

# Fast Maintenance of 2-hop Labels for Shortest Distance Queries on Fully Dynamic Graphs

## — Supplemental Materials

The road map of this supplement is as follows.

- In Section S1, we provide detailed discussions on complexities of algorithms.
- In Section S2, we prove that CLN can eliminate all redundant indexes.

### S1. THE DISCUSSIONS ON COMPLEXITIES OF ALGORITHMS

**The complexities of FastDeM:** FastDeM has a time complexity of

$$O(\delta^2 + Y \cdot (\log Y + d_a \cdot \delta)).$$

The reason is that populating  $CL^c$  takes  $O(\delta^2)$  time, and *DIFFUSE* takes  $O(Y \cdot (\log Y + d_a \cdot \delta))$  time, assumed that labels are stored in hashes. Particularly, *DIFFUSE* pops  $O(Y)$  elements out of the priority queue. Each pop operation takes  $O(\log Y)$  times. After each pop, it searches  $O(d_a)$  neighbors, and a distance query that costs  $O(\delta)$  may be conducted in each search.

Furthermore, like DeAsyn, FastDeM has a space complexity of  $O(|L| + |PPR|)$ , which is  $O(|V|^2)$  in the worst case, *e.g.*, when  $G$  is a complete graph in which each shortest path is a single edge. Nevertheless, our experiments show that FastDeM scales to large graphs in practice.

**The complexities of FastInM:** FastInM has a time complexity of

$$O\left(Y \cdot \left(\log Y + d_a \cdot \delta + \kappa \cdot (d_a + \delta)\right)\right),$$

where  $\kappa$  is the average number of *PPR* elements of each vertex-hub pair. The details are as follows. First, FastInM pushes labels into  $AL_1$  in  $O(\delta)$  time. Then, it performs *SPREAD*<sub>1</sub> in  $O(Y \cdot d_a)$  time, since there are  $O(Y)$  labels deactivated, and each deactivation is followed by  $O(d_a)$  neighbor searches. Subsequently, it performs *SPREAD*<sub>2</sub> in  $O(Y \cdot \kappa \cdot (d_a + \delta))$  time, since for each of  $O(Y)$  tuples in  $AL_2$ , it checks  $O(\kappa)$  *PPR* elements, while checking each *PPR* element takes  $O(d_a + \delta)$  time, *e.g.*, it compares  $O(d_a)$  values in Line 17 and performs the query in  $O(\delta)$  time in Line 18 [1]. After that, it performs *SPREAD*<sub>3</sub> in  $O(Y \cdot (\log Y + d_a \cdot \delta))$  time, just like *DIFFUSE* in FastDeM. Furthermore, like the analyses of FastDeM, FastInM has a space complexity of  $O(|L| + |PPR|)$ .

**The complexities of CLN and Algorithm 1:** We eliminate redundant indexes from time to time such that redundant indexes do not become orders of magnitude more than non-redundant ones. Then, we ignore the change of  $\delta$  in the following analyses.

Like the above analyses, both CLN and Algorithm 1 have a space complexity of  $O(|L| + |PPR|)$ . We explain that Algorithm 1 has a time complexity of

$$O(|E| \cdot \delta^2 + |E| \cdot \delta \cdot \log |V|)$$

as follows. For each generated label associated with a vertex  $v \in V$ , Algorithm 1 inserts  $O(\deg(v))$  elements into  $Q$ . There are  $O(|E| \cdot \delta)$  elements in  $Q$  in total. For each element in  $Q$ , Algorithm 1 takes  $O(\log |V|)$  time to pop it out, and also takes  $O(\delta)$  time to query a distance. In comparison, CLN has a time complexity of

$$O(|V| \cdot \delta^2 + |E| \cdot \delta \cdot \log |V|).$$

First, cleaning  $L$  takes  $O(|V| \cdot \delta^2)$ , since there are  $O(|V| \cdot \delta)$  labels, and checking whether a label is redundant or not using a distance query takes  $O(\delta)$ . Second, re-generating *PPR* takes  $O(|E| \cdot \delta \cdot \log |V| + |V| \cdot \delta^2)$ , since this process is similar to the process of Algorithm 1 with  $O(|PPR|)$  distance queries, and we generally have  $|PPR| \ll |L| = |V| \cdot \delta$ . Moreover, since we often have  $|E| \gg |V|$  and  $\delta \gg \log |V|$ , it can be considered that CLN has a smaller time complexity than Algorithm 1.

## S2. THE EFFECTIVENESS OF CLN

We show that CLN can eliminate all redundant indexes as follows. First, we present the canonical constraint [2–4] below.

