

# Model Solutions

## Assignment 2

November 9, 2014

### 1 Splitting a red-black tree in $O(\log n)$ time

As the question asks us to use *specialUnion*( $T_1, T_2$ ), we take a closer look at that algorithm.

#### Understanding specialUnion

$T = \text{specialUnion}(T_1, T_2)$  merges two RBT,  $T_1$  and  $T_2$  (given  $T_1 < T_2$ ) in  $O(\max(h_1, h_2))$  time, where  $h_1$  and  $h_2$  are the black heights of  $T_1$  and  $T_2$ , respectively. The bottleneck step here is finding minimum element  $x$  in  $T_2$ . Note that, if  $x$  is already known special union runs in  $O(|h_1 - h_2|)$  time.

For this problem we modify *specialUnion*( $T_1, T_2$ ) to

$T = \text{specialUnionMod}(T_1, x, T_2)$ , where  $x$  is such that  $T_1 < x < T_2$  and  $T = T_1 \cup \{x\} \cup T_2$ . In this modification,  $x$  is made the root of  $T$  and  $T_1$  and  $T_2$  are made  $x$ 's left and right sub-tree respectively. This modification of *specialUnion* runs in  $O(|h_1 - h_2|)$  time.

Also, note that the height of  $T$  returned by *specialUnionMod* is at most  $\max(h_1, h_2) + 1 = O(\max(h_1, h_2))$ .

#### 1.1 Algorithm *split*( $T, x$ )

##### $O(\log^2 n)$ algorithm

Using a naive approach we can achieve  $O(\log^2 n)$  time. Initialize two empty trees  $T_1$  and  $T_2$ .  $T_1$  will store all the elements less than  $x$  and  $T_2$  will store all the elements greater than  $x$ . Simply traverse from root to  $x$  and at every node, union the sub-tree not traversed with  $T_1$  or  $T_2$  appropriately using *specialUnion*( $T_1, T_2$ ). There can be at most  $O(\log n)$  union operations and each union operation will take at most  $O(\log n)$  time.

##### $O(\log n)$ algorithm

We can improve the above naive strategy by using the modified *specialUnion* wisely. We note that using *specialUnionMod* while traversing from root to

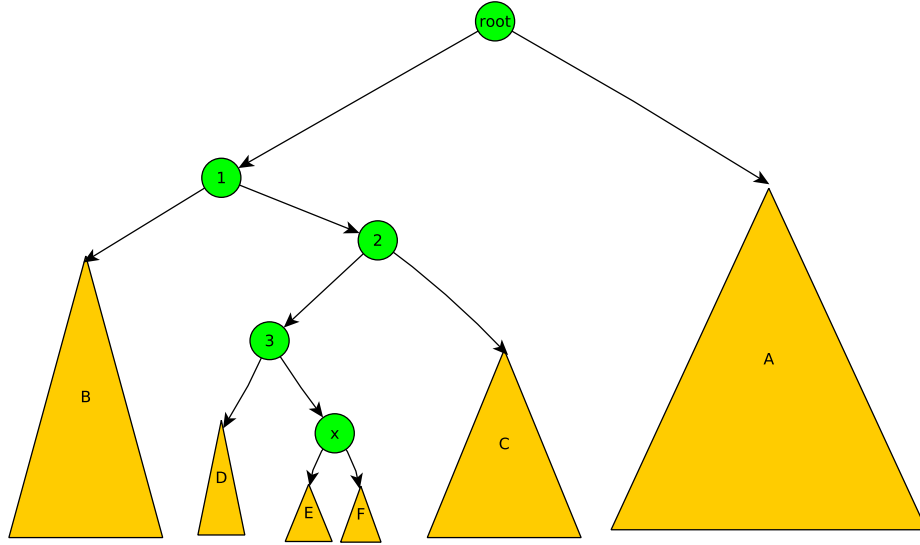


Figure 1: Iterationwise progression of split trees  $T_1$  and  $T_2$ . Initialization:  $T_1 = E$ ,  $T_2 = F$ , Iteration 1:  $T_1 = E \cup D$ ,  $T_2 = F$ , Iteration 2:  $T_1 = E \cup D$ ,  $T_2 = F \cup C$ , Iteration 3:  $T_1 = E \cup D \cup B$ ,  $T_2 = F \cup C$ , Iteration 4:  $T_1 = E \cup D \cup B$ ,  $T_2 = F \cup C \cup A$

$x$  will not give us much advantage because the height of  $T_1$  and  $T_2$  will be  $O(\log n)$  on the first union itself and all the later union operations will take  $O(\log n)$  time. The 'aha!' step is to note that a bottom-up traversal will build  $T_1$  and  $T_2$  incrementally, keeping the costs of union step in check. For example see figure 1

## Pseudo Code

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### Algorithm 1 *split*( $T, x$ )

