

# Fully Dynamic Maximal Independent Set in Expected Poly-Log Update Time

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**Abstract**—In the fully dynamic maximal independent set (MIS) problem our goal is to maintain an MIS in a given graph  $G$  while edges are inserted and deleted from the graph. The first non-trivial algorithm for this problem was presented by Assadi, Onak, Schieber, and Solomon [STOC 2018] who obtained a deterministic fully dynamic MIS with  $O(m^{3/4})$  update time. Later, this was independently improved by Du and Zhang and by Gupta and Khan [arXiv 2018] to  $\tilde{O}(m^{2/3})$  update time<sup>1</sup>. Du and Zhang [arXiv 2018] also presented a randomized algorithm against an oblivious adversary with  $\tilde{O}(\sqrt{m})$  update time.

The current state of art is by Assadi, Onak, Schieber, and Solomon [SODA 2019] who obtained randomized algorithms against oblivious adversary with  $\tilde{O}(\sqrt{n})$  and  $\tilde{O}(m^{1/3})$  update times.

In this paper, we propose a dynamic randomized algorithm against oblivious adversary with expected worst-case update time of  $O(\log^4 n)$ . As a direct corollary, one can apply the black-box reduction from a recent work by Bernstein, Forster, and Henzinger [SODA 2019] to achieve  $O(\log^6 n)$  worst-case update time with high probability. This is the first dynamic MIS algorithm with very fast update time of poly-log.

## I. INTRODUCTION

A maximal independent set (MIS) of a given graph  $G = (V, E)$  is a subset  $M$  of vertices such that  $M$  does not contain two neighbor vertices and every vertex in  $V \setminus M$  has a neighbor vertex in  $M$ . In this paper, we study the maximal independent set (MIS) problem in the dynamic setting, where the graph  $G$  is undergoing a sequence of edge insertions and deletions.

MIS is a fundamental problem with both theoretical and practical importance and is used as a fundamental building block in many applications. For instance, MIS has been used for resource scheduling for parallel threads in a multi-core environment, for leader election [7], for resource allocation [13], etc.

The MIS had received a lot of attention in the distributed and parallel settings since the influential works of [1], [10], [11]. It is considered a central

<sup>1</sup>As usual  $n$  is the number of vertices,  $m$  is the number of edges and  $\tilde{O}(\cdot)$  suppresses poly-logarithmic factors.

problem in distributed computing and in particular in the symmetry breaking field. Specifically, attaining a better understanding of MIS in the distributed setting is of particular interest not only because it is a fundamental problem but also because other fundamental problems reduce to it.

Censor-Hillel, Haramaty, and Karnin [6] in their pioneering paper studied the problem of maintaining an MIS in the distributed dynamic setting where the graph changes over time. They gave a randomized algorithm for maintaining an MIS against an oblivious adversary in the distributed dynamic setting; as an open question, the authors asked whether it is possible to maintain an MIS in a dynamic graph with update time faster than recomputing everything from scratch, which triggered a recent line of research.

The first non-trivial algorithm was proposed by Assadi, Onak, Schieber and Solomon [2] who presented a deterministic algorithm with  $O(m^{3/4})$  amortized update time. This was the first dynamic algorithm that maintains an MIS with sublinear update time in the sequential model. This upper bound was later improved to  $\tilde{O}(m^{2/3})$  independently by Du and Zhang [8] and by Gupta and Khan [9]. In the same paper Du and Zhang [8] overcame the  $\tilde{O}(m^{2/3})$  barrier by assuming an oblivious adversary and a randomized algorithm with amortized  $\tilde{O}(\sqrt{m})$  was proposed. This randomized upper bound was recently improved to  $\tilde{O}(\sqrt{n})$  by Assadi *et al.* [3]. For graphs with bounded arboricity  $\alpha$ , a deterministic algorithm with amortized update time of  $O(\alpha^2 \log^2 n)$  was proposed in [12].

### A. Our contribution

In this paper we present the first dynamic MIS algorithm with very fast update time of poly-logarithmic in  $n$ . We obtain a randomized dynamic MIS algorithm that works against an oblivious adversary. Moreover, our algorithm can actually maintain a greedy MIS with respect to a random order on the set of vertices; the concept of greedy MIS is defined as follows.

**Definition 1.** Given any order  $\pi$  on all vertices in  $V$ , the greedy MIS  $M_\pi$  with respect to  $\pi$  is uniquely defined by the following procedure that gradually builds an MIS: starting with  $M_\pi = \emptyset$ , for each vertex in  $V$  under order  $\pi$ , if it is not yet dominated by  $M_\pi$ , add it to  $M_\pi$ .

We say that an algorithm has worst-case expected update time  $\alpha$  if for every update  $\sigma$ , the expected time to process  $\sigma$  is at most  $\alpha$ .

Our main result argues that when  $\pi$  is a uniformly random permutation, the corresponding greedy MIS can be maintained under edge updates against an oblivious adversary, which is formalized in the following statement.

**Theorem 2.** Let  $\pi$  be a random permutation over  $V$ . The greedy MIS on  $G$  according to order  $\pi$  can be maintained under edge insertions and deletions in worst-case expected  $O(\log^4 n)$  time against an oblivious adversary, where the expectation is taken over the random choice of  $\pi$ .

As a corollary, we can apply a black-box reduction from worst-case time dynamic algorithms to expected worst-case time dynamic algorithms that appeared in a recent paper [5].

**Theorem 3** ([5]). Let  $A$  be an algorithm that maintains a dynamic data structure  $D$  with worst-case expected time  $\alpha$ , and let  $n$  be a parameter such that the maximum number of items stored in the data structure at any point in time is polynomial in  $n$ , and let  $l$  be a parameter for the length of the update sequence to be considered. Then there exists an algorithm  $A'$  with the following properties.

- 1) For any sequence of updates  $\sigma_1, \sigma_2, \dots$ ,  $A'$  processes each update  $\sigma_i$  in  $O(\alpha \log^2 n)$  time with high probability.
- 2)  $A'$  maintains  $\Theta(\log n)$  data structures  $D_1, D_2, \dots, D_{\Theta(\log n)}$ , as well as a pointer to some  $D_i$  that is guaranteed to be correct at the current time. Query operations are answered with  $D_i$ .

**Corollary 4.** There is a dynamic MIS algorithm against an oblivious adversary that handles edge updates in worst-case  $O(\log^6 n)$  time with high probability, and answers MIS membership queries in constant time.

**Independent work:** Independent of our work, Behnezhad *et al.* [4] also presents a fully dynamic algorithm that maintains a greedy MIS with expected poly-logarithmic running time against oblivious adversaries.

## B. Technical overview

Our algorithm is a combination of techniques from [6] and [3]. In paper [6], the authors proved a lemma that the expected number of changes made to a greedy MIS by an edge update is bounded by a constant. Unfortunately, they could not achieve an efficient dynamic algorithm since a straightforward implementation of the lemma has a linear dependence on the maximum degree of the graph which could be large.

The issue with the maximum degree was overcome by the algorithm from [3] which relies on what we informally call the *degree reduction* lemma: if we pick a random subset of  $k$  vertices and build a greedy MIS on this subset, then the maximum degree of the induced subgraph on all the rest un-dominated vertices is at most  $O(\frac{n \log n}{k})$ . Therefore we can do the following to achieve an update time with sub-linear dependence on  $n$ . First build an MIS on a randomly selected subset of  $k$  vertices and then maintain an MIS on the induced subgraph of all the rest vertices in a brute-force manner. If an edge update lies entirely within the induced subgraph, then it takes time proportional to the maximum degree which is  $\tilde{O}(n/k)$ ; if an edge update lies within the random subset, then we rebuild the whole data structure from scratch. The expected running time of this algorithm is a trade-off between two terms. One the one hand, when the edge update occurs within the induced subgraph, the cost would be proportional to the maximum degree which is  $\tilde{O}(n/k)$ ; on the other hand, when the edge update connects two vertices in the random subset, the cost of rebuilding would be  $O(m) = O(n^2)$ , and under the assumption of obliviousness, the probability that an edge update lies within the random subset is roughly  $O(\frac{k^2}{n^2})$ , and so the expected time of rebuilding would be  $\tilde{O}(n^2 \cdot \frac{k^2}{n^2}) = \tilde{O}(k^2)$ . Taking  $k = \lfloor n^{1/3} \rfloor$  gives a balance of  $\tilde{O}(n^{2/3})$  update time. In their paper [3], the authors further refined the running time to  $\tilde{O}(\sqrt{n})$  using a hierarchical approach.

We believe the main bottleneck of the above algorithm is that it takes no effort to utilize the lemma from [6]. As a first attempt one could try to look for expensive parts of [3]'s algorithm and try to plug in [6]'s lemma. For example, instead of directly rebuilding, we could try to apply [6]'s lemma when restoring a greedy MIS among the random subset of  $k$  vertices if an edge update occurs between them. However, we would again encounter the large degree issue within the random subset.