**Definition 1** (Canonical Constraint). *Given a rank of vertices, a set  $L$  of 2-hop labels satisfies the canonical constraint if, a vertex  $v$  is a hub of  $u \in V$ , i.e.,  $v \in C(u)$ , if and only if the rank of  $v$  is the highest among all vertices in all shortest paths between  $u$  and  $v$ .*

For a given rank of vertices, there is only one set of 2-hop labels that satisfies the canonical constraint, e.g.,  $L$  in Figure 1 in the main contents. We refer to a set of 2-hop labels that satisfies the canonical constraint as a canonical set of 2-hop labels, which is minimal in that deleting any label from this set induces that it does not satisfy the 2-hop cover constraint. PLL is a widely-used algorithm for generating a canonical set of 2-hop labels [1, 5].

Suppose that the initial indexes are generated by Algorithm 1. To maintain 2-hop labels after edge weight changes, both InAsyn+RepairedDeAsyn and FastInM+FastDeM generate a new label  $L(u)[v]$  only when  $r(u) < r(v)$ . Let  $L_m$  be the maintained set of 2-hop labels by InAsyn+RepairedDeAsyn or FastInM+FastDeM after a number of edge weight changes. With the input of  $L_m$ , let  $L_c$  and  $PPR_c$  be the cleaned set of 2-hop labels and the cleaned  $PPR$ , respectively, by CLN. Further let  $L_r$  and  $PPR_r$  be the re-generated indexes by Algorithm 1 after the edge weight changes.  $L_r$  is a canonical set of 2-hop labels for the updated graph, and  $PPR_r$  is the record of the pruning information for generating  $L_r$  by PLL. We have the following theorem, which shows that CLN is as effective as Algorithm 1 for eliminating redundant indexes.

**Theorem 3.**  $L_c = L_r$ ,  $PPR_c = PPR_r$ .

*Proof.* First, we prove that  $L_r \subseteq L_m$  as follows. Consider an arbitrary label  $(u', d'_{u's}) \in L_r(s)$ . Since  $L_r$  is a canonical set of 2-hop labels for the updated graph,  $u'$  is the vertex with the highest rank in all shortest paths between  $s$  and  $u'$  on the updated graph, and  $d'_{u's}$  is the distance between  $s$  and  $u'$  on the updated graph. The proofs of Theorems 1-2 show that the vertex with the highest rank in all shortest paths between  $s$  and another vertex on the updated graph is a hub of  $s$  after each edge weight decrease or increase maintenance. Thus,  $(u', d'_{u's}) \in L_m(s)$ , and  $L_r \subseteq L_m$ .

Subsequently, consider an arbitrary label  $(v, d_{uv}) \in L_r(u) \subseteq L_m(u)$ . When CLN computes  $d'_{uv}$  in Line 4, if  $d'_{uv} \leq d_{uv}$ , then there is a vertex  $y \in C_{>r(v)}(u) \cap C(v)$  that is in a shortest path between  $u$  and  $v$ , and  $r(y) > r(v)$ . However, since  $L_r$  is canonical, this contradicts with the fact that the rank of  $v$  is the highest among all vertices in all shortest paths between  $u$  and  $v$ . Thus,  $d'_{uv} > d_{uv}$ , and CLN inserts  $(v, d_{uv})$  into  $L_c(u)$ , i.e.,  $(v, d_{uv}) \in L_c(u)$ . On the other hand, consider an arbitrary label  $(x, d_{ux}) \in L_m(u) \setminus L_r(u)$ . Let  $z \in V$  be the vertex with the highest rank among all vertices in all shortest paths between  $u$  and  $x$ . We have  $r(z) > r(x)$ , and  $z \in C_{>r(x)}(u) \cap C(x)$ . As a result, CLN computes  $d'_{ux} = d_{ux} = d(u, z) + d(z, x)$ , and does not insert  $(x, d_{ux})$  into  $L_c(u)$ , i.e.,  $(x, d_{ux}) \notin L_c(u)$ . Thus,  $L_c = L_r$ . Moreover, since  $PPR_r$  is the record of the pruning information for generating  $L_r$  by PLL; and  $PPR_c$  is the record of the pruning information for generating  $L_c$  by PLL (notably, the process of generating  $PPR_c$  in CLN can be considered as an accelerated version of the process of generating  $PPR$  in Algorithm 1),  $PPR_c = PPR_r$ . Hence, this theorem holds.  $\square$

## REFERENCES FOR THE SUPPLEMENT

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