## 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

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

The reason is that populating  $CL^c$  takes  $O(\delta^2)$  time, and DIFFUSE takes  $O(\Upsilon \cdot (\log \Upsilon + d_a \cdot \delta))$  time, assumed that labels are stored in hashes. Particularly, DIFFUSE pops  $O(\Upsilon)$  elements out of the priority queue. Each pop operation takes  $O(\log \Upsilon)$  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, experiments show that FastDeM scales to large graphs in practice.

The complexities of FastInM: FastInM has a time complexity of

$$O(\Upsilon \cdot (\log \Upsilon + d_a \cdot \delta + \kappa \cdot (d_a + \delta))),$$

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  $SPREAD_1$  in  $O(\Upsilon \cdot d_a)$  time, since there are  $O(\Upsilon)$  labels deactivated, and each deactivation is followed by  $O(d_a)$  neighbor searches. Subsequently, it performs  $SPREAD_2$  in  $O(\Upsilon \cdot \kappa \cdot (d_a + \delta))$  time, since for each of  $O(\Upsilon)$  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  $SPREAD_3$  in  $O(\Upsilon \cdot (\log \Upsilon + d_a \cdot \delta))$  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

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

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.

CLN has the same time complexity. The details are as follows. 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  $O(|E| \cdot \delta \cdot \log |V| + |E| \cdot \delta^2)$ , since this process is similar to the process of Algorithm 1 with O(|PPR|) distance queries, and  $|PPR| = O(|E| \cdot \delta)$ , while each edge may correspond to  $O(\delta)$  pruning operations in PLL.

## 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

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  $d'_{uv} \le d_{uv}$ , 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 v. We have v be the vertex with the highest rank among all vertices in all shortest paths between v and v and v does not insert v

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