# Generation and Maintenance of 2-hop Labels for Accelerating Group Steiner Tree Search on Graphs

The road map of this supplement is as follows.

- In Section S1, we provide proofs that are omitted in the main contents.
- In Section S2, we provide detailed discussions on complexities of algorithms.

## S1. PROOFS OF THEOREMS

We demonstrate theorems and proofs as follows.

**Theorem 1.** A set of 2-hop labels that satisfies the GST-customized 2-hop cover constraint supports the query of a shortest path between every pair of a vertex  $v \in V$  and a candidate vertex group  $g \in \Gamma_{all}$ .

*Proof.* We show that we can query the shortest distance/path between every pair of a vertex  $v \in V$  and a candidate vertex group  $g \in \Gamma_{all}$  using a set L of 2-hop labels that satisfies the above constraint as follows. There are two different cases:

- 1. Case 1: there is a path between v and g on the original graph. Since each mock edge weight M is larger than the total weight of all non-mock edges, a shortest path between v and  $v_g$  on the transformed graph contains a shortest path between v and g on the original graph. Since  $|\{v,v_g\}\cap D|<2$ , there is a common hub vertex  $u\in C(v)\cap C(v_g)$  in a shortest path between v and  $v_g$ . Thus, we can use L to query the shortest distance between v and  $v_g$ , which is smaller than 2M and equals the shortest distance between v and g on the original graph plus M. Hence, equivalently, we can use L to query the shortest distance between v and g. Furthermore, since there is no mock vertex in the middle of a shortest path between v and  $v_g$  on the transformed graph, we can also iteratively query the shortest path between v and  $v_g$  on the original graph by adding predecessors into labels.
- 2. Case 2: there is no path between v and g on the original graph. Thus, v and  $v_g$  are not properly connected, and there may not be a common hub vertex  $u \in C(v) \cap C(v_g)$ . If there is no common hub vertex  $u \in C(v) \cap C(v_g)$ , then we can directly deduce that there is no path between v and g on the original graph. If there is a common hub vertex  $u \in C(v) \cap C(v_g)$ , then since v and  $v_g$  are not properly connected on the transformed graph, there must be at least one mock vertex in the middle of any path between v and  $v_g$ , and as a result the queried distance of any path between v and  $v_g$  is no smaller than 2M, which still indicates that there is no path between v and v on the original graph.

Hence, this theorem holds.

**Theorem 2.** A set of 2-hop labels that has the GST-customized canonical property is minimal for meeting the GST-customized 2-hop cover constraint.

*Proof.* Consider a set *L* of 2-hop labels that has the GST-customized canonical property. For every pair of properly connected vertices  $v_i$  and  $v_j$  such that  $|\{v_i, v_j\} \cap D| < 2$ , let  $v_k$  be the vertex with the highest rank in all shortest paths between  $v_i$  and  $v_j$ . We have  $v_k \notin D$  and  $v_k \in C(v_i) \cap C(v_j)$ , since  $v_k$  has the highest rank in all shortest paths between  $v_i$  and  $v_k$ , as well as between  $v_j$  and  $v_k$ , and no vertices in these shortest paths can be hubs of  $v_k$ , except itself. Thus, *L* meets the GST-customized 2-hop cover constraint. Moreover, consider an arbitrary label  $(v_i, d_{v_i v_j}) \in L(v_j)$ . We have (i)  $v_i$  and  $v_j$  are properly connected; (ii)  $|\{v_i, v_j\} \cap D| < 2$ ; and (iii) the rank of  $v_i$  is the highest among all vertices in all shortest paths between  $v_i$  and  $v_j$ . Thus, there is no other vertex  $v_k \in C(v_i) \cap C(v_j)$  in a shortest path between  $v_i$  and  $v_j$ , *i.e.*, deleting this arbitrary label makes *L* not satisfy the GST-customized 2-hop cover constraint any more. Hence, this theorem holds. □

**Theorem 3.** The set L of 2-hop labels generated by HL4GST has the GST-customized canonical property, and thus is minimal for meeting the GST-customized 2-hop cover constraint.

