to its children from which it got these IDs (i.e., it sends the pair $\langle x, y\rangle$ to its children that send it $x$ or $y$ ). A node receiving this pair continues to send it to its children who sent it $x$ or $y$. This message backtracks until it gets to $x$ and $y$. It is easy to see that all this informed edges (that got an $\langle x, y\rangle$ message) are precisely the edges on the $x-y$ path in $T$. Moreover, all these computed paths are edge disjoint and the messages containing the matching result traverses only the edges belonging to the path between the matched pair. If the number of nodes in $M_{i}$ is not even, then the last unmatched node is matched to an arbitrary (already matched) node. This increases the congestion on the path edges by at most 1 . By using Claim 16 we have:

Claim 17. (I) The Block Matching problem can be solved in $O(D)$ rounds.
(II) The Block Matching problem can be solved for all blocks simultaneously in $O(D)$ rounds.

Algorithm DistNonTreeCover. As in our centralized construction, we first consider the non-tree edges $E^{\prime}=E(G) \backslash E(T)$ and construct a cycle cover $\mathcal{C}_{1}$ for these edges.

We begin with an high level description. The algorithm has $\ell=c \cdot\lceil 1 / \epsilon\rceil$ phases. During each phase, some virtual edges $\widetilde{E}$ are added to the set of edges $E^{\prime}$ that we wish to cover. For clarity, in our description we treat the virtual edges as standard edges in $G$. Informally speaking, whenever the algorithm adds a virtual edge between two nodes $u$ and $v$, it implies that the algorithm has already computed an $u-v$ path, and the virtual edge $(u, v)$ indicates the need for computing another $u-v$ path so that we will endup with a cycle. In other words, adding a virtual edge means that we deffer the closure of the cycle to the future iterations. These cycles get completed only at a point where the congestion (or overlap) between cycles can be bounded.

In phase $i$, we are given the set of edges $E^{\prime}$ in $G$ and a subset $\widetilde{E}_{i}$ of virtual edges that should be covered by cycles. The algorithm assumes at that phase that all the edges in $E^{\prime} \cup \widetilde{E}_{i}$ between nodes belonging to different $j$-level blocks for some $j \leq i-1$ are already taken care of (in a way that becomes clear later). It then restricts attention to cover the edges $(u, v)$ in $E^{\prime} \cup \widetilde{E}_{i}$ between nodes $u$ and $v$ that belong to the same block in all first ( $i-1$ )-levels of the hierarchy, but belong
to different blocks in level $i$. To handle these edges we will have $n^{2 \epsilon}$ applications of Algorithm DistBlockMatching for each pair of $i$ level blocks $B_{x}$ and $B_{y}$ which belong to the same block in level $i-1$. The subset of marked nodes in each application of Algorithm DistBlockMatching depends on the edges in $E^{\prime} \cup \widetilde{E}_{i}$ that connects these pair of blocks. Based on the output matching of Algorithm DistBlockMatching, the algorithm defines a new set of virtual edges which are internal to $i$-level blocks. After $\ell=c \cdot\lceil 1 / \epsilon\rceil$ phases, the remaining edges in $E^{\prime} \cup \widetilde{E}_{\ell}$ are between nodes that belong to the same $\ell$-level block. As will be shown in the analysis section, at that point the total number of edges $E^{\prime} \cup \widetilde{E}_{\ell}$ that have both endpoint in the same $\ell$-level block is $O(1)$. Hence, the algorithm covers these remaining edges by taking their fundamental cycles in $T$.

We explain the procedure in more details. The algorithm starts by computing $\ell=c \cdot\lceil 1 / \epsilon\rceil$ levels of block hierarchies (with respect to the input $E^{\prime}$ of non-tree edges). By letting all nodes exchange their information on their $\ell$ blocks with their neighbors, all nodes know which of the $E^{\prime}$-edges should be considered in a given phase $i$. (The knowledge about the virtual edges is acquired at the point of their creation). Initially, let $\widetilde{E}_{1}=\emptyset$ be the (empty) set of virtual edges of phase 1 .

We now focus on phase $i \geq 1$ and on one particular $(i-1)$-level block $B$ that is partitioned (in level $i$ of the hierarchy) into $k=\left\lceil n^{\epsilon}\right\rceil$ blocks $\mathcal{B}=\left\{B_{1}, \ldots, B_{k}\right\}$. We assume that we have in addition to $E^{\prime}$ a collection of virtual edges $\widetilde{E}_{i}$ and that each of the endpoints of a virtual edge in $\widetilde{E}_{i}$ knows about its second endpoint. Let $\mathcal{E}_{i}=E^{\prime} \cup \widetilde{E}_{i}$. The goal of each block $B_{x} \in \mathcal{B}$ at that phase is to take care of all its edges in $E^{\prime} \cup \widetilde{E}_{i}$ whose second endpoint is in $B_{y}$ for every $B_{y} \in \mathcal{B} \backslash B_{x}$. To take care of these $k$ different types of edges, we run (in $B_{x}$ ) $k$ applications of Algorithm DistBlockMatching simultaneously.

Consider a pair of blocks $B_{x}, B_{y} \in \mathcal{B}$, and let $E_{x, y}$ be the edges in $\mathcal{E}_{i}$ with one endpoint in $B_{x}$ and one endpoint in $B_{y}$. We divide $E_{x, y}$ into two subsets. Let $V_{x, y} \subseteq B_{x}$ be the nodes in $B_{x}$ that have at least two incident edges in $E_{x, y}$ and let $E_{x, y}^{1}$ be the subset of edges in $E_{x, y}$ incident to the nodes of $V_{x, y}$. The set of these $E_{x, y}^{1}$ edges are handled locally with a single round of communication. We let each node $u \in V_{x, y}$, compute an arbitrary matching between its $E_{x, y}$-neighbors in $B_{y}$. We then connect each of the matched pairs by a virtual edge to be added to $\widetilde{E}_{i+1}$. To do that, we simply let each $u \in V_{x, y}$ send to each of its $E_{x, y}$-neighbors in $B_{y}$ the ID of their matched node (indicating to them their new virtual edge to this node).

From now on, we restrict attention to the set of $E_{x, y}^{2}=E_{x, y} \backslash E_{x, y}^{1}$ edges, in which each node in $B_{x}$ has exactly one incident edge in $E_{x, y}$. Here we cannot decide locally about which virtual edges (between $B_{y}$ nodes) should be added. To do this, we define a subset of marked nodes $M_{x, y} \subseteq B_{x}$ as the nodes in $B_{x}$ that have exactly one incident edge in $E_{x, y}$.
Case (I): $\left|M_{x, y}\right|=1$. The algorithm covers the single edge between $B_{x}$ and $B_{y}$ in $E_{x, y}$ by taking its fundamental cycle in $T$ and add it to $\mathcal{C}_{1}$.
Case (II): $\left|M_{x, y}\right| \geq 2$. First, if $\left|M_{x, y}\right|$ is odd, the root of $T_{B_{x}}$ picks one edge ( $u, v$ ) in $E_{x, y}$ incident to node in $M_{x, y}$, add the fundamental cycle of that edge to the cycle cover $\mathcal{C}_{1}$ (i.e., by marking the cycle edges) and omit $u$ from $M_{x, y}$. From now on, assume that $\left|M_{x, y}\right|$ is even. The algorithm then applies Algorithm DistBlockMatching on $B_{x}$ with the marked nodes $M_{x, y}$. This algorithm matches the nodes of $M_{x, y}$ into pairs and mark the edge disjoint paths between these paired nodes on $T_{B_{x}}$. This pairing of nodes in $M_{x, y}$ is now used to define the set of virtual edges between nodes of $B_{y}$. In particular, let $\langle a, b\rangle$ be a matched pair in $M_{x, y}$ and let $\left(a, a^{\prime}\right),\left(b, b^{\prime}\right) \in E_{x, y}$ where $a^{\prime}, b^{\prime} \in B_{y}$. The algorithm will connect $a^{\prime}$ and $b^{\prime}$ in $B_{y}$ by a virtual edge to be added to $\widetilde{E}_{i+1}$. Recall that Algorithm DistBlockMatching computes an edge disjoint path $\pi\left(a, b, T_{B_{x}}\right)$. If $\left(a, a^{\prime}\right),\left(b, b^{\prime}\right)$ are true edges in $G$ it is easy to see how to inform $a^{\prime}$ about its virtual endpoint $b^{\prime}$
and vice versa. In such a case, we also send the information of the edges ( $a, a^{\prime}$ ) and $\left(b, b^{\prime}\right)$ on the path $W_{a^{\prime}, b^{\prime}}=\left(a^{\prime}, a\right) \circ \pi\left(a, b, T_{B_{x}}\right) \circ \pi\left(b, b^{\prime}\right)$. This way the edges on the path $P_{a^{\prime}, b^{\prime}}$ of every virtual edge ( $a^{\prime}, b^{\prime}$ ) added know the $G$-edges that this path (which will become later a cycle) covers. In case where one of the edges $\left(a, a^{\prime}\right),\left(b, b^{\prime}\right)$ is virtual, say the edge $\left(a, a^{\prime}\right)$, the algorithm sends this information using the $a-a^{\prime}$ walk $W_{a, a^{\prime}}$ computed in previous phases. These walks are shown to have length at most $O\left(2^{1 / \epsilon} \cdot D\right)$ and overlap of $O\left(n^{\epsilon}\right)$ and hence this information passing on these walks can be done efficiently of all virtual edges simultaneously using the random delay approach. These details are further explained in Appendix B. For an illustration see Figure 18.

