

L4G: Two-hop Label Management for Group Steiner Tree Search on Graphs – The Supplement

The road map of this supplement is as follows.

- In Section S1, we illustrate candidate group sizes used in the main experiments.
- In Section S2, we provide proofs that are omitted in the main contents.
- In Section S3, we discuss complexities of the proposed algorithms in detail.
- In Section S4, we demonstrate that the label update process in state-of-the-art methods [1, 2] has a larger time cost than the new label update process in L4G-M.

S1. CANDIDATE GROUP SIZES

In Figure S1, we illustrate candidate group sizes used in the main experiments. It can be seen that candidate groups may contain hundreds or thousands of vertices.

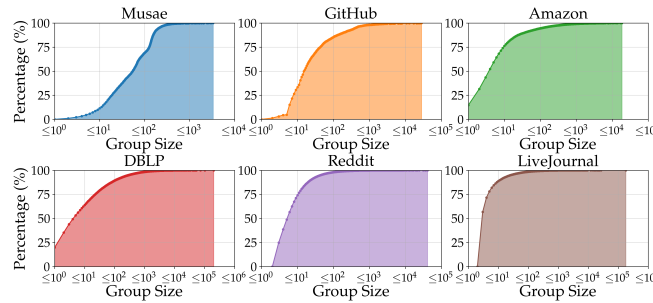


Fig. S1. Candidate group sizes in used real datasets.

S2. PROOFS OF THEOREMS

We demonstrate theorems and proofs as follows.

Theorem 1. *Given an enhanced graph with mock vertices that represent candidate vertex groups, if each mock edge weight M is larger than the total weight of all non-mock edges, then a set of 2-hop labels which 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 $g \in \Gamma_{all}$ on the original graph without mock vertices.*

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 enhanced graph contains (i) a shortest path between v and g on the original graph; and (ii) a mock edge between v_g and a vertex in g on enhanced graph. Since $|\{v, v_g\} \cap D| < 2$, the GST-customized 2-hop cover constraint shows that 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 . Since each mock edge weight M is larger than the total weight of all non-mock edges, the queried shortest distance between v and v_g 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 on the original graph. Furthermore, since there is no mock vertex in the middle of a shortest path between v and v_g on the enhanced graph, we can also iteratively query the shortest path between v and 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 enhanced graph, there must be at least one mock vertex in

the middle of any path between v and v_g , *i.e.*, there are at least two mock edges on any path between v and v_g . As a result, the queried distance of any path between v and v_g is no smaller than $2M$. Since the queried distance in Case 1 is always smaller than $2M$, the above queried distance that is no smaller than $2M$ still indicates that there is no path between v and g on the original graph.

Hence, this theorem holds. \square

Theorem 2. *For an enhanced graph with mock vertices, 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 $v_k \in C(v_i) \cap C(v_j)$, 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. \square

Theorem 3. *Consider an enhanced graph with mock vertices, if each mock edge weight M is larger than the total weight of all non-mock edges, then the set L of 2-hop labels generated by L4G-G 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 . The proof of Theorem 1 shows that, since each mock edge weight M is larger than the total weight of all non-mock edges, we have $d_{vu} < 2M$. Thus, the generation of the above label cannot be pruned by GST-customized pruning technique in Line 11 of L4G-G. Further consider the labeling process for hub u in L4G-G. L4G-G 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 L4G-G. 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 L4G-G inserts (u, d_{vu}) into $L'(v)$.

We further show that L4G-G 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 L4G-G 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, since each mock edge weight M is larger than the total weight of all non-mock edges, we have $d_{vu} \geq 2M$. As a result, the generation of the above label will be pruned by GST-customized pruning technique in Line 11 of L4G-G, 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, L4G-G 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. \square

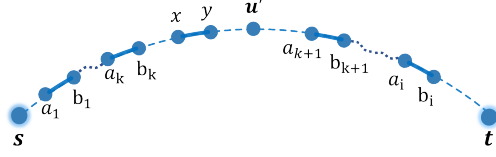


Fig. S2. An illustration for the correctness proofs of the proposed maintenance algorithms.

Theorem 4. *Consider an enhanced graph with mock vertices, and a corresponding set of 2-hop labels that satisfies the GST-customized 2-hop cover constraint, if each mock edge weight M is larger than the total weight of all non-mock edges, then, after a batch of edge insertions and edge weight decreases, the maintained set of labels by L4G-M-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 $d'(s, t) < 2M$. The proof of Theorem 1 shows that, since each mock edge weight M is larger than the total weight of all non-mock edges, s and t are properly connected after the change.