*Proof.* Let  $L^{can}$  be the set of labels that has the GST-customized canonical property. We prove that every label in  $L^{can}$  is also in L as follows.

Consider an arbitrary label  $(u, d_{vu}) \in L^{can}(v)$ . We observe that (i) u and v are properly connected; (ii)  $|\{u, v\} \cap D| < 2$ ; and (iii) the rank of u is the highest among all vertices in all

shortest paths between u and v. Thus, we have  $d_{vu} < 2M$ , and the generation of the above label cannot be pruned by GST-customized pruning technique in Line 11 of HL4GST. Further consider the labeling process for hub u in HL4GST. HL4GST generates labels with hub u by spreading u to other vertices, starting from u, via a Dijkstra-style process. Suppose that the spread of hub u to v along a shortest path between u and v stops at a middle vertex v' due to the query pruning technique in Line 6 of HL4GST. This means that, before inserting  $(u, d_{v'u})$  into L'(v'), the queried distance between u and v' is no larger than  $d_{v'u}$ . As a result, there must be a common hub vertex  $z \in C(u) \cap C(v')$  such that z is in a shortest path between u and v', and r(z) > r(u). This contradicts with the fact that the rank of u is the highest among all vertices in all shortest paths between u and v'. Consequently, the spread of hub u to v along a shortest path between u and v cannot stop at a middle vertex, and as a result HL4GST inserts  $(u, d_{vu})$  into L'(v).

We further show that HL4GST also inserts  $(u, d_{vu})$  into L(v) as follows. When it checks  $(u, d_{vu}) \in L'(v)$  in Line 18 during the cleaning process, it computes d'(u, v) using  $L'_{>r(u)}(v)$  and L'(u). If  $d'(u, v) \leq d_{vu}$ , then there must be a common hub vertex  $z \in C(u) \cap C(v)$  such that z is in a shortest path between u and v, and r(z) > r(u). This contradicts with the fact that the rank of u is the highest among all vertices in all shortest paths between u and v. Consequently,  $d'(u, v) > d_{vu}$  and HL4GST also inserts  $(u, d_{vu})$  into L(v). Hence, every label in  $L^{can}$  is also in L.

We further prove that every label not in  $L^{can}$  is not in L as follows. Suppose that there is a label  $(u, d_{vu}) \in L(v) \setminus L^{can}(v)$ . We observe that (i) u and v are not properly connected; or (ii)  $|\{u,v\}\cap D|=2$ ; or (iii) the rank of u is not the highest among all vertices in all shortest paths between u and v. If u and v are not properly connected or  $|\{u,v\}\cap D|=2$ , then  $d_{vu}\geq 2M$ , and the generation of the above label will be pruned by GST-customized pruning technique in Line 11 of HL4GST, which means that  $(u,d_{vu})\notin L(v)$ . Otherwise, we consider the case where the rank of u is not the highest among all vertices in all shortest paths between u and v as follows. Let z be the vertex with the highest rank in all shortest paths between u and v. We have  $(z,d_{zv})\in L'_{>r(u)}(v)$  and  $(z,d_{zu})\in L'(u)$ . As a result, HL4GST computes d'(u,v) as a value no larger than  $d_{vu}$  in Line 19, and does not insert  $(u,d_{vu})$  into L. Hence, every label not in  $L^{can}$  is not in L. In conclusion, we have  $L^{can}=L$ . By Theorem 2, this theorem holds.

**Theorem 4.** Given a graph and a corresponding set of 2-hop labels that satisfies the GST-customized 2-hop cover constraint, after a batch of edge insertions and edge weight decreases, the maintained set of labels by BatchM4GST-Insert satisfies the above constraint for the updated graph.