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//Search  $x$  in  $T$ 
1:  $currNode \leftarrow search(T, x)$ 
   //Initialize  $T_1$  and  $T_2$ 
2:  $T_1 \leftarrow \{\}, T_2 \leftarrow \{\}$ 
3: if ( $currNode \neq null$  and  $left(currNode) \neq null$ ) then
4:    $T_1 = left(currNode)$ 
5: end if
6: if ( $currNode \neq null$  and  $right(currNode) \neq null$ ) then
7:    $T_2 = right(currNode)$ 
8: end if
   //Traverse from  $x$  to root and perform specialUnion on the way
9: while  $parent(currNode) \neq currNode$  do
10:   $currNode \leftarrow parent(currNode)$ 
11:  if ( $value(currNode) < x$ ) then
12:     $T_1 \leftarrow specialUnionMod(T_1, currNode, left(currNode))$ 
13:  else
14:     $T_2 \leftarrow specialUnionMod(T_2, currNode, right(currNode))$ 
15:  end if
16: end while

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## Analysis

1. **Sketch of correctness:** Initially  $T_1$  contains left sub-tree of  $x$  and  $T_2$  contain right subtree of  $x$  (line 4-7). We maintain following invariant: At the end of each iteration, we have correctly added the elements in sub-tree rooted at  $currNode$  in the correct tree,  $T_1$  or  $T_2$ .

As we move from  $x$  to root, at any node ( $currNode$  in pseudo code), all the elements in the sub-tree of  $currNode$  containing  $x$  have already been added to either  $T_1$  or  $T_2$ , in the previous iterations. In this iteration, if  $x$  is contained in the left sub-tree of  $currNode$  we add  $currNode$  and right sub-tree to  $T_2$  as they are greater than  $x$  (line14). Otherwise, if  $x$  is contained in the left sub-tree of  $currNode$  we add  $currNode$  and left sub-tree to  $T_1$  as they are smaller than  $x$  (line 12). We terminate when we reach the root node.

2. **Time complexity** Suppose for creating  $T_1$ , a sequence of unions of trees with following heights is performed  $h_1, h_2, \dots, h_k$ . As we move from  $x$  to root we know  $h_1 \leq h_2 \leq \dots \leq h_k$ .

When we perform a union operation between trees of height  $a$  and  $b$ , the height of resultant tree is  $O(\max(a, b))$  Therefore, height of  $T_1$

follows the following sequence  $h_1, h_2, \dots, h_k$ . Using this, we can easily claim that the  $i^{th}$  union operation takes  $O(h_{i+1} - h_i)$  time. Hence time complexity for creating  $T_1$  is

$$\sum_{i=1}^{O(\log n)} O(h_{i+1} - h_i) = O(\log n)$$

3. **Space Complexity** We just manipulate the pointers in the input tree. Therefore, only  $O(1)$  extra space is required.

## 2 Thinking beyond limits and beyond marks

### 2.1 Finding $min\_L(s)$ for vertex $s$

We know a DFS from  $s$  will visit all the vertices reachable from  $s$ . We can then simply compare the label  $L(v)$ ,  $\forall v$  visited during call to  $DFS(s)$ .

#### Pseudo Code

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##### Algorithm 2 $DFS(G, s)$

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1: for  $v \in G.V$  do
2:    $visited[v] \leftarrow false$ 
3: end for
4:  $min\_L(s) \leftarrow \infty$ 
5:  $DFS\_visit(G, s, s)$ 

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##### Algorithm 3 $DFS\_visit(G, v, source)$