Our new algorithm is a direct way of combining [6]'s lemma and the degree reduction lemma. The

algorithm keeps a random ordering  $\pi : V \rightarrow [n]$  of all vertices and tries to maintain the random greedy MIS. In order to do so, we explicitly maintain all the induced subgraphs  $G_i = (V_i, E_i)$  ( $0 \leq i \leq \log n$ ) on all vertices which are not dominated by MIS vertices from  $\pi^{-1}(1), \pi^{-1}(2), \dots, \pi^{-1}(2^i)$ . For simplicity assume edge  $(u, v)$  is inserted where  $2^b < \pi(u) < \pi(v) \leq 2^{b+1}$  for some integer  $b$ . Then, on the one hand, this event happens with probability  $O(2^{2b}/n^2)$  when  $\pi$  is uniformly random; on the other hand, all changes to the MIS could only take place in  $G_b$  whose maximum degree is bounded by  $O(\frac{n \log n}{2^b})$ .

Let  $S \subseteq V_b$  be the set of all influenced vertices (we will formally define what  $S$  is later on; basically  $S$  contains all vertices that could possibly enter or leave the MIS during this update). Following similar proofs of [6], we could prove the conditional expectation of  $S$  is at most  $O(n/2^b)$ . As the maximum degree of  $G_b$  is bounded by  $O(\frac{n \log n}{2^b})$ , we could go over all neighbors of  $S$  in  $G_b$  and maintain memberships of vertices from  $S$  in subgraphs  $G_{b+1}, G_{b+2}, \dots$ , which takes  $\tilde{O}(n^2/2^{2b})$  time, perfectly canceling out the probability  $O(2^{2b}/n^2)$  we just mentioned. However, this is not the end of the story. Not only could vertices from  $S$  change their memberships in subgraphs  $G_{b+1}, G_{b+2}, \dots$ , but neighbors of vertices in  $S$  as well, which could be as many as  $O(n^2/2^{2b})$  in the worst-case. The key to the running time analysis is that  $\pi$  roughly assigns the set  $S$  uniform-random positions in  $[2^b + 1, n]$  even when  $S$  is given as prior knowledge. Therefore, on average, the number of neighbors in  $G_b$  of a vertex in  $S$  is bounded by  $\tilde{O}(1)$ .

## II. PRELIMINARIES

For any subgraph  $H \subseteq G$ , let  $\Delta(H)$  be its maximum vertex degree. For any  $U \subseteq V$ , define  $\Gamma(U)$  to be the set of all neighbors of  $U$  in  $G$ , and  $G[U]$  the induced subgraph of  $G$  on  $U$ . For any permutation  $\pi$  on  $V$  and vertex  $u \in V$ , define  $I_u^\pi$  to be the set of neighbor predecessors of  $u$  with respect to  $\pi$ . For any two different vertices  $u, v \in V$ , we say  $u$  has a *higher priority* than  $v$  if  $\pi(u) < \pi(v)$ . For any pair of indices  $i, j$ , define  $\pi[i, j] = \{w \mid i \leq \pi(w) \leq j\}$ . The following lemma states the basic characterization of a greedy MIS.

**Lemma 5** (folklore). *An MIS  $M$  is the greedy MIS with respect to order  $\pi$  if and only if for all  $z \in V$ , it satisfies the constraint that either  $z \in M$  or  $I_z^\pi \cap M \neq \emptyset$ . For the rest, we will call this constraint the greedy MIS constraint for  $z$ .*

The following lemma appeared in [3].

**Lemma 6** ([3]). *Let  $\pi$  be a uniformly random permutation on  $V$  and let  $k$  be an integer in  $[n]$ . Let  $U$  be the set of all vertices not dominated by  $M_\pi \cap \pi[1, k]$ , then with high probability of  $1 - n^{-4}$ ,  $\Delta(G[U]) \leq O(\frac{n \log n}{k})$ .*

The next lemma is a slight modification of the previous lemma where we show that even if we fix the position in the permutation of two vertices the lemma still holds.

**Lemma 7.** *Let  $u_1, u_2 \in V$  be two different vertices and  $k_1, k_2 \in [n]$  be two different indices, and let  $1 \leq k \leq n$  be an integer. Let  $\pi$  be a uniformly random permutation on  $V$  under the condition that  $\pi(u_i) = k_i, i \in \{1, 2\}$ . Let  $U$  be the set of all vertices not dominated by  $M_\pi \cap \pi[1, k]$ , then with high probability  $1 - n^{-2}$ ,  $\Delta(G[U]) \leq O(\frac{n \log n}{k})$ .*

*Proof:* Call a permutation  $\pi$  *bad* if  $\Delta(G[U]) \geq \Omega(\frac{n \log n}{k})$ . Noticing that  $\Pr_\pi[\pi(u_i) = k_i, i \in \{1, 2\}] = \frac{1}{n(n-1)/2}$ , by Lemma 6 we have:

$$\begin{aligned} n^{-4} &\geq \Pr_\pi[\pi \text{ is bad}] \\ &= \frac{1}{n(n-1)/2} \Pr_\pi[\pi \text{ is bad} \mid \forall i, \pi(u_i) = k_i] \\ &\quad + (1 - \frac{1}{n(n-1)/2}) \Pr_\pi[\pi \text{ is bad} \mid \exists i, \pi(u_i) \neq k_i] \\ &\geq \frac{1}{n(n-1)/2} \Pr_\pi[\pi \text{ is bad} \mid \pi(u_i) = k_i] \end{aligned}$$

Hence,  $\Pr_\pi[\pi \text{ is bad} \mid \forall i, \pi(u_i) = k_i]$  is at most  $n^{-2}$  as well, which concludes the proof.  $\blacksquare$

For the rest of this section, we review the notion of *influenced set* which was studied in [6]. Given a total order  $\pi$ , an MIS  $M = M_\pi$ , as well as an edge update between  $u, v$ , we turn to define  $v$ 's influenced set  $S_v^\pi$ . If  $v$  does not violate the greedy MIS constraint after the edge update, then define  $S_v^\pi = \emptyset$ ; notice that  $v$  always preserves the greedy MIS constraint if  $\pi(v) < \pi(u)$ . Otherwise, initially set  $S_0 = \{v\}$ . For any  $i \geq 1$ , define  $S_i$  to be the set of all non-MIS vertices whose MIS predecessors are all in  $S_{i-1}$ , plus the set of every MIS vertex who has at least one predecessor in  $S_{i-1}$ , namely:

$$\begin{aligned} S_i &= \{w \mid w \in M, S_{i-1} \cap I_w^\pi \neq \emptyset\} \\ &\cup \{w \mid w \notin M, I_w^\pi \cap M \subseteq \bigcup_{j=0}^{i-1} S_j\} \end{aligned}$$

Note that the set  $M$  refers to the greedy MIS in the old graph, not in the new graph. Eventually, define  $v$ 's influenced set to be  $S_v^\pi = \bigcup_{i=0}^{\infty} S_i$ . When  $S_v^\pi \neq \emptyset$ , there is a simple characterization which will be used later.

**Lemma 8.** Let  $M$  be the greedy MIS in the old graph. When  $S_v^\pi \neq \emptyset$ , it is equal to the smallest set  $S$  that contains  $v$  and satisfies the following two conditions. (1)  $\forall z \in M, I_z^\pi \cap S \neq \emptyset$  iff  $z \in S$ . (2)  $\forall z \notin M, I_z^\pi \cap M \subseteq S$  iff  $z \in S$ .

*Proof:* Since  $S_v^\pi$  satisfies both of (1) and (2), it suffices to prove that any  $S$  containing  $v$  that satisfies both (1) and (2) would contain  $S_v^\pi$  as a subset. This is done by an easy induction on  $i \geq 0$  that  $S$  contains every  $S_i$ . ■

The proofs of the following two lemmas are given for completeness in the appendix.

**Lemma 9** ([6]). Let  $\pi, \sigma$  be two permutations,  $S \subseteq V$  a nonempty set, and  $v \in V$  be an arbitrary vertex. Suppose an edge update occurs between  $u, v$ . Assume  $S_v^\pi = S$ ,  $\pi(u) < \pi(v)$ ,  $\sigma(u) < \sigma(v)$ ,  $\sigma, \pi$  have the same induced relative order on both  $S$  and  $V \setminus S$ , namely  $\pi_S = \sigma_S$ ,  $\pi_{V \setminus S} = \sigma_{V \setminus S}$ , then  $M_\pi = M_\sigma$  in the old graph before the edge update, and  $S_v^\sigma = S$ .

**Lemma 10** ([6]). Let  $\pi, \sigma$  be two permutations,  $S \subseteq V$  a vertex subset, and  $v \in V$  be an arbitrary vertex. Suppose an edge update occurs between  $u, v$ . If  $S_v^\pi = S \neq \emptyset$ , and  $\pi(u) < \pi(v)$ ,  $\sigma(u) < \sigma(v)$ ,  $\sigma, \pi$  have the same induced relative order on both  $S \setminus \{v\}$  and  $V \setminus S$ , namely  $\pi_{S \setminus \{v\}} = \sigma_{S \setminus \{v\}}$ ,  $\pi_{V \setminus S} = \sigma_{V \setminus S}$ . If  $v \neq \arg \min_{z \in S} \{\sigma(z)\}$  then  $S_v^\sigma = \emptyset$ .

It was also shown in [6] that for an edge update  $(u, v)$  the expected size of  $S_v^\pi$  is constant. In our algorithm we need the following different variants of this claim; the proofs are deferred to the appendix.