Overall, for each $B_{x} \in \mathcal{B}$, we run simultaneously ${ }^{12} k$ applications of Algorithm DistBlockMatching with different subsets of marked nodes $M_{x, y}$ for every $B_{y} \in \mathcal{B}$. At the end of all these applications, there are new virtual edges connecting nodes that belong to the same block $B_{y}$. For each such virtual edge, we have computed a path in $G$ between its endpoints. This completes the description of phase $i \leq \ell$.

In the last phase $\ell$, we consider all edges $e=(u, v)$ in $E^{\prime} \cup \widetilde{E}_{\ell}$ such that $u$ and $v$ belong to the same $\ell$-level block. We connect these virtual edge endpoints by taking their paths in $T$. Since each virtual edge $(a, b) \in E^{\prime} \cup \widetilde{E}_{\ell}$ already has a walk $W_{u, v}$ which is shown to be of length $O\left(2^{1 / \epsilon} D\right)$, we get a cycle $P_{u, v} \circ \pi(u, v, T)$ which is added to the cycle collection. This cycle is not yet simple but by letting all edges on the cycle learn all the edges of the cycle (using the random delay approach), the cycle can be made simple locally (by internal computation at each node of the cycle) using the centralized algorithm of Section 4.1. In Claim 22 we show that making the cycle simple does not violate the covering property. Further implementation Details are provided in Appendix B. This completes the description of the main algorithm for covering non-tree edges. Let $\mathcal{C}_{1}$ be the output cycle collection.


Figure 17: Dealing with nodes with more than one endpoint in a different block. The node $u$ locally matched its neighbors in $B_{y}$ which defines a new set of virtual edges to be covered.

### 7.1.2 Analysis of Algorithm DistNonTreeCover

Congestion and Dilation Analysis. We start by showing that each edge appears on $O\left(1 / \epsilon \cdot n^{2 \epsilon}\right)$ cycles. Recall that $b_{i}(u)$ is the $i$-level block of $u$ and that $\ell=c \cdot\lceil 1 / \epsilon\rceil$. For every $i=\{1, \ldots, \ell\}$ and $u \in V$, let $\mathcal{E}_{i}(u)=\left\{(u, v) \in \mathcal{E}_{i} \mid b_{i-1}(u)=b_{i-1}(v)\right\}$ be the set of all $u$ edges in $\mathcal{E}_{i}$ that are internal

[^12]

Figure 18: Illustration of the hierarchical block partitioning and the addition of virtual edges. Top: In phase 1, we have $n^{\epsilon}$ blocks. Zoom into two blocks $B_{1}$ and $B_{2}$. The pair $\langle u, v\rangle$ got matched in $B_{1}$ and hence the virtual edge is added in $B_{2}$. Bottom: In phase 2, we consider only internal edges in 2-level blocks. Zoom into $B_{1}^{\prime}$ and $B_{2}^{\prime}$ which were part of the same block in level 1 . The matched pair $\langle x, y\rangle$ has two virtual edges to $x^{\prime}$ and $y^{\prime}$ respectively and define a new virtual edge between $x^{\prime}$ and $y^{\prime}$. This define a path $P_{x^{\prime}, y^{\prime}}$ obtained by concatenating the path $P_{x, x^{\prime}}, P_{y, y^{\prime}}$ and $\pi\left(x, y, T_{B_{1}^{\prime}}\right)$ computed in the tree of $B_{1}^{\prime}$.
to its $(i-1)$-level block. Note that Algorithm DistNonTreeCover, in phase $i$, considers only edges belonging to $\bigcup_{u} \mathcal{E}_{i}(u)$.
Claim 18. The number of edges in $\mathcal{E}_{\ell}$ that have both endpoints in a given $\ell$-level block $B$ is $O(1)$.
Proof. For every $i=\{1, \ldots, \ell\}$ and $u \in V$, let $\operatorname{deg}_{i}(u)=\left|\mathcal{E}_{i}(u)\right|$. We now claim that $\operatorname{deg}_{i}(u) \geq$ $\operatorname{deg}_{i+1}(u)$ for every $i=\{1, \ldots, \ell\}$. To see this, observe that each edge $(u, v) \in \mathcal{E}_{i}(u) \backslash \mathcal{E}_{i+1}(u)$ (i.e., in this case, $b_{i-1}(u)=b_{i-1}(v)$ but $\left.b_{i}(u) \neq b_{i}(v)\right)$ leads to the introduction of at most one virtual edge $\left(u, v^{\prime}\right) \in \mathcal{E}_{i+1}(u) \backslash \mathcal{E}_{i}(u)$ where $b_{i}(u)=b_{i}\left(v^{\prime}\right)$. In addition, all the virtual edges added to $\mathcal{E}_{i+1}(u) \backslash \mathcal{E}_{i}(u)$ are due to edges in $\mathcal{E}_{i}(u) \backslash \mathcal{E}_{i+1}(u)$. Combining these two observations, we get that $\operatorname{deg}_{i+1}(u) \leq \operatorname{deg}_{i}(u)$.

We therefore have that for every $u \in V, \operatorname{deg}_{1}(u) \geq \operatorname{deg}_{\ell}(u)$. By definition

$$
\mathcal{E}_{1}(u)=\left\{(u, v) \mid(u, v) \in E^{\prime}\right\}
$$

and hence $\operatorname{deg}_{1}(u)=\operatorname{deg}\left(u, E^{\prime}\right)$. Let $B^{\prime}$ be an $(\ell-1)$ level block such that $B \subseteq B^{\prime}$. By the construction of the $(\ell-1)$ level block $B^{\prime}$, we get that $\sum_{u \in B^{\prime}} \operatorname{deg}_{\ell}(u) \leq \mathrm{b}_{\ell-1}=O(1)$. In particular, since $B \subseteq B^{\prime}$, it also holds that $\sum_{u \in B} \operatorname{deg}_{\ell}(u)=O(1)$, since $\operatorname{deg}_{\ell}(u)$ is the number of $u$ edges in $\mathcal{E}_{\ell}$ whose second endpoint is in the same $(\ell-1)$-level block, it also holds that number of $u$ edges that have both endpoints in its $\ell$-level block is bounded by $O(1)$.

Let $\mathcal{C}_{i}^{\prime}$ be the collection of fundamental cycles added at phase $i$. Recall that we cover an edge $e \in \mathcal{E}_{i}$ by its fundamental cycle in $T$ if the edge $e$ is the only edge in that set the connects two brother blocks $B_{x}$ and $B_{y}$.

Claim 19. Every edge appears on at most $O\left(n^{2 \epsilon}\right)$ cycles in $\mathcal{C}_{i}^{\prime}$.
Proof. Let $B_{1}, \ldots, B_{k}$ be the collection of all $(i-1)$-level blocks. We now consider a particular block $B_{j}$ which is partitioned into $B_{j, 1}, \ldots, B_{j, \ell}$ blocks in level $i$ where $\ell=O\left(n^{\epsilon}\right)$. For such $B_{j}$ we add at most $O\left(\ell^{2}\right)$ fundamental cycles to $\mathcal{C}_{i}^{\prime}$ (one per pair of blocks in $B_{j, 1}, \ldots, B_{j, \ell}$ ). Note that all these cycles are edges in $T_{B_{j}}$. By Claim 3, every edge $e$ appears on at the trees of at most two ( $i-1$ )-blocks, hence overall each edge appears in $O\left(n^{2 \epsilon}\right)$ cycles.

We now turn to consider cycles that are computed using the addition of the virtual edges.
Claim 20. For each virtual edge $(u, v)$ added in phase $i$, the algorithm computes a $u-v$ walk $W_{u, v}$ of length at most $2^{i} \cdot D$. In addition, this walk "covers" $O\left(2^{i}\right)$ edges in $E^{\prime}$.

Proof. Both claims are shown by induction on $i$. For $i=1$, when adding a virtual edge $(u, v)$, we have two edges $\left(u, u^{\prime}\right)$ and $\left(v, v^{\prime}\right)$ in $E^{\prime}$ such that $u, v$ (resp., $u^{\prime}, v^{\prime}$ ) are in the same 1-level block. Using Algorithm DistBlockMatching, we have marked the tree path between $u^{\prime}$ and $v^{\prime}$ and hence we have a path $W_{u, v}=\left(u, u^{\prime}\right) \circ \pi\left(u^{\prime}, v^{\prime}, T\right) \circ\left(v^{\prime}, v\right)$. Clearly, $\left|W_{u, v}\right| \leq 2 D+2$. This path covers exactly two edges $\left(u, u^{\prime}\right)$ and $\left(v, v^{\prime}\right)$ which proves the claims for the induction base.