*Proof.* Consider a pair of vertices s and t. We use p'(s,t) to denote a shortest path between s and t after the change. Let u' be the vertex with the highest rank in all shortest paths between s and t after the change, and  $u' \in p'(s,t)$ . Also let d'(s,t) be the shortest distance between s and t after the change. Suppose that  $|\{s,t\} \cap D| < 2$  and d'(s,t) < 2M, *i.e.*, s and t are properly connected after the change.

Let  $E_{in} = \{(a_1, b_1), ..., (a_i, b_i)\}$  be the set of changed edges on p'(s, t), i.e.,  $E_{in} \subseteq E_c$ . Notably,  $E_{in}$  could be empty. Without loss of generality, suppose that  $a_t$  is closer to s than  $b_t$  along p'(s, t) for each  $t \in [1, i]$ , and u' is between  $(a_k, b_k)$  and  $(a_{k+1}, b_{k+1})$ , as illustrated in Figure S1. Since there are only edge insertions and edge weight decreases,  $p'(u', b_k)$  is a shortest path between u' and  $b_k$  both before and after the change, and u' has the highest rank in all shortest paths between u' and  $b_k$  both before and after the change. We also have  $|\{u', b_k\} \cap D| < 2$  and  $d'(u', b_k) < 2M$ .

Suppose that u' is not a hub of  $b_k$  before the maintenance. Since the given set of 2-hop labels satisfies the GST-customized 2-hop cover constraint, to correctly query  $d'(u',b_k)$ , there must be a vertex z such that  $z \in C(u') \cap C(b_k)$ , r(z) > r(u'), and z is in a shortest path between u' and  $b_k$ . This contradicts with the assumption that u' has the highest rank in all shortest paths between u' and  $b_k$ . Thus,  $(u',d'(b_k,u')) \in L(b_k)$ , and further  $(u',d'(a_{k+1},u')) \in L(a_{k+1})$ , before the maintenance. BatchM4GST-Insert calls DIFFUSE to re-spread hub u' from  $b_k$  to s, and from  $a_{k+1}$  to t, along p'(s,t). Thus,  $(u',d'(s,u')) \in L(s)$ , and similarly  $(u',d'(t,u')) \in L(t)$ , after the maintenance. Hence, for every pair of properly connected vertices  $v_i$  and  $v_j$  such that  $|\{v_i,v_j\}\cap D|<2$ , there is a common hub vertex  $u\in C(v_i)\cap C(v_j)$  on a shortest path between  $v_i$  and  $v_j$ , i.e., the set of labels maintained by BatchM4GST-Insert still satisfies the GST-customized 2-hop cover constraint. This theorem holds.

**Theorem 5.** Given a graph and a corresponding set of 2-hop labels that satisfies the GST-customized 2-hop cover constraint, after a batch of edge deletions and edge weight increases, the maintained set of labels by BatchM4GST-Delete satisfies the above constraint for the updated graph.

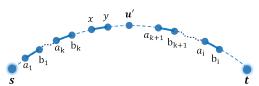


Fig. S1. An illustration for the correctness proofs of the proposed maintenance algorithms.

*Proof.* We use the notations in the proof of Theorem 4. Consider an arbitrary label  $(z, d(z, x)) \in L(x)$  before the change. If this label corresponds to a path that passes through a changed edge, *i.e.*,  $d'(z, x) \ge d(z, x)$ , then BatchM4GST-Delete must call  $SPREAD_1$  to set L(x)[z] to  $\infty$ .

Subsequently, consider a pair of vertices s and t such that  $|\{s,t\} \cap D| < 2$  and d'(s,t) < 2M, i.e., s and t are properly connected after the change. We use p'(s,t) to denote a shortest path between s and t after the change. Let u' be the vertex with the highest rank in all shortest paths between s and t after the change, and  $u' \in p'(s,t)$ .