**Lemma 11.** Suppose an edge update occurs between  $u, v$ . Let  $1 \leq A < B \leq n$  be two integers. Then

$$\mathbb{E}_\pi[|S_v^\pi| \mid \pi(u) = A, \pi(v) \in [A + 1, B]] < \frac{n}{B - A}$$

**Lemma 12.** Suppose an edge update occurs between  $u, v$ . Let  $1 \leq A < B \leq n$  be two integers. Then

$$\mathbb{E}_\pi[|S_v^\pi| \mid A < \pi(u) < \pi(v) \leq B] < \frac{2n}{B - A}$$

### III. THE MAIN ALGORITHM

In this section we describe our fully dynamic MIS algorithm.

#### A. Data structure

When  $\pi$  is a fixed permutation over  $V$ , our algorithm is entirely deterministic. Let  $M \subseteq V$  be the greedy MIS with respect to  $\pi$ , and for any  $1 \leq k \leq n$ , define  $M_k = M \cap \pi[1, k]$ . Since  $M$  is defined in a greedy manner,  $M_k$  dominates the entire set  $\pi[1, k]$ .

The algorithm explicitly maintains the induced subgraph  $G_i = (V_i, E_i), \forall 0 \leq i \leq \log n - 1$ , where  $V_i = V \setminus (M_{2^i} \cup \Gamma(M_{2^i}))$ ; by definition  $G_0 \supseteq G_1 \supseteq G_2 \supseteq \dots \supseteq G_{\log n - 1}$ . More precisely, given a permutation  $\pi$ , our algorithm maintains at any given point of time the graphs  $G_i$  for  $0 \leq i \leq \log n - 1$  and the greedy MIS  $M_\pi$ . In the following subsection we describe our update algorithm to maintain both the graphs  $G_i$  and the MIS  $M_\pi$ .

#### B. Update algorithm

Suppose an edge is updated, either inserted or deleted, between  $u, v \in V$  with  $\pi(u) < \pi(v)$ . Suppose  $2^a < \pi(u) \leq 2^{a+1}$  and  $2^b < \pi(v) \leq 2^{b+1}$  for integers  $a$  and  $b$ . There are several *easy cases*, where  $S_v^\pi = \emptyset$  and thus we do not need to make changes to  $M$  as  $M$  stays the greedy MIS with respect to  $\pi$ , and we only need to maintain the subgraphs  $G_0, G_1, \dots, G_{\log n - 1}$ .

- (i)  $u \notin M$ . In this case, we simply add or remove, depending whether the edge update is an insertion or deletion, the edge  $(u, v)$  to/from  $E_0, E_1, \dots, E_i$ , where  $i$  is the largest index such that  $u, v \in V_i$ .
- (ii)  $u \in M, v \notin M$ , the update is a deletion and  $I_v^\pi \cap M \neq \{u\}$ . This case can be handled in the same way as in (1): remove the edge  $(u, v)$  in  $E_0, E_1, \dots, E_i$ , where  $i$  is the largest index such that  $u, v \in V_i$ , and recompute  $v$ 's position in the subgraphs  $G_a, G_{a+1}, \dots, G_{\log n - 1}$ .
- (iii)  $u \in M, v \notin M$  and the update is an insertion. In this case, if  $v \in V_a$ , then since now  $v$  is dominated by  $u \in V_a$  we should remove  $v$  from all subgraphs  $G_k, \forall k > a$ . After that, add  $(u, v)$  to  $E_0, E_1, \dots, E_i$ , where  $i$  is the largest index such that  $u, v \in V_i$ .

For the rest of this section we consider the case where an edge is inserted between  $u, v \in M$ , or deleted between  $u \in M, v \notin M$  with  $I_v^\pi \cap M = \{u\}$ . In both of these cases,  $S_v^\pi \neq \emptyset$  and thus we need to change  $v$ 's status in the MIS, and then we must try to fix the greedy MIS  $M$  within  $G_b$ . We start by computing the nonempty influenced set  $S_v^\pi$  with respect to edge update between  $u, v$ .

- (1) Initialize an output set  $S = \emptyset$  that is promised to be equal to  $S_v^\pi$  by the end of the algorithm, as well as a set  $T = \{v\}$  that contains a set of candidate vertices that might be included in  $S$  during the process.
- (2) In each iteration, extract  $z = \arg \min_{z \in T} \{\pi(z)\}$  from  $T$ . If  $z \in M$ , then suppose  $2^k < \pi(z) \leq 2^{k+1}$ ; by definition it must be  $z \in V_k$ . First we add  $z$  to  $S$ , and scan all neighbors  $w$  of  $z$  in  $V_k$  such that  $\pi(w) > \pi(z)$  and add  $w$  to  $T$ .

If  $z \notin M$ , first scan its adjacency list in  $G_b$ ; if all its MIS neighbors with higher priority are in  $S$ , then add  $z$  to  $S$  and add all of its MIS neighbors  $w \in V_b$  with  $\pi(w) > \pi(z)$  to  $T$ .

(3) When  $T$  becomes empty, output  $S$  as  $S_v^\pi$ .

For convenience we summarize the above procedure as pseudo-code 1.

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**Algorithm 1: FindInfluencedSet( $u, v, b$ )**

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1  $S \leftarrow \emptyset$ , in easy cases (i)(ii)(iii)  $T \leftarrow \emptyset$ , and
   otherwise  $T \leftarrow \{v\}$ ;
2 while  $T \neq \emptyset$  do
3    $z \leftarrow \arg \min_{z \in T} \{\pi(z)\}$ ,  $T \leftarrow T \setminus \{z\}$ ;
4   if  $z \in M$  then
5      $S \leftarrow S \cup \{z\}$ ;
6     suppose  $2^k < \pi(z) \leq 2^{k+1}$ , and assert
        $z \in V_k$ ;
7     for neighbor  $w \in V_k$  of  $z$  such that
        $\pi(w) > \pi(z)$  do
8        $T \leftarrow T \cup \{w\}$ ;
9   else
10    flag  $\leftarrow$  true;
11    for neighbor  $w \in V_b \cap M$  of  $z$  such that
        $\pi(w) < \pi(z)$  do
12      if  $w \notin S$  then
13         $\text{flag} \leftarrow$  false and break;
14    if flag then
15       $S \leftarrow S \cup \{z\}$ ;
16      for neighbor  $w \in V_b \cap M$  of  $z$  with
        $\pi(w) > \pi(z)$  do
17         $T \leftarrow T \cup \{w\}$ ;
18 return  $S$ ;
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It will be proved that the output  $S$  of Algorithm 1 is equal to  $S_v^\pi$ . Once we have found  $S = S_v^\pi$ , we can try to fix the greedy MIS by only looking at  $G[S]$ ; note that it might be the case that not every vertex in  $S$  needs to change its status in the MIS (for example if  $G[S]$  is a triangle and  $v$  is removed from  $M$  due to an insertion, we would not add both vertices in  $S$  to  $M$ ). If the edge update is an insertion, we first remove  $v$  from all  $V_k, k > a$ , and then compute the greedy MIS on  $G[S \setminus \{v\}]$  with respect to  $\pi$ ; if the edge update is a deletion, we add  $v$  to all  $V_k, \forall a < k \leq b$ , and then compute the greedy MIS on  $G[S]$  with respect to  $\pi$ .

Last but not least, we also need to update  $G_k, k \geq b + 1$ . This is done in the straightforward manner:

go over every vertex  $z$  that has changed its status in MIS in the increasing order with respect to  $\pi(z)$ . Assuming  $2^k < \pi(z) \leq 2^{k+1}$ , directly go over all of its neighbors in  $G_k$  and recompute their memberships in  $G_{b+1}, \dots, G_{\log n-1}$ . More specifically, consider the following two cases.

- (1) If  $z$  has been added to  $M$ , then for every neighbor  $w \in \Gamma(z) \cap V_k$ , we remove  $w$  from all  $G_l, l > k$ .
- (2) If  $z$  has been removed from  $M$ , then  $z$  belonged to  $V_k$  before the update. Instead of enumerating every neighbor from the current version of  $\Gamma(z) \cap V_k$ , we go over all of its old neighbors  $w \in V_k$  before the update, and compute their memberships in  $G_{b+1}, \dots, G_{\log n-1}$ .

We also summarize this procedure as pseudo-code 2. After that we can summarize the main update algorithm as pseudo-code 3.

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**Algorithm 2: FixSubgraphs( $S, b$ )**

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1 for  $z \in S$  that has changed its status, in the
   increasing order in terms of  $\pi$  do
2   assume  $2^k < \pi(z) \leq 2^{k+1}$ ;
3   if  $z$  has joined  $M$  then
4     for  $w \in V_k \cap \Gamma(z)$  do
5       remove  $w$  from all  $G_l, l > k$ ;
6   else if  $z$  has left  $M$  then
7     for neighbor  $w$  of  $z$  in the old version of  $V_k$ 
       before the edge update do
8       compute  $w$ 's memberships in
          $G_k, G_{k+1}, \dots, G_{\log n-1}$ ;
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### C. Correctness

In this section we prove the correctness of our algorithm. We start by proving that the algorithm correctly computed the set  $S_v^\pi$ .