Assume that the claims hold up to phase $i-1$ and consider phase $i$. Here, when adding a virtual edge $(u, v)$ in phase $i$, same story holds as in the induction phase only that the two edges $\left(u, u^{\prime}\right)$ and $\left(v, v^{\prime}\right)$ might be virtual edges that were added in previous phases. The $u-v$ walk is given by $W_{u, v}=W_{u, u^{\prime}} \circ \pi\left(u^{\prime}, v^{\prime}, T\right) \circ W_{v^{\prime}, v}$ where $W_{u, u^{\prime}}, W_{v^{\prime}, v}$ is either an edge in $G$ or a walk computed in previous phases. Using the induction assumption, we get that the length of the walk $W_{u, v}$ at most $2 \cdot 2^{i-1} \cdot D+2 D \leq 2^{i} \cdot D$. In addition, if $\left(u, u^{\prime}\right)$ is a virtual edge, then by induction assumption, $W_{u, u^{\prime}}$ covers at most $2^{i-1}$ edges in $E^{\prime}$ and same goes to $W_{v, v^{\prime}}$. If $\left(u, u^{\prime}\right)$ is an edge in $E^{\prime}$ than it covers only itself (same for $\left(v, v^{\prime}\right)$ ). Overall the walk $W_{u, v}$ covers $2^{i}$ edges. The claim follows.

Let $\mathcal{W}_{i}=\left\{W_{u, v} \mid(u, v) \in \mathcal{E}_{i+1} \backslash \mathcal{E}_{i}\right\}$ be the collection of all $u-v$ walks between all virtual edges $(u, v)$ added in phase $i$. Let $\mathcal{W}=\bigcup_{i=1}^{\ell} \mathcal{W}_{i}$ be the collection of all these walks.

Claim 21. For every $i$, each edge $e \in G$ appears at most $O\left(i \cdot n^{\epsilon}\right)$ in total on all walks of $\mathcal{W}_{i}$ (this takes into account the multiple appearance of the edge on the same walk). Hence, overall, each edge e appears $O\left((1 / \epsilon)^{2} \cdot n^{\epsilon}\right)$ time on all walks of $\mathcal{W}$.

Proof. Recall that we add a virtual edge $(u, v)$ in phase $i \geq 1$ for $u, v \in B_{y}$, in case where there are edges $\left(u, u^{\prime}\right),\left(v, v^{\prime}\right)$ in $\mathcal{E}_{i}$ such that $u^{\prime}, v^{\prime}$ are in $B_{x}$ and $B_{x}$ and $B_{y}$ are brothers (i.e., belong to the same block in level $i-1)$. Note that for $i \geq 2$, the edges $\left(u, u^{\prime}\right),\left(v, v^{\prime}\right)$ might be virtual. In any case, we have have $u-v$ walk $W_{u, v}=W_{u, u^{\prime}} \circ \pi\left(u^{\prime}, v^{\prime}, T\right) \circ W_{v^{\prime}, v}$ where in phase $1, W_{u, u^{\prime}}, W_{v, v^{\prime}}$ are simply the edges $\left(u, u^{\prime}\right)$ and $\left(v, v^{\prime}\right)$.

We prove the claim by induction on $i$. For the base of the induction, consider $i=1$. In this case, all walks $W_{u, v}$ for the virtual edges added at phase 1 are in fact paths. Thus, we only need to bound the number of an edge $e$ on different such paths. Since we have $n^{\epsilon}$ applications of Algorithm PartitionSingleBlock in each 1-level block, and using Claim 16, we get that each edge $e$ appears on $O\left(n^{\epsilon}\right)$ different paths in $\mathcal{W}_{1}$.

Assume that the claim holds up to phase $i-1$ and consider phase $i$. Here, we can apply the induction assumption for $W_{u, u^{\prime}}$ and $W_{v^{\prime}, v}$ (it is sufficient to consider the case where $\left(u, u^{\prime}\right),\left(v, v^{\prime}\right)$ are virtual edges added in previous phases, as this is the interesting case). The walk $W_{u, v}$ has two types of segments: (I) walks connecting virtual edges that were added in phase $j \leq i-1$, namely,
$W_{u, u^{\prime}}, W_{v, v^{\prime}}$ and (II) tree segment $\pi(u, v)$. By the same reasoning as in the base of the induction, each edge $e$ can appear on at most $O\left(n^{\epsilon}\right)$ tree paths $\pi(u, v)$ for every virtual edge ( $u, v$ ) added in phase $i$. By the induction assumption, each edge $e$ appears at most $(i-1) n^{\epsilon}$ times on the collection of walks in $\mathcal{W}_{j}$ for $j \leq i-1$. Hence, overall an edge $e$ appears $i \cdot n^{\epsilon}$ many time on the walks of $\mathcal{W}_{i}$. The claim holds.

We therefore have:
Corollary 5. Every edge appears on $O\left(n^{2 \epsilon}\right)$ cycles in the final cycle collection. Each cycle has length $O\left(2^{1 / \epsilon} \cdot D\right)$ and covers $O\left(2^{1 / \epsilon}\right)$ many non-tree edges.

Covering Analysis of Algorithm DistNonTreeCover. Note that the cycles computed are not necessarily simple. We next claim that they can be made simple without loosing the covering property. Recall that in each phase $i$, for every virtual edge ( $u, v$ ) added in phase $i$ we compute a walk $W_{u, v}$ in $G$.

Claim 22. For every virtual edge $(u, v)$, any edge that appears more than once on the walk $W_{u, v}$ is a T-edge.

Proof. The proof is shown by induction on the phase number $i$. For $i=1$, there exists two edges $\left(u, u^{\prime}\right)$ and $\left(v, v^{\prime}\right)$ in $E^{\prime}$ such that $u, v$ (resp., $u^{\prime}, v^{\prime}$ ) are in the same 1-level block and $u^{\prime}, v^{\prime}$ have been paired in their block by Algorithm DistBlockMatching. In this case, $W_{u, v}=\left(u, u^{\prime}\right) \circ \pi\left(u^{\prime}, v^{\prime}, T\right) \circ$ $\left(v^{\prime}, v\right)$ which is in fact a path!.

Assume that the claim holds up to phase $i-1$ and consider phase $i$. For any virtual edge $(u, v)$ added in phase $i$, there exists two edges $\left(u, u^{\prime}\right)$ and $\left(v, v^{\prime}\right)$ in $\mathcal{E}$ such that $u, v$ (resp., $\left.u^{\prime}, v^{\prime}\right)$ are in the same $i$-level block and $u^{\prime}, v^{\prime}$ have been paired in their block by Algorithm DistBlockMatching. The only difference to the induction base is that now the edges $\left(u, u^{\prime}\right)$ and $\left(v, v^{\prime}\right)$ might be virtual edges added in some phase $j \leq i-1$. In either case $W_{u, u^{\prime}}$ is either an edge ( $u, u^{\prime}$ ) are of walk for which the induction assumption can be applied. The walk $W_{u, v}=W_{u, u^{\prime}} \circ \pi\left(u^{\prime}, v^{\prime}, T\right) \circ W_{v^{\prime}, v}$ adds a tree segment $\pi\left(u^{\prime}, v^{\prime}, T\right)$ to the existing walks $W_{u, u^{\prime}}, W_{v^{\prime}, v}$. Hence, it can only increase the appearance of the tree edges. Combining with the induction assumption, the claim holds.

Claim 23. Every edge $e^{\prime} \in E^{\prime}$ is covered by the cycles of Algorithm DistNonTreeCover.
Proof. By Claim 22, it remains to show that for every edge $e^{\prime} \in E^{\prime}$, there is a walk $W_{u, v}$ such that $e^{\prime}$ appears on it. Let $e^{\prime}=(x, y)$ and define $i$ to be the first index such that $x$ and $y$ belong to different $i$-level block. The edge $e^{\prime}$ is then considered in phase $i$ of the algorithm. If the algorithm did not cover $e^{\prime}$ by taking its fundamental cycle, it implies that $i$ is not the last phase and necessarily some virtual edge was added due to $e^{\prime}=(w, z)$. By construction, $e^{\prime}$ connects two nodes in the same $i$-level block and hence will be handled in phase $j \geq i+1$ and in addition, $(x, y)$ appears on the walk $W_{w, z}$.