Before the maintenance, let y be the vertex farthest to u' along  $p'(u', b_k)$  such that u' is a hub of y, and the corresponding label-contained distance value is d'(u', y). Let x be the neighbor of y along  $p'(u', b_k)$  such that u' is not a hub of x, or u' is a hub of x but the corresponding label-contained distance value is not d'(u', x).

If u' is a hub of x but the corresponding label-contained distance value is not d'(u',x) before the maintenance, then we must have d'(u',x) > d(u',x). In this case, BatchM4GST-Delete must call  $SPREAD_1$  to set L(x)[u'] to  $\infty$ , and then call  $SPREAD_2$  and  $SPREAD_3$  to re-spread hub u' from y to x, and ultimately to s, along p'(s,t), i.e., to generate a new label  $(u',d'(u',s)) \in L(s)$ .

If u' is not a hub of x before the maintenance, then there must be a vertex z such that  $z \in C(u') \cap C(x)$ , r(z) > r(u'), and z is in a shortest path between u' and x, and  $u' \in PPR[x,z]$  and  $x \in PPR[u',z]$  before the maintenance. Since u' is the vertex with the highest rank in all shortest paths between u' and x after the change, z is not in a shortest path between u' and x after the change. To meet this condition, either L(u')[z] or L(x)[z] increases after the graph change. Thus, BatchM4GST-Delete must call  $SPREAD_1$  to set either L(u')[z] or L(x)[z] to  $\infty$ . In either case, using the above PPR, it performs  $SPREAD_2$  and  $SPREAD_3$  to re-spread hub u' from y to x, and ultimately to x, along x along x in generate a new label x in x in

Therefore, in either case, we have  $(u',d'(u',s)) \in L(s)$ , and similarly  $(u',d'(u',t)) \in L(t)$  after the maintenance. Hence, for every pair of properly connected vertices  $v_i$  and  $v_j$  such that  $|\{v_i,v_j\}\cap D|<2$ , there is a common hub vertex  $u\in C(v_i)\cap C(v_j)$  on a shortest path between  $v_i$  and  $v_j$ , *i.e.*, the set of labels maintained by BatchM4GST-Delete still satisfies the GST-customized 2-hop cover constraint. This theorem holds.

## S2. COMPLEXITIES OF ALGORITHMS

### A. Complexities of HL4GST

The time complexity of the proposed HL4GST is

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

in a single thread environment, where  $\delta$  is the average number of labels associated with each vertex. The details are as follows. First, the labeling process in Lines 1-14 takes  $O(|E| \cdot \delta \cdot (\delta + \log |V|))$  time. The reason is that, for each label inserted into L'(v) in Line 8, it may insert |N(v)| elements into Q in Line 12, and for each element in Q, it takes  $O(\log |V|)$  time to pop it out in Line 5 (e.g., using Fibonacci heap [1]) and  $O(\delta)$  time to query a distance in Line 6. Second, the sorting process in Lines 15-16 takes  $O(|V| \cdot \delta \cdot \log \delta)$  time. Since we generally have  $|E| \gg |V|$ , the above complexity is covered by that of the labeling process. Third, the cleaning process in Lines 17-21 takes  $O(|V| \cdot \delta^2)$  time, given that a distance query that costs  $O(\delta)$  is required for cleaning each label. The cost of  $O(|V| \cdot \delta^2)$  is also covered by that of the labeling process.

# B. Complexities of BatchM4GST-Insert

The proposed BatchM4GST-Insert has a time complexity of

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

in a single thread environment, where  $\Upsilon$  is the number of update labels, and  $d_a$  is the average degree of vertices that are associated with these labels. The details are as follows. First, populating

CL in Lines 1-18 takes  $O(|E_c| \cdot \delta^2)$  time, since each distance query in Lines 5 and 13 costs  $O(\delta)$ . Second, DIFFUSE takes  $O(\Upsilon \cdot (\log \Upsilon + d_a \cdot \delta))$  time. The reason is that, since  $O(|CL|) = O(\Upsilon)$ , the initialization steps in Lines 21-23 takes  $O(\Upsilon)$  time. Moreover, 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. On the other hand, due to the cost of PPR, the space complexity of BatchM4GST-Insert is  $O(|E| \cdot \delta)$ .