Let $E_{in} = \{(a_1, b_1), \dots, (a_i, b_i)\}$ 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_x is closer to s than b_x along $p'(s, t)$ for each $x \in [1, i]$, and u' is between (a_k, b_k) and (a_{k+1}, b_{k+1}) , as illustrated in Figure S2. 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. L4G-M-Insert calls *DIFUSE* 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 L4G-M-Insert still satisfies the GST-customized 2-hop cover constraint. This theorem holds. \square

Theorem 5. *Consider an enhanced graph with mock vertices, and a corresponding set of 2-hop labels that satisfies the GST-customized 2-hop cover constraint, if each mock edge weight M is larger than the total weight of all non-mock edges, then, after a batch of edge deletions and edge weight increases, the maintained set of labels by L4G-M-Delete satisfies the above constraint for the updated graph.*

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., $d'(z, x) \geq d(z, x)$, then L4G-M-Delete must call *SPREAD*₁ to set $L(x)[z]$ to ∞ .

Subsequently, consider a pair of vertices s and t such that $|\{s, t\} \cap D| < 2$ and $d'(s, t) < 2M$. The proof of Theorem 1 shows that, since each mock edge weight M is larger than the total weight of all non-mock edges, 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, L4G-M-Delete must call *SPREAD*₁ to set $L(x)[u']$ to ∞ , 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, L4G-M-Delete must call $SPREAD_1$ to set either $L(u')[z]$ or $L(x)[z]$ to ∞ . 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 s , along $p'(s, t)$, i.e., to generate a new label $(u', d'(u', s)) \in L(s)$.

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 L4G-M-Delete still satisfies the GST-customized 2-hop cover constraint. This theorem holds. \square

S3. COMPLEXITIES OF ALGORITHMS

A. Complexities of L4G-G

The time complexity of the proposed L4G-G is

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

in a single thread environment, where δ 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 [3]) 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. On the other hand, the space complexity of L4G-G is $O(|V| \cdot \delta)$ without PPR , and $O(|E| \cdot \delta)$ with PPR .

B. Complexities of L4G-M-Insert

The proposed L4G-M-Insert has a time complexity of

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

in a single thread environment, where Y 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 $O(Y \cdot (\log Y + d_a \cdot \delta))$ time. The reason is that, since $O(|CL|) = O(Y)$, the initialization steps in Lines 21-23 takes $O(Y \cdot \log Y)$ time. Moreover, $DIFFUSE$ 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. On the other hand, due to the cost of PPR , the space complexity of L4G-M-Insert is $O(|E| \cdot \delta)$.

C. Complexities of L4G-M-Delete

The proposed L4G-M-Delete has a time complexity of

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

in a single thread environment, where κ 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 $O(Y \cdot d_a)$ time, since there are $O(Y)$ labels deactivated, and each deactivation is followed by $O(d_a)$ neighbor searches. Subsequently, it performs $SPREAD_2$ 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 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 $DIFFUSE$ in L4G-M-Insert, it performs $SPREAD_3$ in $O(Y \cdot (\log Y + d_a \cdot \delta))$ time, given that $O(|AL_3|) = O(Y)$. In the end, due to the cost of PPR , the space complexity of L4G-M-Delete is $O(|E| \cdot \delta)$.

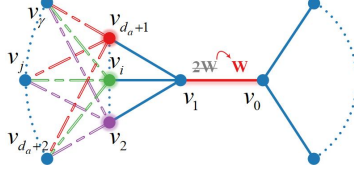


Fig. S3. An example to show the label update process in DecreaseAsync.

S4. THE LABEL UPDATE PROCESS IN STATE-OF-THE-ART METHODS

As discussed in the main contents, the label update process in state-of-the-art methods [1, 2] has a larger time cost than the new label update process in the proposed L4G-M. Specifically, the state-of-the-art methods may unnecessarily update a label multiple times, while L4G-M updates a specific label at most once even in batch cases. We discuss this in detail.

We take DecreaseAsync in [1] as an example to show the label update process in state-of-the-art methods. We refer to DecreaseAsync as InsertAsync in the main experiments. It maintains labels after an edge weight decrease or an edge insertion, and can be parallelly implemented in batch cases [2]. Another state-of-the-art method, IncreaseAsync in [1] that deals with an edge weight increase or an edge deletion, has a similar label update process with DecreaseAsync.