We can continue with the argument with the virtual edge ( $w, z$ ) which is handled only at later phase $j$, claiming that the edge $(x, y)$ appears when another virtual edge $\left(w^{\prime}, z^{\prime}\right)$ is added when considering $(w, z)$ and hence $(x, y)$ appears on the walk $W_{w^{\prime}, z^{\prime}}$. This continues until we get to a point where the fundamental cycle of some virtual edge $(a, b)$ such that $(x, y) \in W_{a, b}$ is added and at that point, Claim 22 guarantees that when making the cycle $W_{a, b} \circ \pi(a, b, T)$ simple the edge $(x, y)$ is covered. Recall that indeed in the last phase $\ell$ we handle all remaining edges (also the virtual onces) by taking their fundamental cycles. The claim follows.

## Round Complexity of Algorithm DistNonTreeCover.

Claim 24. Algorithm DistTreeCover has round complexity $\widetilde{O}\left(1 / \epsilon \cdot\left(2^{1 / \epsilon} D+n^{2 \epsilon}\right)\right)$
Proof. By Claim 15, the hierarchical block partitioning can be computed in $O(d)$ rounds. The algorithm has $O(1 / \epsilon)$ phases. We show that each phase can be implemented can in $O\left(2^{1 / \epsilon} D+n^{\epsilon}\right)$ rounds. We have $O\left(n^{\epsilon}\right)$ applications of Algorithm DistBlockMatching in each block $B_{x}$. Using the random delay approach and Claim 17 all the $O\left(n^{\epsilon}\right)$ applications in a given block $B_{x}$ can be done in $\widetilde{O}\left(n^{\epsilon}+D\right)$ rounds. Using Claim 16, this can be done for all blocks in $\widetilde{O}\left(n^{\epsilon}+D\right)$ rounds as well. We next bound the number of rounds required to form the virtual edges $\left(a^{\prime}, b^{\prime}\right)$ and exchange information about the $G$ edges that are covered by the $W_{a^{\prime}, b^{\prime}}$ path. By Claims 20 and 21, using the random delay approach it can be done in $\widetilde{O}\left(n^{2 \epsilon}+2^{1 / \epsilon} D\right)$ rounds. Finally, once the cycles are computed, using the random delay approach again, all edges can learn all the cycles that go through it in $\widetilde{O}\left(n^{2 \epsilon}+2^{1 / \epsilon} D\right)$ round and locally make them simple.

### 7.1.3 Distributed Algorithm for Covering Tree Edges

We turn to consider the remaining tree edges $E(T)$. Algorithm DistTreeCover essentially mimics the centralized construction of Section 4.1. Recall that $p(v)$ is the parent of $v$ in the BFS tree $T$. A non-tree edge $e^{\prime}=\left(u^{\prime}, v^{\prime}\right)$ is a swap edge for the tree edge $e=(p(v), v)$ if $e \in \pi\left(u^{\prime}, v^{\prime}\right)$, let $s(v)=v^{\prime}$ by the endpoint of $e^{\prime}$ that is not in $T(v)$. By using the algorithm of Section 4.1 in [GP16], we can make every node $v$ know $s(v)$ in $O(D)$ rounds.

A key part in the algorithm of Section 4.1 is the definition of the path $P_{e}=\pi\left(v, u^{\prime}\right) \circ\left(u^{\prime}, s(v)\right)$ for every tree edge $e=(p(v), v)$. By computing swap edges using Section 4.1 in [GP16] all the edges of each $P_{e}$ get marked.
Computing the set $I(T) \subseteq E(T)$. We next describe how to compute a maximal collection of tree edges $I=\left\{e_{i}\right\}$ whose paths $P_{e_{i}}$ are edge disjoint and in addition for each edge $e_{j} \in E(T) \backslash I$ there exists an edge $e_{i} \in T^{\prime}$ such that $e_{j} \in P_{e_{i}}$. To achieve this, we start working on the root towards the leaf. In every round $i \in\{1, \ldots, D\}$, we consider only active edges in layer $i$ in $T$. Initially, all edges are active. An edge becomes inactive in a given round if it receives an inactivation message in any previous round. Each active edge in layer $i$, say $e_{j}$, initiates an inactivation message on its path $P_{e_{j}}$. An inactivation message of an edge $e_{j}$ propagates on the path $P_{e_{j}}$ round by round, making all the corresponding edges on it to become inactive.

Note that the paths $P_{e_{j}}$ and $P_{e^{\prime}}$ for two edges $e_{j}$ and $e_{j^{\prime}}$ in the same layer of the BFS tree, are edge disjoint and hence inactivation messages from different edges on the same layer do not interfere each other. We get that an edge in layer $i$ active in round $i$ only if it did not receive any prior inactivation message from any of its BFS ancestors. In addition, any edge that receives an inactivation message necessarily appears on a path of an active edge. It is easy to see that within $D$ rounds, all active edges $I$ on $T$ satisfy the desired properties (i.e., their $P_{e_{i}}$ paths cover the remaining $T$ edges and these paths are edge disjoint).

Distributed Implementation of Algorithm TreeCover. First, we mark all the edges on the $P_{e}$ paths for every $e \in I(T)$. As every node $v$ with $e=(p(v), v)$ know its swap edge, it can send information along $P_{e}$ and mark the edges on the path. Since each edge appears on the most two $P_{e}$ paths, this can be done simultaneously for all $e \in I(T)$.

From this point on we follow the steps of Algorithm TreeCover. The partitioning of Appendix A can be done in $O(D)$ rounds as it only required nodes to count the number of nodes in their subtree.

We define the ID of each tree $T_{1}^{\prime}, T_{2}^{\prime}$ to be the maximum edge ID in the tree (as the trees are edge disjoint, this is indeed an identifier for the tree). By passing information on the $P_{e}$ paths, each node $v$ can learn the tree ID of its swap endpoint $s(v)$. This allows to partition the edges of $T^{\prime}$ into $E_{x, y}^{\prime}$ for $x, y \in\{1,2\}$. Consider now the $i^{t h}$ phase in the computation of cycle cover $\mathcal{C}_{1,2}$ for the edges $E_{1,2}^{\prime}$.

Applying Algorithm TreeEdgeDisjointPath can be done in $O(D)$ round. At the end, each node $v_{j}$ knows its matched pair $v_{j}^{\prime}$ and the edges on the tree path $\pi\left(v_{j}, v_{j}^{\prime}, T_{1}^{\prime}\right)$ are marked. Let $\Sigma$ be the matched pairs. We now the virtual conflict graph $G_{\Sigma}$. Each pair $\left\langle v_{j}, v_{j}^{\prime}\right\rangle \in \Sigma$ is simulated by the node of higher ID, say, $v_{j}$. We say that $v_{j}$ is the leader of the pair $\left\langle v_{j}, v_{j}^{\prime}\right\rangle \in \Sigma$. Next, each node $v$ that got matched with $v^{\prime}$ activates the edges on its path $P_{e} \cap E\left(T_{1}^{\prime}\right)$ for $e=(p(v), v)$. Since the $\pi$ edges of the matched pairs are marked as well, every edge $e^{\prime} \in \pi\left(v_{k}, v_{k}^{\prime}, T_{1}^{\prime}\right)$ that belongs to an active path $P_{e}$ sends the ID of the edge $e$ to the leader of the pair $\left\langle v_{k}, v_{k}^{\prime}\right\rangle$. By Claim 6, every pair $\sigma^{\prime}$ interferes with at most one other pair and hence there is no congestion and a single message is sent along the edge-disjoint paths $\pi\left(v_{j}, v_{j}^{\prime}, T_{1}^{\prime}\right)$ for every $\left\langle v_{j}, v_{j}^{\prime}\right\rangle \in \Sigma$. Overall, we get the the construction of the virtual graph can be done in $O(D)$ rounds.

We next claim that all leaders of two neighboring pairs $\sigma, \sigma^{\prime} \in G_{\Sigma}$ can exchange $O(\log n)$ bits of information using $O(D)$ rounds. Hence, any $r$-round algorithm for the graph $G_{\Sigma}$ can be simulated in $T_{1}^{\prime}$ in $O(r \cdot D)$ rounds. To see this, consider two neighbors $\sigma=\langle x, y\rangle, \sigma^{\prime}=\left\langle x^{\prime}, y^{\prime}\right\rangle$ where $\sigma^{\prime}$ interferes $\sigma$. Without loss of generality, assume that the leader $x^{\prime}$ of $\sigma^{\prime}$ wants to send a message to the leader $x$ of $\sigma$. First, $x^{\prime}$ sends the message on the path $\pi\left(x^{\prime}, y^{\prime}, T_{1}^{\prime}\right)$. The edge $e^{\prime} \in \pi\left(x^{\prime}, y^{\prime}, T_{1}^{\prime}\right) \cap P_{e}$ for $e=(p(x), x)$ that receives this message sends it to the leader $x$ along the path $P_{e}$. Since we only send messages along edge disjoint paths, there is no congestion and can be done in $O(D)$ rounds.