**Lemma 13.** *Algorithm 1 correctly outputs the influenced set with respect to  $v$ , namely  $S = S_v^\pi$  when it terminates.*

*Proof:* Let  $v = z_1, z_2, \dots, z_l$  be the sequence of vertices that are added to  $S$  sorted in the increasing order with respect to  $\pi$ . We prove inductively that for any  $1 \leq i \leq l$ ,  $S_v^\pi \cap \pi[\pi(z_1), \pi(z_i)] = \{z_1, z_2, \dots, z_i\}$ . When  $i = 1$ , the equality holds trivially as  $S_v^\pi \cap \pi[\pi(z_1), \pi(z_i)] = \{v\} = \{z_1\}$ . For the inductive step, suppose we have  $S_v^\pi \cap \pi[\pi(z_1), \pi(z_i)] =$

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**Algorithm 3: Update( $u, v$ )**

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1 suppose  $\pi(u) < \pi(v)$ , and  $2^a < \pi(u) \leq 2^{a+1}$ ,  
    $2^b < \pi(v) \leq 2^{b+1}$ ;  
2  $S \leftarrow \text{FindInfluencedSet}(u, v, b)$ ;  
3 if  $S = \emptyset$  then  
4   recompute  $v$ 's memberships among  
    $G_a, G_{a+1}, \dots, G_{\log n-1}$ ;  
5 else  
6   if the update is insertion then  
7     remove  $v$  from  $V_k, k > a$ ;  
8     run the greedy MIS algorithm on  
      $G[S \setminus \{v\}]$  with respect to order  $\pi$ ;  
9   else  
10    add  $v$  to all  $V_k, a < k \leq b$ ;  
11    run the greedy MIS algorithm on  $G[S]$   
    with respect to order  $\pi$ ;  
12 FixSubgraphs( $S, b$ );
```

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$\{z_1, z_2, \dots, z_i\}$  for some  $i \geq 1$ . Next we prove  $S_v^\pi \cap \pi[\pi(z_i) + 1, \pi(z_{i+1})] = \{z_{i+1}\}$  in two steps.

- $z_{i+1} \in S_v^\pi$ .  
This can be verified according to the specification of Algorithm 1 and definition of  $S_v^\pi$  in the following way. If  $z_{i+1}$  were added to  $S$  on line-5, namely  $z_{i+1} \in M$ , then it must have been introduced to  $T$  on line-17 by a neighboring by one of its neighbor that appears before in  $\pi$ . Note that this predecessor cannot in  $M$ , and so it was added to  $S$  on line-15, and thus  $z_{i+1}$  was added to  $T$  on line-17. Then according to the definition of  $S_v^\pi$ ,  $z_{i+1} \in S_v^\pi$ . If otherwise  $z_{i+1}$  was added to  $S$  as a non-MIS vertex, then on the one hand  $z_{i+1}$  does not have MIS predecessor neighbors not in  $V_b$  as  $z_{i+1} \in V_b$ ; on the other hand,  $z_{i+1}$  can be added to  $S$  only when all of MIS its neighboring predecessors belong to  $\{z_1, z_2, \dots, z_i\} \subseteq S_v^\pi$ . Therefore, according to the definition of  $S_v^\pi$ , it should be  $z_{i+1} \in S_v^\pi$ .
- For any  $w \in \pi[\pi(z_i) + 1, \pi(z_{i+1}) - 1]$ ,  $w \notin S_v^\pi$ .  
Suppose we choose  $w \in S_v^\pi \cap \pi[\pi(z_i) + 1, \pi(z_{i+1}) - 1]$  with the smallest order in  $\pi$ . We first rule out the case where  $w \in M$ . If this should be the case, the  $w$  must be adjacent to a vertex  $z \in \{z_1, z_2, \dots, z_i\}$ ; this is not possible because  $w$  would have been added to  $T$ , when  $z$  was added to  $S$  on line-15, and then later it would be added to  $S$ .  
Now we suppose  $w \notin M$ . By definition of  $S_v^\pi$

and the inductive hypothesis, all preceding MIS neighbors of  $w$  belong to  $\{z_1, z_2, \dots, z_i\}$ . Let  $z \in \{z_1, z_2, \dots, z_i\}$  be the one with the smallest order among its MIS neighbors, and suppose  $2^k < \pi(z) < 2^{k+1}$ . Since  $z$  is the MIS vertex that dominates  $w$  with the smallest order, it must be  $w \in V_k$ , and therefore when  $z$  was added to  $S$  on line-5,  $w$  would be added to  $T$  on line-8, and later to  $S$  on line-15, which is a contradiction. ■

**Lemma 14.** *Algorithm 3 correctly restores the greedy MIS with respect to  $\pi$ .*

*Proof:* We only need to consider the case when  $S = S_v^\pi \neq \emptyset$  since otherwise no changes are made to the greedy MIS. We first claim that none of the vertices outside  $S$  need to change their status in the greedy MIS. This is because, on the one hand, for any  $z \in M \setminus S$ , by Lemma 8 we know  $I_z^\pi \cap S = \emptyset$ , and so any changes within  $S$  cannot affect  $z$ ; on the other hand, for any  $z \in V \setminus (M \cup S)$ , by Lemma 8, there exists a vertex from  $M \setminus S$  that dominates  $z$  as a predecessor, and therefore  $z$  stays a non-MIS vertex, irrespective of changes in  $S$ .

Secondly, we claim that recomputing the greedy MIS on  $G[S \setminus \{v\}]$  or  $G[S]$ , depending on whether the update is an insertion or a deletion, has no conflicts with MIS vertices in  $M \setminus S$ . This is because, again by Lemma 8, for any  $z \in S$  that was originally a non-MIS vertex,  $z$  is not adjacent to any MIS vertex from  $M \setminus S$ , and so adding  $z$  to  $M$  has no conflicts with vertices in  $M \setminus S$ . ■

**Lemma 15.** *In each iteration of the outermost loop of Algorithm 2, by the time when line-2 is executed,  $V_k$  is already equal to  $V \setminus (M_{2^k} \cup \Gamma(M_{2^k}))$ .*

*Proof:* We prove the claim by an induction on the value of  $\pi(z)$ . For the base case where  $z = v, k = b$ , note that the only possible change to  $V_b$  is  $v$ : if the edge update is an insertion, then  $v$  would leave  $V_b$ ; if the edge update is a deletion, then  $v$  might join  $V_b$ . In both cases, we have already fixed it right before recomputing the greedy MIS on  $G[S \setminus \{v\}]$  or  $G[S]$ . Since we turn to fix subgraphs  $G_b, G_{b+1}, \dots, G_{\log n-1}$  after we have finished restoring the greedy MIS, it should be  $V_b = V \setminus (M_{2^b} \cup \Gamma(M_{2^b}))$  at the beginning of Algorithm 2.

Next we turn to look at the inductive step. We first argue that any vertex  $w$  that leaves  $V \setminus (M_{2^k} \cup \Gamma(M_{2^k}))$  due to changes in  $S$  has already been removed from  $V_k$  in previous iterations. This is because we iterate over  $S$  in the increasing order with respect to  $\pi$ , and we must have already visited another vertex  $z' \in S \cap M$

with  $2^l < \pi(z') \leq 2^{l+1} \leq 2^k$  who is the earliest neighbor of  $w$ . By the inductive hypothesis, when  $z'$  was enumerated in the for-loop,  $V_l = V \setminus (M_{2^l} \cup \Gamma(M_{2^l}))$ , and thus  $w$  is removed from  $V_k$  by then.

Secondly we argue that any vertex  $w$  that joins  $V \setminus (M_{2^k} \cup \Gamma(M_{2^k}))$  due to changes in  $S$  has already been added to  $V_k$  in previous iterations. For  $w$  to join  $V \setminus (M_{2^k} \cup \Gamma(M_{2^k}))$ , it must have lost all of its MIS neighbors whose order is less or equal to  $2^k$ . Let  $z' \in S \setminus M$  be the one with smallest order and assume  $2^l < \pi(z') \leq 2^{l+1} \leq 2^k$ , and so  $z'$  must have been removed from  $M$  by Algorithm 3. By the inductive hypothesis, by the time when  $z'$  was enumerated by Algorithm 2, we fix all old neighbors of  $z'$  in  $V_l$ , which include  $w$ , and hence  $w$ 's memberships in  $G_l, G_{l+1}, \dots, G_{\log n-1}$  were already recomputed from scratch by then. ■