We first briefly introduce the state-of-the-art DecreaseAsync, and then analyze the label update process in DecreaseAsync in detail, as follows.

A. A brief introduction on the state-of-the-art DecreaseAsync

We show the pseudo code of DecreaseAsync 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 DecreaseAsync 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.

DecreaseAsync 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. DecreaseAsync 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 h_c 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 necessary for combining DecreaseAsync and IncreaseAsync 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).

Subsequently, while $CL^c \neq \emptyset$, DecreaseAsync 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, DecreaseAsync returns the updated L and PPR (Line 19).

B. Detailed analyses on the label update process in DecreaseAsync

After a change, DecreaseAsync 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

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

Algorithm S1. The DecreaseAsyn algorithm

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

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1:  $CL^c = CL^n = \emptyset$ 
2: for each label  $(v, d_{va}) \in L(a)$  do
3:   if  $r(v) \geq r(b)$  then
4:     if  $Query(v, b, L) > d_{va} + w_1(a, b)$  then
5:        $L(b)[v] = d_{va} + w_1(a, b)$ ,  $CL^c.push((b, v, d_{va} + w_1(a, b)))$ 
6:     else
7:       if  $v \in C(b)$  &  $L(b)[v] > d_{va} + w_1(a, b)$  then
8:          $L(b)[v] = d_{va} + w_1(a, b)$ ,  $CL^c.push((b, v, d_{va} + w_1(a, b)))$ 
9:        $PPR[b, h_c].push(v)$ ,  $PPR[v, h_c].push(b)$ 
10: for each label  $(v, d_{vb}) \in L(b)$  do
11:   if  $r(v) \geq r(a)$  then
12:     if  $Query(v, a, L) > d_{vb} + w_1(a, b)$  then
13:        $L(a)[v] = d_{vb} + w_1(a, b)$ ,  $CL^c.push((a, v, d_{vb} + w_1(a, b)))$ 
14:     else
15:       if  $v \in C(a)$  &  $L(a)[v] > d_{vb} + w_1(a, b)$  then
16:          $L(a)[v] = d_{vb} + w_1(a, b)$ ,  $CL^c.push((a, v, d_{vb} + w_1(a, b)))$ 
17:        $PPR[a, h_c].push(v)$ ,  $PPR[v, h_c].push(a)$ 
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
23:       if  $Query(v, u_n, L) > d_{new} = d_u + w_1(u, u_n)$  then
24:          $L(u_n)[v] = d_{new}$ ,  $CL^n.push((u_n, v, d_{new}))$ 
25:       else
26:         if  $v \in C(u_n)$  &  $L(u_n)[v] > d_{new}$  then
27:            $L(u_n)[v] = d_{new}$ ,  $CL^n.push((u_n, v, d_{new}))$ 
28:          $PPR[u_n, h_c].push(v)$ ,  $PPR[v, h_c].push(u_n)$ 

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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 $O(Y \cdot d_a)$ 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 [4, 5], the above label update process takes $O(Y \cdot d_a^2 \cdot \delta)$ time. We show an example as follows.

Consider the graph in Figure S3, where $W \gg Y \gg d_a^2$; $w_0(v_0, v_1) = 2W$; $w_0(v_1, v_i) = 1$ 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 $j \in [d_a + 2, Y]$, 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 $j \in [d_a + 2, Y]$, it sequentially updates $L(v_j)[v_0] = W + 1 + d_a - i + 2$ through the path between v_i and v_j for each $i \in [2, d_a + 1]$. That is to say, it could unnecessarily update a specific label-contained distance value multiple times. In particular, it updates label-contained distance values $O(Y \cdot d_a)$ times. Since each label update in an iterative call of *ProDecrease* induces $O(d_a)$ distance queries in the next iterative call, DecreaseAsyn conducts $O(Y \cdot d_a^2)$ distance queries. Since each distance query takes $O(\delta)$ time, the label update process in DecreaseAsyn costs $O(Y \cdot d_a^2 \cdot \delta)$. In comparison, as discussed before, the new label update process in the proposed L4G-M has a smaller time cost of

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

given that we generally have $d_a^2 \cdot \delta \gg \log Y + d_a \cdot \delta$.

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