Since the graph $G_{\Sigma}$ has arboricity $O(1)$, it can be colored with $O(1)$ colors and $O(\log n)$ rounds using the algorithm of [BE10]. By the above, simulating this algorithm in $G$ takes $O(D \log n)$ rounds. We then consider each color class at a time where at step $j$ we consider $\Sigma_{i, j}$. For every $\sigma=\langle x, y\rangle, x$ sends the ID of $s(y)$ to $s(x)$ along the $P_{e}$ path for $e=(p(x), x)$. In the same manner, $y$ sends the ID of $s(x)$ to $s(y)$. This allows each node in $T_{2}^{\prime}$ know its virtual edge. At that point we run Algorithm DistNonTreeCover to cover the virtual edges. Each virtual edge is later replaced with a true path in $G$ in a straightforward manner.

## Analysis of Algorithm DistTreeCover.

Claim 25. Algorithm DistTreeCover computes a $\left(2^{1 / \epsilon} D, n^{\epsilon} \log ^{2} n\right)$ cycle cover $\mathcal{C}_{2}$ for the tree edges $E(T)$ and has round complexity of $O\left(2^{1 / \epsilon} D \cdot n^{\epsilon} \cdot \log ^{2} n\right)$.

Proof. The correctness follows the same line of arguments as in the centralized construction (see the Analysis of Section 4.1.2), only the here we use Algorithm DistNonTreeCover. Each cycle computed by Algorithm DistNonTreeCover has length $O\left(2^{1 / \epsilon} D\right)$ and the cycle covers $O\left(2^{1 / \epsilon}\right)$ non-tree edges. In our case, each non-tree edge is virtual and replaced by a path of length $O(D)$ hence the final cycle has still length $O\left(2^{1 / \epsilon} D\right)$. With respect to congestion, we have $O(\log n)$ levels of recursion and in each level when working on the subtree $T^{\prime}$ we have $O(\log n)$ applications of Algorithm DistNonTreeCover which computes cycles with congestion $O\left(n^{\epsilon}\right)$. The total congestion is then bounded by $O\left(n^{\epsilon} \cdot \log ^{2} n\right)$.

We proceed with round complexity. The algorithm has $O(\log n)$ levels of recursion. In each level we work on edge disjoint trees simultaneously. Consider a tree $T^{\prime}$. The partitioning into $T_{1}^{\prime}, T_{2}^{\prime}$ takes $O(D)$ rounds. We now have $O(\log n)$ phases. We show that each phase takes $O\left(2^{1 / \epsilon} D \cdot n^{\epsilon}\right)$ rounds, which is the round complexity of Algorithm DistNonTreeCover. In particular, In phase $i$
we have the following procedures. Applying Algorithm TreeEdgeDisjointPath in $T_{1}^{\prime}, T_{2}^{\prime}$ takes $O(D)$ rounds. The computation of the conflict graph $G_{\Sigma}$ takes $O(D)$ rounds as well and coloring it using the coloring algorithm for low-arboricity graphs of [BE10] takes $O(D \log n)$ rounds. Then we apply Algorithm DistNonTreeCover which takes $\widetilde{O}\left(2^{1 / \epsilon} \cdot D+n^{2 \epsilon}\right)$ rounds. Translating the cycles into cycles in $G$ takes $\widetilde{O}\left(2^{1 / \epsilon} \cdot D+n^{2 \epsilon}\right)$ rounds. Overall, we have $\widetilde{O}\left(2^{1 / \epsilon} \cdot D+n^{2 \epsilon}\right)$ rounds.

Theorem 3 follows by combining Lemma 8 and Claims 24 and 25.

### 7.2 Additional Low Congestion Covers

Private Neighborhood Cover. We now show how to use the distributed construction of cycle covers to construct private neighborhood covers. The distribute output format of private neighborhood cover $\mathcal{N}$ is that each node $u$ knows its parent in the spanning tree $T(v) \in \mathcal{N}$ for every $v \in V$.

Lemma 9. There exists an $\widetilde{O}\left(4^{1 / \epsilon} \cdot \Delta \cdot D+n^{\epsilon} \cdot D\right)$-round algorithm that computes an $\left(4^{1 / \epsilon} \cdot D \cdot \Delta\right.$. $\left.\log \Delta, n^{\epsilon} \cdot D \cdot \log ^{2} n \cdot \log \Delta\right)$ private neighborhood cover.

Proof. We start by running Algorithm DistCycleCover which computes a ( $\mathrm{d}, \mathrm{c}$ ) cycle cover $\mathcal{C}$ for $\mathrm{d}=O\left(4^{1 / \epsilon} \cdot D\right)$ and $\mathrm{c}=O\left(n^{\epsilon} \cdot \log ^{2} n\right)$. By using $\widetilde{O}(\mathrm{~d}+\mathrm{c})$ rounds, we can make each edge $(u, v)$ know the edges of all the cycles it belongs to in $\mathcal{C}$. We mimic the centralized reduction to cycle cover. In this reduction, we have $O(\log \Delta)$ applications of Algorithm DistCycleCover on some virtual graph. Since a node $v$ knows the cycles of its edges, it knows which virtual edges it should add in phase $i$. Simulating the virtual graph can be done with no extra congestion in $G$. In each phase $i$, we compute ( $\mathrm{d}, \mathrm{c}$ ) cycle cover $\mathcal{C}_{i}$ in $G$ obtained by computing the cycle cover in a virtual graph. By the same argument as in Claim 11, translating these cycles to cycles in $G$ does not increase the congestion. Using $\widetilde{O}(\mathrm{~d}+\mathrm{c})$ rounds, each edge $e$ can learn all the edges on the cycles that pass through it appears in $\mathcal{C}_{i}$. At the last phase $\ell=O(\log \Delta)$, the graph $G_{\ell}\left(u_{j}\right)$ consists of $O(\log \Delta \cdot \Delta)$ cycles. In particular,

$$
G_{\ell}\left(u_{j}\right)=\bigcup_{i=1}^{\ell}\left\{C \in \mathcal{C}_{i} \mid \quad\left(u_{j}, v\right) \in C, v \in \Gamma\left(u_{j}, v\right)\right\}
$$

By the same argument of Claim 11, each edge $e$ appears on $O\left(\log \Delta \cdot \log ^{2} n \cdot n^{\epsilon} \cdot D\right)$ different subgraphs $G_{\ell}\left(u_{j}\right)$ for $u_{j} \in V$. The diameter of each subgraph can $G_{\ell}\left(u_{j}\right)$ can be clearly bounded by the number of nodes it contained which is $O\left(4^{1 / \epsilon} \cdot \log \Delta \cdot \Delta \cdot D\right)$. Since each edge $e$ knows all cycles it appears on ${ }^{13}$, it also knows all the graphs $G_{\ell}\left(u_{j}\right)$ to which it belongs. Computing a spanning tree in $G_{\ell}\left(u_{i}\right) \backslash\left\{u_{i}\right\}$ can be done in $\widetilde{O}\left(4^{1 / \epsilon} \Delta \cdot D\right)$ rounds. Using the standard random delay approach, and using the fact that each edge appears on $O\left(\log \Delta \cdot \log ^{2} n \cdot n^{\epsilon} \cdot D\right)$, all the spanning trees in $G_{\ell}\left(u_{j}\right) \backslash\left\{u_{j}\right\}$ can be construct simultaneously in $\widetilde{O}\left(4^{1 / \epsilon} \cdot \Delta \cdot D+n^{\epsilon} \cdot D\right)$ rounds.
$P_{2}$ Cycle Cover Recall that in $P_{2}$ cycle cover $\mathcal{C}$, for every two-edge path $P=[x-y-z]$, there is a cycle $C \in \mathcal{C}$ that contains $P$. We now show how to construct low-congestion $P_{2}$ cycle cover using private neighborhood cover. The distributed output of $P_{2}$ cycle cover is that each edge $e \in C$ knows that identifier of all the two-edge paths that are covered by the cycles that pass through $e$.

[^13]The cycles that cover of all paths $P=[x-y-z]$ for every $x, z \in \Gamma(y, G)$ are all contained in the tree $T(y) \in \mathcal{N}$ where $\mathcal{N}$ is the private neighborhood cover. As shown in Lemma 9, every edge $e \in \Gamma(y, G)$ knows all the graphs $y$ to which it belongs. We first explain how to mark the collection of $\Delta$ paths between each $x$ and $z$ for every $x, z \in \Gamma(y, G)$. Since the tree $T(y)$ contains $\widetilde{O}\left(4^{1 / \epsilon} \cdot D \cdot \Delta\right)$ nodes, by standard pipeline procedure, all nodes in $T(y)$ can learn all the edges of the tree in $O(|T(y)|)$ rounds. This allows every edge $e$ that belongs to an $x-z$ path in $T(y)$ mark itself on the cycle that covers $[x-y-z]$.