**Corollary 16.** *The update algorithm correctly maintains subgraphs  $G_0, G_1, \dots, G_{\log n-1}$  by the end of its execution.*

#### D. Running time analysis

Define  $\mathcal{B}$  to be the set of all permutations  $\pi$  such that there exists an index  $0 \leq k \leq \log n - 1$  for which  $\Delta(G_k) \geq \Omega(n \log n / 2^k)$  either before or after the edge update; we need to emphasize it here that the constant hidden in the  $\Omega(\cdot)$  notation is larger than the constant hidden in the notation  $O(\cdot)$  in the statement of Lemma 7.

**Lemma 17.** *Let  $a, b$  be fixed integers. Denote  $\mathcal{E} = \{\pi(u) < \pi(v), \pi(u) \in [2^a + 1, 2^{a+1}], \pi(v) \in [2^b + 1, 2^{b+1}]\}$ . Let  $T_0$  be the set of all vertices that have once belonged to  $T$ , and let  $T_1$  be the set of all vertices that need to change their memberships among subgraphs  $G_{b+1}, \dots, G_{\log n-1}$  during the execution of Algorithm 2. Note that in the easy cases where  $S_v^\pi = \emptyset$ , we have  $T = T_0 = T_1 = \emptyset$ . Then we have the following bound on the conditional expectation:*

$$\mathbb{E}_\pi[|T_0 \cup T_1| \mid \mathcal{E}] = O(n \log^2 n / 2^b + n^2 \cdot \Pr[\pi \in \mathcal{B} \mid \mathcal{E}])$$

We break the proof of the above lemma into several steps.

**Lemma 18.**  $\mathbb{E}_\pi[|S_v^\pi| \mid \mathcal{E}] = O(n/2^b)$ .

*Proof:* If  $a < b$ , then  $u$  belongs to  $\pi[1, 2^b]$ . Directly apply Lemma 11 by fixing an arbitrary position  $\pi(u) \in [2^a + 1, 2^{a+1}]$  and setting  $A = \pi(u), B = 2^{b+1}$ , and then we would have  $\mathbb{E}_\pi[|S_v^\pi| \mid \mathcal{E}] \leq n/(2^{b+1} - A) \leq n/2^b$ . If  $a = b$ , then  $u, v \in \pi[2^b + 1, 2^{b+1}]$ . Apply

Lemma 12 with  $A = 2^b, B = 2^{b+1}$ , and then we also have  $\mathbb{E}_\pi[|S_v^\pi| \mid \mathcal{E}] \leq n/2^{b-1}$ . ■

Fix any set  $S$  such that  $v \in S \subseteq V$ , as well as any relative order  $\sigma_+$  on  $S$  and any relative order  $\sigma_-$  on  $V \setminus S$ , such that there exists a permutation  $\pi$  with  $S_v^\pi = S, \pi_S = \sigma_+, \pi_{V \setminus S} = \sigma_-$ . Therefore, we can further decompose the conditional expectations as follows:

$$\begin{aligned} & \mathbb{E}_\pi[|T_0 \cup T_1| \mid \mathcal{E}] \\ &= \sum_{S, \sigma_+, \sigma_-} \Pr[S_v^\pi = S, \pi_S = \sigma_+, \pi_{V \setminus S} = \sigma_- \mid \mathcal{E}] \\ & \cdot \mathbb{E}_\pi[|T_0 \cup T_1| \mid \mathcal{E}, S_v^\pi = S, \pi_S = \sigma_+, \pi_{V \setminus S} = \sigma_-] \end{aligned}$$

Therefore, it suffices to study the upper bound on  $\mathbb{E}_\pi[|T_0 \cup T_1| \mid \mathcal{E}, S_v^\pi = S, \pi_S = \sigma_+, \pi_{V \setminus S} = \sigma_-]$ . For notational convenience, define  $\Omega = \{\pi \mid \mathcal{E}, \pi_S = \sigma_+, \pi_{V \setminus S} = \sigma_-\}$ . By Lemma 9, if there exists one  $\pi \in \Omega$  such that  $S_v^\pi = S$ , then all  $\pi \in \Omega$  would satisfy  $S_v^\pi = S$ ; plus  $\forall \pi \in \Omega$ , all  $M_\pi$ 's are the same in the old graph before the edge update, which we can safely denote as a common MIS  $M$ .

First we study the conditional expectation  $\mathbb{E}_\pi[|T_0| \mid \pi \in \Omega]$ . As can be seen from Algorithm 1, any vertex, which belonged to  $M$  before the edge update, that has once been added to  $T$  must have eventually joined  $S$ . So we only need to bound the total number of vertices in  $T_0 \setminus M$ . Again by Algorithm 1, any  $w \in T_0 \setminus M$  was added to  $T$  by an MIS predecessor  $z \in S$  on line-8. Therefore,  $|T_0 \setminus M|$  is bounded by the sum of (lower priority) neighbors of all  $z \in S \cap M$ . So it suffices to study individual contribution of all  $z \in S \cap M$  to  $T_0 \setminus M$ . Formally,  $\forall z \in S \setminus M, w \in T_0 \setminus M$ , we say  $z$  contributes  $w$  to  $T_0$  if  $w$  was added to  $T$  on line-8 when  $z$  is being processed. First we notice a basic property of  $T_0$ .

**Lemma 19.**  $v = \arg \min_{z \in T_0} \{\pi(z)\}$ , for any  $\pi \in \Omega$ .

*Proof:* This property is directly guaranteed by Algorithm 1: on line-8 or line-17, it only adds vertices  $w$  to  $T$  whose order is strictly larger than  $z$  who has just entered  $S$ . Since  $v$  is the first vertex that has been added to  $S$ , all vertices that join  $T$  should have larger order than  $v$ . ■

**Lemma 20.** For any  $k > b$ ,  $\mathbb{E}_{\pi \in \Omega}[|(S \setminus \{v\}) \cap \pi[2^b + 1, 2^k]|] < \frac{2^k |S|}{n}$ .

*Proof:* Decompose the expectation as following:

$$\begin{aligned} \mathbb{E}_{\pi \in \Omega}[|(S \setminus \{v\}) \cap \pi[2^b + 1, 2^k]|] &= \sum_{j=2^b+1}^{2^{b+1}} \Pr_{\pi \in \Omega}[\pi(v) = j] \\ & \cdot \mathbb{E}_{\pi \in \Omega}[|(S \setminus \{v\}) \cap \pi[2^b + 1, 2^k]| \mid \pi(v) = j] \end{aligned}$$

When  $\pi(v) = j$ , the rest of  $S \setminus \{v\}$  are free to choose positions on  $[j+1, n]$ , as  $v$  always takes the smallest order among  $S$ , which is guaranteed by Lemma 19 as  $S \subseteq T_0$ . Hence, for any  $l \in [1, \min\{2^k - j, |S| - 1\}]$ , conditioned on  $\pi(v) = j$ , the probability that  $|(S \setminus \{v\}) \cap \pi[2^b + 1, 2^k]| = l$  is equal to  $\frac{\binom{2^k - j}{l} \cdot \binom{n - 2^k}{|S| - 1 - l}}{\binom{n - j}{|S| - 1}}$ . Therefore, the expectation is computed as follows:

$$\begin{aligned}
& \mathbb{E}_{\pi \in \Omega} [|(S \setminus \{v\}) \cap \pi[2^b + 1, 2^k]| \mid \pi(v) = j] \\
&= \sum_{l=1}^{\min\{2^k - j, |S| - 1\}} l \cdot \Pr_{\pi \in \Omega} [|(S \setminus \{v\}) \cap \pi[2^b + 1, 2^k]| = l \mid \pi(v) = j] \\
&= \sum_{l=1}^{\min\{2^k - j, |S| - 1\}} l \cdot \frac{\binom{2^k - j}{l} \cdot \binom{n - 2^k}{|S| - 1 - l}}{\binom{n - j}{|S| - 1}} \\
&= \sum_{l=1}^{\min\{2^k - j, |S| - 1\}} (2^k - j) \cdot \binom{2^k - j - 1}{l - 1} \\
&\quad \cdot \binom{n - 2^k}{|S| - 1 - l} / \binom{n - j}{|S| - 1} \\
&= (2^k - j) \cdot \binom{n - j - 1}{|S| - 2} / \binom{n - j}{|S| - 1} \\
&= \frac{(2^k - j)(|S| - 1)}{n - j} < \frac{2^k |S|}{n}
\end{aligned}$$

Finally, we have:

$$\begin{aligned}
& \mathbb{E}_{\pi \in \Omega} [|(S \setminus \{v\}) \cap \pi[2^b + 1, 2^k]|] \\
&= \sum_{j=2^{b+1}}^{2^{b+1}} \Pr_{\pi \in \Omega} [\pi(v) = j] \\
&\quad \cdot \mathbb{E}_{\pi \in \Omega} [|(S \setminus \{v\}) \cap \pi[2^b + 1, 2^k]| \mid \pi(v) = j] \\
&< \sum_{j=2^{b+1}}^{2^{b+1}} \Pr_{\pi \in \Omega} [\pi(v) = j] \cdot \frac{2^k |S|}{n} = \frac{2^k |S|}{n}
\end{aligned}$$

**Lemma 21.** *The expected contribution of all  $z \in S \cap M \setminus \{v\}$  to  $T_0$  is at most  $O(|S| \log^2 n + |S|n \cdot \Pr_{\pi \in \Omega} [\pi \in \mathcal{B}])$ .*

*Proof:* Consider any index  $b \leq k \leq \log n - 1$ . When  $2^k < \pi(z) \leq 2^{k+1}$ , the total number of neighbors of  $z$  in  $V_k$  is at most  $O(n \log n / 2^k + n \cdot \mathbb{1}[\pi \in \mathcal{B}])$ , by definition of  $\mathcal{B}$ . Therefore, by Lemma 20 the expected total contribution of  $z \in S \cap M \setminus \{v\}$  to  $T_0$  that lies in  $[2^k + 1, 2^{k+1}]$  is bounded by  $O(|S| \log n + 2^k |S| \cdot \mathbb{1}[\pi \in \mathcal{B}])$ . Taking a summation over all  $k$  we can finalize the proof. ■