## C. Complexities of BatchM4GST-Delete

The proposed BatchM4GST-Delete has a time complexity of

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

in a single thread environment, where  $\kappa$  is the average number of PPR elements of each vertex-hub pair. The details are as follows. First, it pushes labels into  $AL_1$  in Lines 2-8 in  $O(|E_c| \cdot \delta)$  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 in Line 19, while checking each PPR element takes  $O(d_a + \delta)$  time, due to the cost of  $O(d_a)$  for computing d1(x,t) and d1(t,x) in Lines 21 and 28, and the cost of  $O(\delta)$  for querying distances in Lines 23 and 30. After that, similar to D1FFUSE in BatchM4GST-Insert, it performs  $SPREAD_3$  in  $O(\Upsilon \cdot (\log \Upsilon + d_a \cdot \delta))$  time, given that  $O(|AL_3|) = O(\Upsilon)$ . In the end, due to the cost of PPR, the space complexity of BatchM4GST-Delete is  $O(|E| \cdot \delta)$ .

#### D. Comparison between the proposed BatchM4GST and state-of-the-art methods

As discussed above, the proposed batch-friendly label maintenance framework: BatchM4GST contains a new integrated label update process that takes

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

time. In comparison, the label update process in state-of-the-art algorithms [2-4] takes

$$O(\Upsilon \cdot d_a^2 \cdot \delta)$$

time. Since we generally have

$$d_a^2 \cdot \delta \gg \log \Upsilon + d_a \cdot \delta \tag{S1}$$

in practice, BatchM4GST obtains a smaller time cost than state-of-the-art algorithms [2–4]. We present details of the label update process in state-of-the-art algorithms as follows.

We take DecreaseAsyn in [2] as an example to show the label update process in state-of-the-art algorithms. We refer to DecreaseAsyn as InsertAsyn in the main experiments. It maintains 2-hop labels after an edge weight decrease or an edge insertion, and can be parallelly implemented in batch cases [4]. IncreaseAsyn in [2] that deals with an edge weight increase or an edge deletion has a similar label update process with DecreaseAsyn. These two algorithms, including their parallel versions [4], are state-of-the-art 2-hop label maintenance methods.

We show the pseudo code of DecreaseAsyn as Algorithm S1. Suppose that the weight of an edge  $(a,b) \in E$  decreases from  $w_0(a,b)$  to  $w_1(a,b)$ . The idea of DecreaseAsyn is to update all outdated label-contained distance values that correspond to paths that pass through the old (a,b), first for labels of a and b, and then neighbors of a and b, and then neighbors of neighbors, etc.

DecreaseAsyn first initializes two empty sets:  $CL^c$  and  $CL^n$  (Line 1). It uses  $CL^c$  to record new labels of a and b in Lines 2-17. DecreaseAsyn uses  $CL^c$  to record new labels of b as follows. For each  $(v,d_{va}) \in L(a)$ , if  $r(v) \geq r(b)$  (Line 3), which means that v could be a hub of b, then it checks whether the queried distance between v and b is larger than  $d_{va} + w_1(a,b)$ . If it is, then we can generate a new label  $(v,d_{va}+w_1(a,b)) \in L(b)$ . Thus, it sets  $L(b)[v] = d_{va}+w_1(a,b)$ , and pushes  $(b,v,d_{va}+w_1)$  into  $CL^c$  (Line 5), for iteratively updating more labels in later processes. Otherwise, it conducts the above step when  $v \in C(b)$  &  $L(b)[v] > d_{va}+w_1(a,b)$  (Line 7), and then inserts v into  $PPR[b,h_c]$ , and also inserts b into  $PPR[v,h_c]$ , where b is the common hub responsible for the queried distance between v and b. The condition that  $v \in C(b)$  &  $L(b)[v] > d_{va}+w_1(a,b)$  means that v is already a hub of b, and L(b)[v] is larger than, but should be decreased to,  $d_{va}+w_1(a,b)$ . This step is not in [2], but is necessary for combining DecreaseAsyn and IncreaseAsyn together to deal with fully dynamic cases where edge weights may alternately increase and decrease. After that, it uses  $CL^c$  to record new labels of a similarly (Lines 10-17).