Since each edge $e$ appears on $\widetilde{O}\left(D \cdot n^{\epsilon}\right)$ different trees $T\left(u_{j}\right)$ for $u_{j} \in V$, we can work on all trees simultaneously using the random delay approach. Overall the $P_{2}$ cover is computed in $\widetilde{O}\left(4^{1 / \epsilon} \cdot D \cdot \Delta+D \cdot n^{\epsilon}\right)$ rounds. We have:

Lemma 10. An $\left(4^{1 / \epsilon} \cdot D \cdot \Delta \cdot \log \Delta, D \cdot \Delta^{2} \cdot n^{\epsilon} \cdot \log ^{2} n \cdot \log \Delta\right) P_{2}$ cycle cover $\mathcal{C}$ can be computed in $\widetilde{O}\left(4^{1 / \epsilon} \cdot D \cdot \Delta+D \cdot n^{\epsilon}\right)$ rounds.

By combining our general compiler Theorem 6 with our distributed low-congestion covers of Lemmas 9 and 10, we get:

Corollary 6 (Distributed PSM with no pre-processing). Let $G$ be a 2-vertex connected n-vertex graph with diameter $D$ and maximal degree $\Delta$. Let $\mathcal{A}$ be a natural distributed algorithm that runs on $G$ in $r$ rounds. Then, $\mathcal{A}$ can be transformed to an equivalent algorithm $\mathcal{A}^{\prime}$ with perfect privacy which runs in $\widetilde{O}\left(r \cdot 2^{\sqrt{\log n}} \cdot D \cdot \operatorname{poly}(\Delta)\right)$

### 7.3 Pre-processing for Improved Cover Structures

We next show how the nearly optimal covers of Theorems 1,4 and 5 can be constructed using $\widetilde{O}(n+D \cdot \Delta)$ rounds in the distributed setting. This pre-processing step should be done only once and any general compiler that will run in the network in the future will be able to use it (i.e., these structures should be computed once for a given network).

Lemma 11. For every 2 -edge connected $n$-vertex graph a $\left(D \log n, \log ^{3} n\right)$ cycle cover can be computed distributively in $\widetilde{O}(n)$ rounds of pre-processing.

Proof. Compute a BFS tree $T$ and consider the set of non-tree edges $E^{\prime}$. Let $E_{0}=E^{\prime}$. So long that number of edges $E_{i}$ to be covered in $E^{\prime}$ is at least $O\left(\log ^{c} n \cdot n\right)$, we do as follows in phase $i$.

Let $\Delta_{i}=\left|E_{i}\right| / n$. We partition the edges of $E_{i}$ into $\ell_{i}=\Delta_{i} /(c \cdot \log n)$ edge-disjoint subgraphs by letting each edge in $E_{i}$ pick a number in $\left[1, \ell_{i}\right]$ uniformly at random. We have that w.h.p. each subgraph $E_{i, j}$ contains $\Theta(n \log n)$ edges of $E_{i}$.

At the point, we work on each subgraph $E_{i, j}$ independently. We compute a BFS tree $T_{i, j}$ in each $E_{i, j}$ (using only communication on $E_{i, j}$ edges). We then collect all edges of $E_{i, j}$ to the root by pipelining these edges on $T_{i, j}$. At that point, each root of $T_{i, j}$ can partition all but $2 n$ edges of $E_{i, j}$ into edge disjoint cycles of length $O(\log n)$. The root also pass these cycle information to the relevant edges using the communication on $T_{i, j}$. Note that since the $E_{i, j}$ subgraphs are disjoint, this can be done simultaneously for all subgraphs $E_{i, j}$. At the end of that phase, we are left with $2 n \cdot \ell_{i}=O\left(\left|E_{i}\right| / \log n\right)$ uncovered edges $E_{i+1}$ to be handled in the next phase. Overall, after $O(\log n / \log \log n)$ phases, we are left with $O(n \log n)$ uncovered edges. At the point, we can pipeline these edges to the root of the BFS tree, along with the $n-1$ edges of the BFS tree and let the root compute it locally as explained in Section 4.1. The lemma follows.

Combining that with Lemmas 9 and 10, we have:
Corollary 7. For every 2 -vertex connected n-vertex graph $G=(V, E)$ with diameter $D$ and maximum degree $\Delta$ : (I) an $O\left(D \cdot \Delta \cdot \log n, D \cdot \log ^{3} n \cdot \log \Delta\right)$ private neighborhood cover $\mathcal{N}$ can be constructed using $\widetilde{O}(n+D \cdot \Delta)$ rounds.
(II) an $O\left(D \cdot \Delta \cdot \log n \cdot \log \Delta, D \cdot \Delta^{2} \cdot \log ^{3} n \cdot \log \Delta\right) P_{2}$ cycle cover $\mathcal{C}$ can be constructed in $\widetilde{O}(n+D \cdot \Delta)$ rounds.

Distributed Construction of Instance-Optimal Covers. As discussed in Section 5, it is desirable to obtain cover constructions with bounds that are competitive with the best possible covers for the given graph. Recall that Section 5 (specifically, Lemma 4) provides a recipe for transforming an algorithm for $\left(f(D)\right.$, c) cycle cover, to an algorithm that constructs $\left(f\left(\widetilde{O}\left(\mathrm{OPT}_{G}\right)\right), \widetilde{O}(\mathrm{c})\right)$ cycle cover where $\mathrm{OPT}_{G}$ is the best dilation possible. Also, recall that $\mathrm{OPT}_{G}=\alpha_{G}$ where $\alpha_{G}$ is the maximum over all edges $e \in G$ of the shortest cycle containing an edge $e$ in $G$.

To apply this transformation in the distributed setting, we first observe that Alg. DistCycleCover is nice, that is, it covers every edge $e \in G$ for which there exists some cycle containing it in $G$ (does not require $G$ to be bridgeless). Now, by using the network decomposition construction of Appendix C, and applying Alg. DistCycleCover in each cluster of the decomposition, we get an $\left(\widetilde{O}\left(4^{1 / \epsilon} \cdot \mathrm{OPT}_{G}\right), \widetilde{O}\left(n^{\epsilon}\right)\right)$ cycle cover. Since each vertex belongs to $\widetilde{O}(1)$ clusters, this increases the round complexity of Algorithm DistCycleCover by a logarithmic factor. The distributed construction of private neighborhood cover and $P_{2}$ cover can be obtained by replacing Algorithm DistCycleCover with the approximation distributed algorithm for cycle cover, describe above.

## 8 Discussion and Future Work

In this paper we introduce a new framework for secure distributed graph algorithms. We present the construction of a general compiler that can turn any natural non-secure distributed algorithm to a secure one, while increasing the round complexity by factor $\widetilde{O}(D \cdot \Delta)$. There are many remaining interesting research directions.

First, we note that our protocols are secure against an adversary that controls a single node in the graph. To see this, imagine our secret sharing approach in which a node $w \in \Gamma_{2}(v, G)$ sends a secret to $v$ along two vertex disjoint paths $P, P^{\prime}$ (which are given by the $P_{2}$ cycle cover). If the adversary takes over two nodes, one on each path $P, P^{\prime}$, the communication between $w$ and $v$ is no longer secure. This example also suggests that to provide resilience against coalition of $k$ nodes, the graph is required to be $(k+1)$ vertex connected and in addition, the low-congestion covers should be adopted accordingly. For instance, such $k$-resilience calls for private neighborhood covers in which every neighborhood is $k$-connected, and $P_{2}$ cycle covers in which each $P_{2}$ is covered by $k$ (almost vertex disjoint) cycles, etc.

An additional research direction involves the distribution computation of $(\widetilde{O}(D), \widetilde{O}(1))$ cycle covers in sublinear number of rounds. Our current efficient distributed algorithm obtains ( $\widetilde{O}\left(4^{1 / \epsilon}\right.$. $\left.D), \widetilde{O}\left(n^{\epsilon}\right)\right)$ cycle covers in $O\left(4^{1 / \epsilon} \cdot D+n^{\epsilon}\right)$ rounds. We note that computing $(\widetilde{O}(D), \widetilde{O}(1))$ cycle covers boils into the following problem:

Given a graph $G=(V, E)$ compute (in the CONGEST model) a maximal collection of edge disjoint $O(\log n)$-length cycles.

Note that in the LOCAL model, this problem is easy, and the main challenges comes from the bandwidth limitations of the CONGEST model.

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## A Balanced Partitioning of a Tree

We show that every rooted tree $T$ can be partitioned into two edge-disjoint rooted trees $T_{1}$ and $T_{2}$ such that (I) $E\left(T_{1}\right) \cup E\left(T_{2}\right)=E(T)$ and (II) $V\left(T_{1}\right), V\left(T_{2}\right) \leq 2 / 3 \cdot N$ where $N=|T|$. In addition, this partitioning maintains the layering structure of $T$ as will be described later. To compute this partitioning, define the weight $w(v)$ of each vertex $v$ in $T$ to be the number of vertices in its subtree $T(v)$. First, consider the case, where there is a vertex $v^{*}$ with weight $w\left(v^{*}\right) \in[1 / 3 N, 2 / 3 N]$. In such a case, define $T_{1}=T_{v^{*}}$ and $T_{2}=T \backslash E\left(T\left(v^{*}\right)\right)$. By definition, both $T_{1}$ and $T_{2}$ are trees, all edges of $T$ are covered and $\left|T_{1}\right|,\left|T_{2}\right| \in[1 / 3 N, 2 / 3 N]$.