By Lemma 21, we have an upper bound on conditional expectation:

$$\begin{aligned}
& \mathbb{E}_{\pi} [|T_0| \mid \pi \in \Omega] \\
&\leq O(|S| \log^2 n + n \log n / 2^b + |S|n \cdot \Pr_{\pi \in \Omega} [\pi \in \mathcal{B}])
\end{aligned}$$

Here we have an extra additive term as an upper bound on the contribution of  $v$  to  $T_0$ .

Next we try to study  $\mathbb{E}_{\pi} [|T_1| \mid \pi \in \Omega]$ . By Algorithm 2, for any  $z \in S$  that has changed its status in  $M$ , we go over some of the neighbors of  $z$  and update their memberships in  $G_{k+1}, \dots, G_{\log n - 1}$  using brute force, and by definition these neighbors would belong to  $T_1$ . Similar to what we did with  $T_0$ , we say  $z$  *contributes* these neighbors to  $T_1$ . Next we need to carefully analyze the total number of these neighbors.

**Lemma 22.** *The expected contribution of all  $z \in S \setminus \{v\}$  to  $T_1$  is at most  $O(|S| \log^2 n + |S|n \cdot \Pr_{\pi \in \Omega} [\pi \in \mathcal{B}])$ .*

*Proof:* Let  $k \in [b, \log n - 1]$  be any index. Assume  $2^k < \pi(z) \leq 2^{k+1}$ . Consider the following two possibilities.

- $z$  has joined  $M$  during the update algorithm. In this case,  $z$  must belong to  $V_k$  and thus the total number of its neighbors in  $V \setminus (M_{2^k} \cup \Gamma(M_{2^k}))$  is at most  $O(n \log n / 2^k + n \cdot \mathbb{1}[\pi \in \mathcal{B}])$ , and by Lemma 15 we know  $V_k = V \setminus (M_{2^k} \cup \Gamma(M_{2^k}))$  by the time  $z$  is processed by Algorithm 2, and thus the total number of its neighbors in  $V_k$  is at most  $O(n \log n / 2^k)$ .
- $z$  has just left  $M$  during the update algorithm. In this case,  $z$  was selected by  $M$  and thus belonged to  $V_k$  before the update. As Algorithm 2 only iterates over  $z$ 's old neighbors in  $V_k$ , the total number of these neighbors is also bounded by  $O(n \log n / 2^k + n \cdot \mathbb{1}[\pi \in \mathcal{B}])$ .

In any case, the contribution of  $z$  to  $T_1$  is at most  $O(n \log n / 2^k + n \cdot \mathbb{1}[\pi \in \mathcal{B}])$ . Therefore, by Lemma 20 the expected total contribution of  $z \in S \cap M \setminus \{v\}$  to  $T_1$  that lies in  $[2^k + 1, 2^{k+1}]$  is bounded by  $O(|S| \log n + 2^k |S| \cdot \mathbb{1}[\pi \in \mathcal{B}])$ . Taking a summation over all  $k$  we can finalize the proof. ■

Taking a summation over all  $z \in S \setminus \{v\}$  that has changed its status in the MIS we have:

$$\begin{aligned}
& \mathbb{E}_{\pi} [|T_1| \mid \pi \in \Omega] \\
&\leq O(|S| \log^2 n + n \log n / 2^b + |S|n \cdot \Pr_{\pi \in \Omega} [\pi \in \mathcal{B}])
\end{aligned}$$

Here the extra additive term also stands for  $v$ 's contribution to  $T_1$ .

*Proof of Lemma 17:* To summarize, by Lemma 21 and Lemma 22, we have proved:

$$\begin{aligned} & \mathbb{E}_\pi[|T_0 \cup T_1| \mid \pi \in \Omega] \\ & \leq O(|S| \log^2 n + n \log n/2^b + |S|n \cdot \Pr_{\pi \in \Omega}[\pi \in \mathcal{B}]) \end{aligned}$$

Recall a previous decomposition we would then have:

$$\begin{aligned} & \mathbb{E}_\pi[|T_0 \cup T_1| \mid \mathcal{E}] \\ & = \sum_{S, \sigma_+, \sigma_-} \Pr_\pi[S_v^\pi = S, \pi_S = \sigma_+, \pi_{V \setminus S} = \sigma_- \mid \mathcal{E}] \\ & \cdot \mathbb{E}_\pi[|T_0 \cup T_1| \mid \mathcal{E}, S_v^\pi = S, \pi_S = \sigma_+, \pi_{V \setminus S} = \sigma_-] \\ & \leq \sum_{S, \sigma_+, \sigma_-} O(|S| \log^2 n + n \log n/2^b + |S|n \cdot \Pr_{\pi \in \Omega}[\pi \in \mathcal{B}]) \\ & \cdot \Pr_\pi[S_v^\pi = S, \pi_S = \sigma_+, \pi_{V \setminus S} = \sigma_- \mid \mathcal{E}] \\ & = \sum_S O(|S| \log^2 n \cdot \Pr_\pi[S_v^\pi = S \mid \mathcal{E}]) + O(n \log n/2^b) \\ & + \sum_{S, \sigma_+, \sigma_-} |S|n \cdot \Pr_\pi[S_v^\pi = S, \pi_S = \sigma_+, \pi_{V \setminus S} = \sigma_- \mid \mathcal{E}] \\ & \cdot \Pr_\pi[\pi \in \mathcal{B} \mid \mathcal{E}, S_v^\pi = S, \pi_S = \sigma_+, \pi_{V \setminus S} = \sigma_-] \\ & \leq O(\mathbb{E}_\pi[|S_v^\pi| \log^2 n \mid \mathcal{E}] + n \log n/2^b + n^2 \Pr_\pi[\pi \in \mathcal{B} \mid \mathcal{E}]) \\ & \leq O(n \log^2 n/2^b + n^2 \Pr_\pi[\pi \in \mathcal{B} \mid \mathcal{E}]) \end{aligned}$$

The last inequality holds by Lemma 18.  $\blacksquare$

To remove the extra term  $\Pr_\pi[\pi \in \mathcal{B} \mid \mathcal{E}]$ , apply Lemma 7 by fixing values of  $\pi(u), \pi(v)$  and taking union bound over all  $k$  equal to powers of 2, we would know that  $\pi \notin \mathcal{B}$  with high probability, namely  $\Pr_\pi[\pi \in \mathcal{B} \mid \mathcal{E}] = n^{-2} \log n$ , and thus  $\mathbb{E}_\pi[|T_0 \cup T_1| \mid \mathcal{E}_0] \leq O(n \log^2 n/2^b + \log n) = O(n \log^2 n/2^b)$ .

By definition of  $T_0$  and  $T_1$ , the total update time is proportional to  $\Delta(G_b) \cdot (|T_0| + |T_1|)$  whose expectation is then bounded by  $O(n^2 \log^3 n/2^{2b})$ . Since fixing the memberships of  $v$  takes time at most  $O(n \log^2 n/2^a)$ , it immediately says that the expected update time is  $O(n^2 \log^3 n/2^{2b} + n \log^2 n/2^a)$ . Since the adversary is oblivious to the randomness used in the algorithm, the probability of inserting an edge between  $V_a$  and  $V_b$  with respect to  $\pi$  is  $O(2^{a+b}/n^2)$ . Hence, the expected running time would be  $O(2^{a+b}/n^2 \cdot (n^2 \log^3 n/2^{2b} + n \log^2 n/2^a)) = O(2^{a-b} \log^3 n + \log^2 n)$ . Summing over all different indices  $a, b$ , the total time would be  $O(\log^4 n)$ .

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## APPENDIX

### A. Proof of Lemma 9

*Proof:* Here we present an conceptually simpler proof than the one presented in [6]. Notice that it suffices to consider the case where  $\sigma(z) = \pi(z), \forall z \notin \{x, y\}$  and  $\sigma(x) = \pi(y), \sigma(y) = \pi(x)$ , where  $x, y$  is an arbitrary pair of consecutive vertices in  $\pi$  such that  $x \in S$  and  $y \notin S$ . As  $\sigma(u) < \sigma(v), \pi(u) < \pi(v)$ , it can never be  $x = v$  and  $y = u$ . Denote  $M = M_\pi$  be the greedy MIS in the old graph. The proof follows from the two statements below.

**Claim 23.**  $M$  was also the greedy MIS on the old version of  $G$  with respect to  $\sigma$ .

*Proof:* Recall from Lemma 5,  $M$  is the greedy MIS with respect to  $\sigma$  in the old graph if  $M$  is an MIS and for all  $z \in V \setminus M, I_z^\sigma \cap M \neq \emptyset$ . The first half is easy:  $M$  was the greedy MIS in the old version of  $G$  with respect to  $\pi$ , so  $M$  is certainly an MIS in the old graph. Now we turn to verify the greedy MIS constraints.