## Algorithm S1. The DecreaseAsyn algorithm

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Input: the original G_0(V,E_0,w_0), the updated G_1(V,E_1,w_1), (a,b), L, PPR Output: the maintained L and PPR
 1: CL^c = CL^n = \emptyset
 2: for each label (v, d_{va}) \in L(a) do
          Fractional (v, w<sub>a</sub>) = -(v, w<sub>a</sub>)

if r(v) \ge r(b) then

if Query(v,b,L) > d_{va} + w_1(a,b) then

L(b)[v] = d_{va} + w_1(a,b), CL^c.push((b,v,d_{va} + w_1(a,b)))
                     if v \in C(b) & L(b)[v] > d_{va} + w_1(a,b) then

L(b)[v] = d_{va} + w_1(a,b), CL^c.push((b,v,d_{va} + w_1(a,b)))
 9.
                     PPR[b,h_c].push(v), PPR[v,h_c].push(b)
10: for each label (v, d_{vb}) \in L(b) do
            if r(v) \ge r(a) then
                 (v) \ge f(a) then
if Query (v, a, L) > d_{vb} + w_1(a, b) then
L(a)[v] = d_{vb} + w_1(a, b), CL^c.push((a, v, d_{vb} + w_1(a, b)))
12:
13:
14:
                       \begin{array}{l} \text{if } v \in C(a) \& L(a)[v] > d_{vb} + w_1(a,b) \text{ then} \\ L(a)[v] = d_{vb} + w_1(a,b), CL^c.push((a,v,d_{vb} + w_1(a,b))) \end{array} 
15:
16:
                      PPR[a,h_c].push(v), PPR[v,h_c].push(a)
17:
18: while CL^c \neq \emptyset do ProDecrease(CL^c, CL^n), CL^c = CL^n, CL^n = \emptyset
19: Return L and PPR
      Procedure ProDecrease(CL^c, CL^n)
20: for each (u, v, d_u) \in CL^c do
21: for each u_n \in N(u) do
22: if r(v) > r(u_n) then
21:
22:
23:
24:
25:
26:
27:
                      if Query(v, u_n, L) > d_{new} = d_u + w_1(u, u_n) then L(u_n)[v] = d_{new}, CL^n.push((u_n, v, d_{new}))
                          if v \in C(u_n) \& L(u_n)[v] > d_{new} then

L(u_n)[v] = d_{new}, CL^n.push((u_n, v, d_{new}))
                           PPR[u_n, h_c].push(v), PPR[v, h_c].push(u_n)
28.
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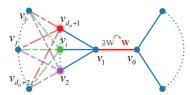


Fig. S2. An example to show the label update process in DecreaseAsyn.