Else, if no such balanced vertex exists, there must be a vertex $v^{*}$ such that $w\left(v^{*}\right) \geq 2 / 3 N$ but for each of its children in $T, u_{i}$, it holds that $w\left(u_{i}\right) \leq 1 / 3 N$. In such a case, we consider the children of $v^{*}$ from left to right $u_{1}, \ldots, u_{k}$ and sum up their weights until we get to a value in the range $[1 / 3 N, 2 / 3 N]$. Formally, let $\ell \in\{1, \ldots, k\}$ be the minimal index satisfying that
$\sum_{i=1}^{\ell} w\left(u_{i}\right) \in[1 / 3 N, 2 / 3 N]$. Since each $w\left(u_{i}\right) \leq 1 / 3 N$, such an index $\ell$ exists. We then set $T_{1}=$ $\bigcup_{i=1}^{\ell}\left(T\left(u_{i}\right) \cup\left\{\left(u_{i}, v^{*}\right)\right\}\right)$ and $T_{2}=T \backslash \bigcup_{i=1}^{\ell} V\left(T\left(u_{i}\right)\right)$. By construction, all edges of $T$ are covered by $T_{1}$ and $T_{2}$. In addition, by definition, $\left|T_{1}\right| \in[1 / 3 N, 2 / 3 N]$ and hence also $T_{2} \in[1 / 3 N, 2 / 3 N]$.

Finally, we pick the roots $r_{1}, r_{2}$ of $T_{1}, T_{2}$ (respectively) to be the vertices the are close-most to the root $r$ in $T$. We then get for $u, v \in T_{1}$, that if $u$ is closer to the root than $v$ in $T$, then also $u$ is closer to the root $r_{1}$ than $v$ in $T_{1}$.

## B Missing Details for Algorithm DistCycleCover

Postorder Numbering on a Tree. Given a BFS tree $T$, We now show a procedure which assigns the vertices numbers $N: V \rightarrow[1, n]$ according to a postorder traversal on $T$ in $O(D)$ rounds. Each node $u$ first computes the number of vertices in $T(u)$. This can be done in $O(D)$ rounds by working from the leafs up. At the end, each node also knows the number of nodes in the subtree of each of its children. For a range of integers $R=[i, j]$, let $\max (R)=j$ and $\min (R)=i$. By working from root to leaf nodes, each vertex $u$ computes a range $R(u)$, indicating the bucket of post-order numbers given to its vertices in $T(u)$. Let $R(s)=[1, n]$. Given that a vertex $v$ has received its range $R(v)$, it sends to its children their ranges in the following manner. Let $u_{1}, \ldots, u_{\ell}$ be the children of $v$ ordered from left to right. Then $R\left(u_{1}\right)=\left[\min (R(v)), \min (R(v))+\left|T\left(u_{1}\right)\right|-1\right]$ and for every $i \geq 1, R\left(u_{i}\right)=\left[\max \left(R\left(u_{i-1}\right)\right)+1, \max \left(R\left(u_{i-1}\right)\right)+\left|T\left(u_{i}\right)\right|\right]$. This proceeds for $O(D)$ rounds, at the end every $u$ knows $R(u)$. Now, each node $u$, sets its own number $N(u)=\max \left(R\left(u_{i}\right)\right)$. Since by this numbering, each vertex has the maximum number in its subtree, it is indeed a postorder numbering.

Sending Information on Virtual Edges. For every virtual edge ( $a, a^{\prime}$ ) added in phase $i \geq 1$, we show in the analysis section, that there is a precomputed $a-a^{\prime}$ walk $W_{a, a^{\prime}}$ in $G$ of length $O\left(2^{i} \cdot D\right)$. We assume that in phase $i$, all the edges on the walk of virtual edge added in phase $j \leq i-1$, are already marked and that each such edge on the walk knows the endpoint of the virtual edges and the edges in $G$ that this walk should cover. The analysis shows that in the virtual edges added in phase $i-1$ are important of $O\left(2^{i-1}\right)$ edges in $G$.

Assume that it is given for all virtual edges added in phase $j \leq i-1$, we now show how the algorithm provides these properties for the virtual edges added in phase $i$. Assume that both edges $\left(a, a^{\prime}\right)$ and $\left(b, b^{\prime}\right)$ incident to the matched pair $\langle a, b\rangle$ are virtual. Then, we send the information on the walks $\pi\left(a, b, T_{B_{x}}\right), W_{a, a^{\prime}}$ and $W_{b, b^{\prime}}$. This allows the endpoints of virtual edge $a^{\prime}, b^{\prime}$ to learn about each other. In addition, by passing the identifiers of the $2^{i-1}$ edges that are supposed to be covered by the walks $W_{a, a^{\prime}}$ and $W_{b, b^{\prime}}$, all the nodes on the new walk $W_{a^{\prime}, b^{\prime}}=W_{a^{\prime}, a} \circ \pi\left(a, b, T_{B_{x}}\right) \circ W_{b, b^{\prime}}$ can get this information. This is done on for all the virtual edges added in phase $i$ simultaneously using the random delay approach. In the analysis section we show that the length of the walks $W_{a, a^{\prime}}$ is bounded by $O\left(2^{1 / \epsilon} D\right)$ and each edge appears on $O\left(n^{2 \epsilon}\right)$ many paths. Thus using the random delay approach, this can be done in $O\left(2^{1 / \epsilon} \cdot D+n^{2 \epsilon}\right)$ rounds.
Marking Edges on Cycles and Making Them Simple. We first make every edge $e$ (i.e., the endpoints of the edge) know the set of edges that are covered by the cycles that go through $e$. To do that, the endpoints of the virtual edges keep information of the covered edge endpoints. In phase 1 , this is easily obtained since the endpoint of the virtual edges are also the vertices whose two edges should be covered by the cycle. Assuming that this information is kept up to phase $i-1$, in phase $i$, when adding the virtual edge $u-v$, the combined information traverses through the
edge disjoint path $\pi(u, v)$ computed on the tree of the block.
We next make every edge $e$ know all the edges of the cycles that go through $e$. Since each edge appears on $n^{\epsilon}$ cycles and the length of the cycles is $O\left(2^{1 / \epsilon} \cdot D\right)$, using standard random delay techniques, it can be done in $\widetilde{O}\left(n^{2 \epsilon}+2^{1 / \epsilon} \cdot D\right)$ rounds. Once each edge $e$ sees the entire cycle, it can locally correct it to be simple as described in Section 4.1.

## C Distributed Construction of Neighborhood Cover in the Congest Model

In this section we describe how to construct neighborhood cover in the CONGEST model. As far as we know, the previous explicit constructions for neighborhood cover (such as [ABCP96]) are in the LOCAL model and use large messages. For the definition of $(k, t, q)$ neighborhood cover, see Definition 8. For ease of presentation, we construct slightly weaker covers where the diameter of each cluster is $O(k \cdot t \cdot \log n)$ rather than $O(k \cdot t)$ as in Definition 8 . This construction is implicit in the recent spanner construction of [EN17]. We show:

Lemma 12. For every integer $t$, and every $n$-vertex graph $G=(V, E)$, one can construct in in $O(k \cdot t \cdot \log n)$ rounds, an $(k, t, q)$ neighborhood cover with $k=2 \log n, q=O(\log n)$ and the strong diameter of each cluster is $O(t \cdot k \cdot \log n)$, w.h.p.

We first describe how using the spanner construction of [EN17], we get a neighborhood cover that succeeds with constant probability. That is, we show that using the algorithm of [EN17] for constructing a $(k \cdot t)$-spanner, one can get in $O(k t \log n)$ rounds, a collection of subsets $\mathcal{S}=$ $\left\{S_{1}, \ldots, S_{n}\right\}$ such that (I) the diameter of each $G\left[S_{i}\right]$ is $O(k \cdot t \cdot \log n)$, (II) w.h.p., each vertex belongs to $O\left(k \cdot n^{1 / k}\right)$ sets and (III) for each vertex $v$, there is a constant probability that there exists $S_{i}$ that contains its entire $t$-neighborhood. Repeating these procedure for $O(\log n)$ many times yields the final cover.