Since  $\sigma$  agrees with  $\pi$  on every vertex except for  $\{x, y\}$ , we only need to verify  $\forall z \in \{x, y\} \setminus M, I_z^\sigma \cap M \neq \emptyset$ . We can assume  $x, y$  are neighbors in the updated graph  $G$ ; otherwise switching the orders between  $x, y$  in  $\pi$  does not affect the greedy MIS constraint. Consider several cases.

- $x \in M, y \notin M$ . In this case, if  $\pi(y) > \pi(x)$ , then  $\sigma(x) > \sigma(y)$ , and thus  $x \in I_y^\sigma \cap M \neq \emptyset$ . If  $\pi(y) < \pi(x)$ , then since  $y \notin S, I_y^\pi \cap M \setminus S$  must be nonempty, and so there exists  $z \in I_y^\pi \cap M \setminus S$  that dominates  $y$ . As  $\sigma(z) = \pi(z) < \pi(x) = \sigma(y)$ ,  $I_y^\sigma \cap M$  is also nonempty.
- $x, y \notin M$ . Since  $x, y$  are consecutive in  $\pi$ , switching their positions in  $\sigma$  does not affect the invariant that  $I_z^\sigma \cap M \neq \emptyset, \forall z \in \{x, y\}$ .
- $x \notin M, y \in M$ . By definition of  $S, \pi(y) < \pi(x)$  as otherwise  $y$  would belong to  $S$ , and so  $\sigma(y) > \sigma(x)$ . If  $x \neq v$ , then  $x$  cannot belong to  $S$  by definition since  $x$  is dominated by some MIS vertices outside of  $S$ . If  $x = v$ , then  $y \neq u$  as  $\sigma(v) > \sigma(u)$ . Right after the edge update  $x$  is still dominated by a vertex in  $M$ , namely  $y$ , which is also a predecessor in  $\pi$ , so  $S = \emptyset$  which is a contradiction. ■

**Claim 24.**  $S_v^\sigma = S$ .

*Proof:* By the previous claim,  $M$  was also the greedy MIS on  $G$  with respect to order  $\sigma$ . We first argue that  $S_v^\sigma \supseteq S$ . To do this, we prove by an induction that for every  $i \geq 0, S_i \subseteq S_v^\sigma$ ; we refer readers to the definition of influenced sets for the meaning of  $S_i$ , where  $S_i$ 's are defined with respect to permutation  $\pi$ , not  $\sigma$ .

- Basis. For  $i = 0$ , to argue  $v \in S_v^\sigma$  we only need to prove  $S_v^\sigma \neq \emptyset$ . As  $S_v^\pi \neq \emptyset$ , the edge update can only be an insertion  $(u, v)$  and  $u, v \in M$ , or an edge deletion  $(u, v)$  and  $u \in M, v \notin M$  plus that  $u$  is the only MIS predecessor that dominates  $v$ . Since  $\sigma$  and  $\pi$  agree on all vertices whose orders are  $\leq \pi(v)$ ,  $v$  would also violate its greedy MIS constraint with respect to  $\sigma$ , and so  $S_v^\sigma \neq \emptyset$ .
- Induction. Suppose we already have  $S_{i-1} \subseteq S_v^\sigma$ . Then, by Lemma 8, any  $z \in M$  such that  $S_{i-1} \cap I_z^\sigma \neq \emptyset$  should belong to  $S_v^\sigma$ . Since  $\pi$  and  $\sigma$  have the same relative order on  $S, S_{i-1} \cap I_z^\sigma$  would be the same as  $S_{i-1} \cap I_z^\pi$  for any  $z \in S_i \cap M$ . On the other hand, for any  $z \in S_i \setminus (M \cup \{v\})$ , we claim  $I_z^\sigma \cap M$  is also equal to  $I_z^\pi \cap M$ . The only possible violation comes from the case that  $z = x$  and  $y \in M$ . However this is also not possible: if  $\pi(y) > \pi(x)$ , then as  $y \notin S$ , by definition when  $x \neq u$ , it would have been excluded from  $S$ , and otherwise if  $x = v$  we would have  $S_v^\pi = \emptyset$ ; if  $\pi(y) < \pi(x)$ , then  $y$  would have been added to  $S$ ; both lead to contradictions. Therefore, by definition of  $S_i$ , we also have  $S_i \subseteq S_v^\sigma$ .

To prove  $S_v^\sigma \subseteq S$ , by Lemma 8 it suffices to verify that (1)  $\forall z \in M, I_z^\sigma \cap S \neq \emptyset$  iff  $z \in S$ ; (2)  $\forall z \notin M, I_z^\sigma \cap M \subseteq S$  iff  $z \in S$ . As  $\sigma$  is equal to  $\pi$  except for  $x, y$ , we only need to consider  $z \in \{x, y\}$  in (1)(2). We can assume  $x, y$  are adjacent; otherwise switching the orders between  $x, y$  in  $\pi$  does not affect the invariant. Then it can never be the case where  $x \notin M, y \in M$  as it would contradict the definition of  $S$ . So it is either  $x \in M, y \notin M$  or  $x, y \notin M$ . Consider two cases.

- $x \in M, y \notin M$ . In this case,  $I_x^\sigma \cap S = \emptyset$  always holds as switching the positions between  $x, y$  does not affect the equality  $I_x^\sigma \cap S = I_x^\pi \cap S \neq \emptyset$ . If  $\pi(y) < \pi(x)$ , then since  $y \notin M$ , it must be  $I_y^\pi \cap M \neq \emptyset$ , and because  $y \notin S$ , there exists  $z \in I_y^\pi \cap M \setminus S$ . So  $\sigma(z) = \pi(z) < \pi(y) = \sigma(x)$ . By the previous claim we already know  $M_\sigma = M$ , and so  $I_z^\sigma \cap M \not\subseteq S$ . If  $\pi(y) > \pi(x)$ , then  $I_z^\sigma \cap S \subseteq I_z^\pi \cap S = \emptyset$ .
- $x, y \notin M$ . Since  $x, y$  are consecutive in  $\pi$ , switching their positions in  $\sigma$  does not change

$I_z^\sigma \cap M, \forall z \in \{x, y\}$ .

- $x \notin M, y \in M$ . By definition of  $S$ ,  $\pi(y) < \pi(x)$  as otherwise  $y$  would belong to  $S$ , and so  $\sigma(y) > \sigma(x)$ . If  $x \neq u$ , then  $x$  cannot belong to  $S$  by definition since  $x$  is dominated by some MIS vertices outside of  $S$ . If  $x = v$ , then  $y \neq u$  as  $\sigma(v) > \sigma(u)$ . Right after the edge update  $x$  is still dominated by a vertex in  $M$ , namely  $y$ , which is also a predecessor in  $\pi$ , so  $S = \emptyset$  which is a contradiction. ■

### B. Proof of Lemma 10

*Proof:* As the lemma is stated in a slightly different way from [6], for completeness we also present a proof here. Define an intermediate permutation  $\tau$  by this operation: remove  $v$  from order  $\sigma$  and reinsert it back right after  $u$ . Then  $\tau(u) < \tau(v), \tau_S = \pi_S, \tau_{V \setminus S} = \pi_{V \setminus S}$ , and thus by Lemma 9 we have  $S_v^\tau = S$ . Namely,  $\tau$  and  $\pi$  satisfy the same condition in the statement of the lemma.

Let  $w = \arg \min_{x \in S \setminus \{v\}} \{\tau(x)\}$ . First we argue that  $w$  and  $v$  are neighbors. If  $w$  was in  $M_\tau$ , then by the inductive definition of  $S_v^\tau$ , there exists  $z \in S \setminus M_\tau$  such that  $z$  is a predecessor neighbor of  $w$ . By minimality of  $w$ ,  $z$  must be equal to  $v$ , and hence  $w$  and  $v$  are adjacent. If  $w$  was not in  $M_\tau$ , then it has an MIS predecessor  $z \in S \cap M_\tau$ , similarly by minimality of  $w$ ,  $z$  must be equal to  $v$ , and hence  $w$  and  $v$  are adjacent.

Recalling the relation between  $\tau$  and  $\sigma$ , we can view  $\sigma$  as a permutation derived from  $\tau$  by first removing  $v$  from  $\tau$  and then reinsert  $v$  back to  $\tau$  at a certain position somewhere behind  $w$ . We claim that right after we remove  $v$  from  $\tau$  before reinsertion,  $w$  belongs to the greedy MIS  $M_\tau$  with respect to the current  $\tau$  (which is without  $v$ ). Consider the only two cases where  $S_v^\tau$  could be nonempty.

- The edge update is an insertion and both of  $u, v$  were in  $M_\tau$ . After the removal of  $v$ ,  $w$  is no longer dominated by any MIS predecessor in  $M_\tau$ , hence  $w$  must join  $M_\tau$ .