Subsequently, while  $CL^c \neq \emptyset$ , DecreaseAsyn iteratively uses the ProDecrease procedure to update more labels (Line 18).  $CL^c$  and  $CL^n$  are the sets of updated labels in the last and current iterations, respectively. To perform these iterations, it sets  $CL^c = CL^n$  and  $CL^n = \emptyset$  after using ProDecrease in each iteration. In ProDecrease, for each  $(u,v,d_u) \in CL^c$  (Line 20), it checks each neighbor  $u_n$  of u (Line 21). If  $r(v) > r(u_n)$ , which means that v could be a hub of  $u_n$ , then it checks whether the queried distance between v and  $u_n$  is larger than  $d_{new} = d_u + w_1(u,u_n)$  (Line 23). If it is, then we can generate a new label  $(v,d_{new}) \in L(u_n)$ . Thus, it updates  $L(u_n)[v] = d_{new}$ , and pushes  $(u_n,v,d_{new})$  into  $CL^n$  (Line 24). Otherwise, it conducts the above step when  $v \in C(u_n)$  &  $L(u_n)[v] > d_{new}$ , and then inserts v into  $PPR[u_n,h_c]$ , and also inserts  $u_n$  into  $PPR[v,h_c]$  (Line 28), where  $h_c$  is the common hub responsible for the queried distance between v and  $u_n$ . The update of PPR is for maintaining 2-hop labels in later edge weight increase or edge deletion cases. In the end, DecreaseAsyn returns the updated L and PPR (Line 19).

After a change, DecreaseAsyn first updates labels of a and b, and then updates labels of neighbors of a and b, and then neighbors of neighbors etc., until no label needs to be updated. The above label update process, i.e., the iterative call of ProDecrease, takes  $O(\Upsilon \cdot d_a^2 \cdot \delta)$  time. The details are as follows. First, note that, the above label update process updates a label-contained distance value L(x)[v] by spreading hub v from a or b to x in a breadth first search way. This process performs a distance relaxation in Lines 23-24 for each searched edge. As a result, this process may update a label-contained distance value L(x)[v] at most  $O(d_a)$  times in the breadth first search process. Thus, this process could perform  $O(\Upsilon \cdot d_a)$  label update operations. Each label update operation in an iterative call of ProDecrease induces  $O(d_a)$  distance queries in the next iterative call of ProDecrease. Since each distance query takes  $O(\delta)$  time [5, 6], the above label update process takes  $O(\Upsilon \cdot d_a^2 \cdot \delta)$  time. We show an example as follows.

Consider the graph in Figure S2, where  $W \gg \Upsilon \gg d_a^2$ ;  $w_0(v_0, v_1) = 2W$ ;  $w_0(v_1, v_i) = 1$  for

Consider the graph in Figure S2, where  $W \gg \Upsilon \gg d_a^2$ ;  $w_0(v_0, v_1) = 2W$ ;  $w_0(v_1, v_i) = 1$  for each  $i \in [2, d_a + 1]$ ; there is a simple path between  $v_i$  and  $v_j$  for every pair of  $i \in [2, d_a + 1]$  and  $j \in [d_a + 2, \Upsilon]$ , and this path contains i - 1 edges and has a total weight of  $d_a - i + 2$ . Suppose that the weight of  $(v_0, v_1)$  decreases from 2W to W. DecreaseAsyn maintains labels as follows. Initially,

it updates  $L(v_1)[v_0] = W$  and  $L(v_i)[v_0] = W + 1$  for each  $i \in [2, d_a + 1]$ . Subsequently, for a certain  $j \in [d_a + 2, \Upsilon]$ , it sequentially updates  $L(v_j)[v_0] = W + 1 + d_a - i + 2$  through the path between  $v_i$ and  $v_i$  for each  $i \in [2, d_a + 1]$ . As a result, they update label-contained distance values  $O(\Upsilon \cdot d_a)$ times. Since each label update in an iterative call of ProDecrease induces  $O(d_a)$  distance queries in the next iterative call, DecreaseAsyn conducts  $O(\Upsilon \cdot d_a^2)$  distance queries. Since each distance query takes  $O(\delta)$  time, the label update process in DecreaseAsyn costs  $O(\Upsilon \cdot d_a^2 \cdot \delta)$ . In comparison, as discussed before, the integrated label update process in the proposed BatchM4GST has a smaller time cost of  $O(\Upsilon \cdot (\log \Upsilon + d_a \cdot \delta))$ .

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