We now describe the phase $i=\{1, \ldots, \Theta(\log n)\}$ where we construct a collection of $n$ sets $S_{i, u_{1}}, \ldots, S_{i, u_{n}}$ that satisfy (I-III). Each vertex $u \in V$ samples a radius $r_{u}$ from the exponential distribution ${ }^{14}$ with parameter $\beta=\ln (c \cdot n) /(3 k \cdot t)$. Each vertex $w$ that received a message originated at $u$, stores $m_{u}(w)=r_{u}-\operatorname{dist}(w, u, G)$, and also a neighbor $p_{u}(w)$ that lies on the shortest path from $w$ to $u$. Let $m(w)=\max _{u \in V} m_{u}(w)$. For every vertex $u$, let $S_{i, u}=\left\{w \mid m_{u}(w) \geq m(w)-1\right\}$. The final neighborhood cover is given by $\mathcal{S}=\bigcup_{i} \bigcup_{u} S_{i, u}$.

We now show that the output collection of sets are indeed neighborhood cover. We fix one phase $i$, and claim the following about the output sets $S_{i, u_{1}}, \ldots, S_{i, u_{n}}$.

Lemma 13. (I) Each $S_{i, u}$ is connected with diameter $O(k \cdot t \log n)$ with high probability.
(II) For every $i$, every vertex $w$ appears in $O\left(\log n \cdot(c n)^{1 /(k t)}\right)$ sets $S_{i, u}$ with high probability.
(III) For every vertex $w$, there exists $S_{i, u}$ such that $\Gamma_{t}(w) \subseteq S_{i, u}$, with constant probability.

Proof. For each of notation, let $S_{i, u}=S_{u}$. For every $u$ and $w$, let $p_{u}(w)$ be the neighbor of $w$ that lies on the shortest path from $w$ to $u$, from which $w$ received the message about $u$ (breaking ties based on IDs).

[^14]To show that each set $S_{u}$ is connected, it is sufficient to show that if $w \in S_{u}$ then also $p_{u}(w) \in S_{u}$. The proof is as Claim 5 in [EN17]. In particular, since $w$ and $w^{\prime}=p_{u}(w)$ are neighbors, it holds that $m(w) \geq m\left(w^{\prime}\right)-1$ and hence $w^{\prime} \in S_{u}$. In addition, by Claim 3 in [EN17] (and plugging our value of $\beta$ ) it holds that w.h.p. for every $u, r_{u} \leq O(k \cdot t \log n)$.

We proceed with Claim (II). For each $w$ and $u$, let $X_{w, u} \in\{0,1\}$ be the random variable indicating that $w \in S_{u}$. Let $Q_{w}=\sum_{u} X_{w, u}$ be the random variable of the number of sets to which $u$ belongs. Let $Z_{w}$ be the random variable of the value $m(w)$. By Lemma 2 of [EN17], $\mathbb{E}\left(Q_{u}\right) \leq(c n)^{1 /(3 k t)}$. By the proof of Lemma 2, it also holds that $\mathbb{E}\left(Q_{u} \mid Z_{w}=z\right) \leq(c n)^{1 /(3 k t)}$ for every value $z$ (that is, the bound holds even when conditioning it on particular value of $Z_{w}$ ). We now claim that for every $z, \mathbb{P}\left(Q_{u} \geq c^{\prime} \cdot \log n \cdot(c n)^{1 /(3 k t)} \mid Z_{w}=z\right) \leq 1 / n^{c^{\prime}}$. We first claim that when conditioning on a specific value of $Z_{w}$, the variables $X_{w, u}$ and $X_{w, u^{\prime}}$ are independent for $u \neq u^{\prime}$. Let $Z_{w}=z$, then $X_{w, y}=1$ if $r_{y} \geq \operatorname{dist}(y, w, G)+z-1$ for $y \in\left\{u, u^{\prime}\right\}$. In particular, $\mathbb{P}\left(X_{w, y}=1 \quad \mid \quad Z_{w}=z\right)=\mathbb{P}\left(r_{y} \geq \operatorname{dist}(y, w, G)+z-1\right)$. Since the random choices of $r_{y}$ are independent, it holds that $\mathbb{P}\left(X_{w, u}=1\right.$ and $\left.\quad X_{w, u^{\prime}}=1 \quad \mid \quad Z_{w}=z\right)=\mathbb{P}\left(X_{w, u}=1 \quad \mid \quad Z_{w}=\right.$ $z) \cdot \mathbb{P}\left(X_{w, u^{\prime}}=1 \mid Z_{w}=z\right)$.

Using Hoeffding Inequality of Lemma 14, we get that $\mathbb{P}\left(Q_{u} \geq c^{\prime} \cdot \log n \cdot(c n)^{1 /(3 k t)} \mid Z_{w}=z\right) \leq$ $1 / n^{c^{\prime}}$. Since it holds for every $z$, the claim follows.

Finally, we consider claim (III). We consider the more strict event, $Y_{w}$, in which the entire $t$-neighborhood of $w$ belongs to $S_{u^{*}}$ where $u^{*}$ is the vertex that attains $m(w)=m_{u^{*}}(w)$ (breaking ties based on IDs). We now show that $Y_{w}=0$ with probability of at most constant. Consider $u^{*}$ as above, we now bound the probability that there is a vertex $y \in \Gamma_{t}(w)$ that does not belong to $S_{u^{*}}$. We therefore have:

$$
\begin{equation*}
m(y)>m_{u^{*}}(y)-1=r_{u^{*}}-\operatorname{dist}\left(u^{*}, y, G\right)-1 \geq r_{u^{*}}-\operatorname{dist}\left(u^{*}, w, G\right)-t-1=m(w)-t-1, \tag{1}
\end{equation*}
$$

and in the same manner, $m(w) \geq m(y)-t-1$. Therefore, $m(w) \in[m(y)-t-1, m(y)+t+1]$. This happens with probability of $1-e^{-3 \beta \cdot t}=1-\ln (c n) / k \leq c^{\prime}$ for $k=2 \log n$. The claim follows.

We are now ready to complete the proof of Lemma 12.
Proof. It is easy to see that each phase can be implemented in $O(k \cdot t \log n)$ rounds (see Sec. 2.1.1 in [EN17] for more details), hence overall the construction takes $O\left(k \cdot t \log ^{2} n\right)$ rounds. By Lemma 13, w.h.p., all subsets have small diameter and bounded overlap. In addition, for every verte $w$, with constant probability, the entire $t$-neighborhood of $w$ is covered by some of the $S_{i, u}$ sets. Since we repeat this process for $O(\log n)$ many times, w.h.p., there exists a set that covers $\Gamma_{t}(w)$. By applying the union bounded overall sets, we get that w.h.p. all vertices are covered, the Lemma follows.

Lemma 14. [Hoeffding Inequality] Let $X_{1}, \ldots, X_{n}$ be independent random variables bounded by the interval $[0,1]$, i.e, $0 \leq X_{i} \leq 1$. Let $Q=\sum_{i} X_{i}$. Then, $\mathbb{P}(Q-\mathbb{E}(Q) \geq t) \leq e^{-2 n \cdot t^{2}}$.


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[^1]:    ${ }^{1}$ The term PSM stands for Private Simultaneous Model.

[^2]:    ${ }^{2}$ The $\widetilde{O}(\cdot)$ notation hides polylog $(n)$ factors where $n$ is the number of vertices in the graph.

[^3]:    ${ }^{3}$ A graph $G=(V, E)$ is 2-vertex connected if for all $u \in V$ the graph $G^{\prime}=(V \backslash\{u\}, E)$ is connected.

[^4]:    ${ }^{4}$ We observe that Algorithm 19 of [BE13] can implemented with $O(\log \Delta)$ memory.

[^5]:    ${ }^{5}$ The graph $G \backslash T$ might be disconnected, when referring to its diameter, we refer to the maximum diameter in each connected component of $G \backslash T$.

[^6]:    ${ }^{6}$ Recall that the graph $G$ is two edge connected.

[^7]:    ${ }^{7}$ This partitioning procedure is described in Appendix A. We note that this partitioning maintains the layering structure of $T^{\prime}$.

[^8]:    ${ }^{8}$ Since our partitioning into $T_{1}^{\prime}, T_{2}^{\prime}$ maintains the layering structure of $T$, it also holds that $P_{e}$ is below $e$ on $T_{1}^{\prime}$.

[^9]:    ${ }^{9}$ The strong diameter of a cluster of the diameter of the subgraph induced by the vertices of the cluster.

[^10]:    ${ }^{10}$ The ID of the $P_{2}$ path can be simply the concatenation of the three node IDs.

[^11]:    ${ }^{11}$ In fact it is sufficient to know if there is exactly one block in $T_{B}(u)$ or more.

[^12]:    ${ }^{12}$ When saying simultaneously we mean using the random delay approach.

[^13]:    ${ }^{13}$ We say that an edge $(u, v)$ knows a piece of information, if at least one of the edge endpoints know that.

[^14]:    ${ }^{14}$ Recall the exponential distribution with parameter $\beta$ where $f(x)=\beta \cdot e^{-\beta \cdot x}$ for $x \geq 0$ and 0 otherwise.