- The edge update is a deletion, and  $u$  was in  $M_\tau$  while  $v$  was not in  $M_\tau$ , plus that  $u$  is the only MIS predecessor that dominates  $v$ . Since  $v$  was not in  $M_\tau$ , then by minimality of  $\tau(w)$  among  $S \setminus \{v\}$ , the only predecessor of  $w$  in  $S$  was  $v$ , and thus  $w \in M_\tau$  before and after  $v$ 's removal.

When we insert  $v$  back to  $\tau$  at some position after  $w$ , which produces permutation  $\sigma$ , since  $w$  is now an MIS predecessor of  $v$ ,  $v$  does not belong to  $M_\sigma$ . If the edge

update is insertion then no changes would be made to  $M_\sigma$  and thus  $S_v^\sigma = \emptyset$ ; if the edge update is deletion, then since  $v$  has a neighboring MIS predecessor other than  $u$ , which is  $w$ ,  $M_\sigma$  would also stay unchanged, and thus  $S_v^\sigma = \emptyset$ . ■

### C. Proof of Lemma 11

*Proof:* For notational convenience, define  $\mathcal{E} = \{\pi(u) = A, \pi(v) \in [A + 1, B]\}$ . For any vertex set  $S \subseteq V \setminus \{u_j\}_{1 \leq j \leq a}$  containing  $v$ , and partial orders  $\sigma_+, \sigma_-$  on  $S \setminus \{v\}$  and  $V \setminus S$ , with the property that there exists at least one permutation  $\pi$  that satisfies event  $\mathcal{E}$ , as well as  $S_v^\pi = S, \pi_{S \setminus \{v\}} = \sigma_+, \pi_{V \setminus S} = \sigma_-$ , define a set of permutations

$$\Omega_{S, \sigma_+, \sigma_-} = \{\pi \mid \mathcal{E}, \pi_{S \setminus \{v\}} = \sigma_+, \pi_{V \setminus S} = \sigma_-\}$$

By Lemma 9 and Lemma 10, for any  $\pi \in \Omega_{S, \sigma_+, \sigma_-}$ ,  $S_v^\pi = S$  when  $\pi(v) = \min_{z \in S} \{\pi(z)\}$ , and  $S_v^\pi = \emptyset$  otherwise. Here is a basic property of  $\Omega_{S, \sigma_+, \sigma_-}$ .

**Claim 25.** For any two different  $\Omega_{S, \sigma_+, \sigma_-} = \{\pi \mid \mathcal{E}, \pi_{S \setminus \{v\}} = \sigma_+, \pi_{V \setminus S} = \sigma_-\}$  and  $\Omega_{S', \sigma'_+, \sigma'_-} = \{\pi \mid \mathcal{E}, \pi_{S' \setminus \{v\}} = \sigma'_+, \pi_{V \setminus S'} = \sigma'_-\}$ ,  $\Omega_{S, \sigma_+, \sigma_-}$  and  $\Omega_{S', \sigma'_+, \sigma'_-}$  are disjoint.

*Proof:* Suppose otherwise there exists  $\tau \in \Omega_{S, \sigma_+, \sigma_-} \cap \Omega_{S', \sigma'_+, \sigma'_-}$ . By definition, there exists  $\pi \in \Omega_{S, \sigma_+, \sigma_-}$  that satisfies event  $\mathcal{E}$ , as well as  $S_v^\pi = S, \pi_{S \setminus \{v\}} = \sigma_+, \pi_{V \setminus S} = \sigma_-$ . By Lemma 10,  $v$  takes the minimum in  $\pi$  among  $S$ .

Remove  $v$  from  $\tau$  and reinsert  $v$  back to  $\tau$  right at position  $A + 1$ . We claim  $\tau$  stays in  $\Omega_{S, \sigma_+, \sigma_-} \cap \Omega_{S', \sigma'_+, \sigma'_-}$ ; this is because removal and reinsertion of  $v$  preserves  $\tau$ 's induced order on  $S \setminus \{v\}, V \setminus S$  and  $S' \setminus \{v\}, V \setminus S'$ . Now, since  $v$  takes the minimum among  $S$  in  $\tau$ , we have  $\tau_S = \pi_S, \tau_{V \setminus S} = \pi_{V \setminus S}$ . By Lemma 9,  $S_v^\tau = S_v^\pi = S$ . Similarly we can also have  $S_v^\tau = S'$ . Therefore,  $S = S'$ . As  $\tau \in \Omega_{S', \sigma'_+, \sigma'_-}$ , we know immediately  $\sigma_+ = \tau_S = \tau_{S'} = \sigma'_+, \sigma_- = \tau_{V \setminus S} = \tau_{V \setminus S'} = \sigma'_-$ , which is a contradiction that  $\Omega$  and  $\Omega'$  are different. ■

By this claim, we can decompose the expectation as a sum of conditional ones:

$$\begin{aligned} & \mathbb{E}_\pi[|S_v^\pi| \mid \mathcal{E}] \\ &= \sum_{S, \sigma_+, \sigma_-} \Pr_\pi[\pi \in \Omega_{S, \sigma_+, \sigma_-} \mid \mathcal{E}] \cdot \mathbb{E}_\pi[|S_v^\pi| \mid \mathcal{E}, \pi \in \Omega_{S, \sigma_+, \sigma_-}] \end{aligned}$$

So it suffices to compute each term in the summation. Fix any  $S, \sigma_+, \sigma_-$  and  $\Omega = \Omega_{S, \sigma_+, \sigma_-}$ . Notice that by Lemma 9 and Lemma 10 we have:

$$\mathbb{E}_\pi[|S_v^\pi| \mid \mathcal{E}, \pi \in \Omega] = |S| \cdot \Pr_\pi[\pi(v) = \min_{z \in S} \{\pi(z)\} \mid \mathcal{E}, \pi \in \Omega]$$

To bound the probability  $\Pr_{\pi}[\pi(v) = \min_{z \in S} \{\pi(z)\} \mid \mathcal{E}, \pi \in \Omega]$ , on the one hand, any permutation  $\pi \in \Omega$  can be constructed by picking an arbitrary position for  $v$  among  $[A + 1, B]$ , and then assign arbitrary positions for  $S \setminus \{v\}$ , so  $|\Omega| = (B - A) \cdot \binom{n-A-1}{|S|-1}$ . On the other hand, the total number of permutations such that  $v$  takes the minimum among  $S$  is  $\binom{n-A}{|S|} - \binom{n-B}{|S|}$ . Therefore, as  $\pi$  is uniformly drawn from  $\Omega$ , we have:

$$\begin{aligned} & \Pr_{\pi}[\pi(v) = \min_{z \in S} \{\pi(z)\} \mid \mathcal{E}, \pi \in \Omega] \\ &= \frac{\binom{n-A}{|S|} - \binom{n-B}{|S|}}{(B-A) \cdot \binom{n-A-1}{|S|-1}} \\ &= \frac{\binom{n-A}{|S|} - \binom{n-B}{|S|}}{(B-A) \cdot \binom{n-A}{|S|} \cdot \frac{|S|}{n-A}} < \frac{n-A}{(B-A)|S|} \end{aligned}$$

Hence,  $\mathbb{E}_{\pi}[|S_v^{\pi}| \mid \mathcal{E}, \pi \in \Omega] = |S| \cdot \Pr_{\pi}[\pi(v) = \min_{z \in S} \{\pi(z)\} \mid \mathcal{E}, \pi \in \Omega] < \frac{n-A}{B-A}$ . Since all  $\Omega$  are disjoint, ranging over all different choices for  $S, \sigma_+, \sigma_-$ , we have

$$\begin{aligned} & \mathbb{E}_{\pi}[|S_v^{\pi}| \mid \pi(u) = A, \pi(v) \in [A+1, B]] \\ & < \frac{n-A}{B-A} < \frac{n}{B-A} \end{aligned}$$

■

#### D. Proof of Lemma 12

*Proof:* For  $u, v$  to both lie in  $[A + 1, B]$ ,  $B$  must be larger than  $A + 1$ . For notational convenience, define  $\mathcal{E} = \{A < \pi(u) < \pi(v) \leq B\}$ . We decompose the expectation as:

$$\begin{aligned} & \mathbb{E}_{\pi}[|S_v^{\pi}| \mid \mathcal{E}] \\ &= \sum_{k=A+1}^{B-1} \Pr_{\pi}[\pi(u) = k \mid \mathcal{E}] \cdot \mathbb{E}_{\pi}[|S_v^{\pi}| \mid \mathcal{E}, \pi(u) = k] \\ &= \sum_{k=A+1}^{B-1} \frac{B-k}{\binom{B-A}{2}} \cdot \mathbb{E}_{\pi}[|S_v^{\pi}| \mid \mathcal{E}, \pi(u) = k] \end{aligned}$$

The second equality holds as  $\Pr_{\pi}[\pi(u) = k \mid \mathcal{E}] = \frac{B-k}{\binom{B-A}{2}}$ ; this is because, conditioned on  $\pi(u) = k$  as well as event  $\mathcal{E}$ , there are  $(B-k) \cdot (n-A-2)!$  permutations  $\pi$ , while there are  $\binom{B-A}{2} \cdot (n-A-2)!$  permutations  $\pi$  that satisfy event  $\mathcal{E}$ . Since  $\pi$  is drawn uniformly at random from the set of all permutations that satisfy event  $\mathcal{E}$ , we have  $\Pr_{\pi}[\pi(u) = k \mid \mathcal{E}] = \frac{B-k}{\binom{B-A}{2}}$ .

Using Lemma 11, we have:

$$\mathbb{E}_{\pi}[|S_v^{\pi}| \mid \mathcal{E}, \pi(u) = k] \leq \frac{n}{B-k}$$

Therefore,

$$\begin{aligned} \mathbb{E}_{\pi}[|S_v^{\pi}| \mid \mathcal{E}] &= \sum_{k=A+1}^{B-1} \frac{B-k}{\binom{B-A}{2}} \cdot \mathbb{E}_{\pi}[|S_v^{\pi}| \mid \mathcal{E}, \pi(u) = k] \\ &< \sum_{k=A+1}^{B-1} \frac{B-k}{(B-A-1)(B-A)/2} \cdot \frac{n}{B-k} < \frac{2n}{B-A} \end{aligned}$$

■