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1: if  $L[v] < min\_L[source]$  then
2:    $min\_L[source] \leftarrow L[v]$ 
3: end if
4:  $visited[v] \leftarrow true$ 
5: for  $u \in G.V$  s.t.  $(v, u) \in G.E$  do
6:   if  $visited[u] = false$  then
7:      $DFS\_visit(G, u, source)$ 
8:   end if
9: end for

```

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#### Analysis

1. **Correctness:** Whenever a node is visited its label is compared with minimum label seen so far and value of  $min\_L$  is updated. From correctness of DFS it follows that all nodes connected to  $s$  are visited.

2. **Time complexity:** Same as DFS  $O(m + n)$
3. **Space complexity:** Same as DFS  $O(n)$

## 2.2 Computing $\text{min\_}L(v)$ for all the vertex

The key idea is to observe that if a node  $v$  is reachable from a set of nodes  $W$  then the set  $W$  is reachable from  $v$  in the reverse graph  $G'$  (graph with same vertices as  $G$ , but with reversed edges). We note that  $\text{min\_}L(v)$  of a node is the minimum label node reachable from  $v$  in  $G$ . Equivalent definition in the reverse graph  $G'$  for  $\text{min\_}L(v)$  is the minimum label among the nodes that can reach  $v$ . Using this property if we start DFS from nodes in increasing order of label value we will be able to mark all the min labels. We follow the following procedure:

1. Create reverse graph  $G'$
2. Sort nodes of  $G'$  in ascending order
3. Do *DFS* on  $G'$  but select source of DFS in the sorted order
4. Every node is visited exactly once and whenever a node  $v$  is visited mark  $\text{min\_}L(v) \leftarrow L[\text{source}]$

## Pseudo Code

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### Algorithm 4 *assgnMinLabel*( $G$ )

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//Reverse G
1:  $G'.V \leftarrow G.V$ 
2: for  $(u, v) \in G.E$  do
3:    $G'.E \leftarrow G'.E \cup \{(v, u)\}$ 
4: end for
5:  $\text{Sort}(G'.V)$ 
6: for  $v \in G'.V$  do
7:    $\text{visited}[v] \leftarrow \text{false}$ 
8: end for
   //call DFS_visit( $G', \text{vertex}, \text{source}$ )
9: for  $v \in G'.V$  do
10:  if  $\text{visited}[v] = \text{false}$  then
11:     $\text{DFS\_visit}(G', v, v)$ 
12:  end if
13: end for

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**Algorithm 5** *DFS\_visit*( $G', v, source$ )

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1: visited[ $v$ ]  $\leftarrow true$ 
2: min_L[ $v$ ]  $\leftarrow L[source]$ 
3: for  $u \in G'.V$  s.t.  $(v, u) \in G'.E$  do
4:   if visited[ $u$ ] = false then
5:     DFS_visit( $G', u, source$ )
6:   end if
7: end for
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**Analysis**

1. **Correctness:** In DFS every node is visited exactly once and once a node  $v$  is visited we update its  $min\_L(v)$  as the label of the source  $s$  of the DFS call. As the source of the DFS call is selected in sorted order it implies that  $v$  is not reachable (in  $G'$ ) from any node with smaller label value than  $s$ . Also, if  $v$  is reachable from  $s$  in  $G'$  then  $s$  is reachable from  $v$  in  $G$ . Therefore,  $min\_L(v)$  is the minimum possible label  $v$  can reach in the original graph  $G$ .
2. **Time Complexity:** Creating reverse graph takes  $O(m + n)$  if adjacency list is used, sorting takes  $O(n \log n)$  and DFS takes  $O(m + n)$ , therefore time complexity is  $O(m + n \log n)$ .
3. **Space Complexity:**  $O(m + n)$  for storing reverse graph and  $O(n)$  for DFS.

### 3 Breadth of special graph

We begin by pondering over the question of the longest path in a rooted tree. The first question that naturally arises is about the structure of such a path. Should the longest path always pass through the root node? It is instructive to look at figure 2 to get a better understanding. Here path P-N-L-J-H-F-I-K-M-O-Q is the longest path and it doesn't pass through the root  $A$ . For this example tree,  $breadth[A]$  i.e. the length of the corresponding path in the sub-tree rooted at  $A$  is 10 and the longest path doesn't pass through  $A$ . From this we get the idea that the longest path for sub-tree rooted at  $A$  could either pass through  $A$  or it is present in one of the sub-trees rooted at its children  $B$ ,  $C$  or  $D$ .

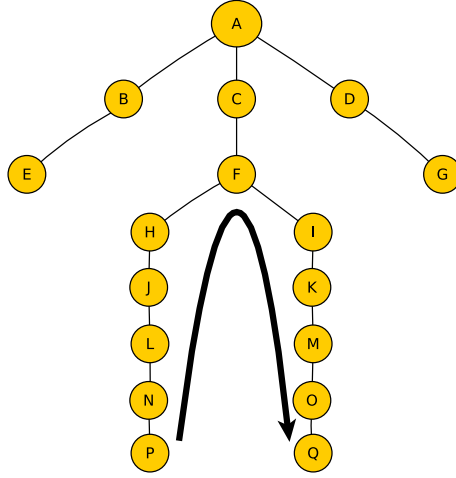


Figure 2: Example illustrating that the longest path doesn't necessarily pass through the root node

In general we should try to answer the question: for a node  $v$ , what is the relation between  $breadth(v)$  and  $breadth(u)$ ,  $\forall u \in v.children$ ? Observe that it is the maximum of the longest paths in sub-tree rooted at each of its children and the longest path passing through  $v$  itself. Notice that the longest path through  $v$  is related to maximum length path from leaf nodes to  $v$ 's children. Longest path through  $v$  is the sum of the top two maximum length paths ( $max_1, max_2$ ) from root to two of  $v$ 's children and 2. This gives rise to following recurrence:

$$breadth(v) = \max(breadth(u), max_1 + max_2 + 2), \forall u \in v.children$$

This recurrence becomes the crux of the recursive algorithm below. The last question that remains is that above discussion is for a rooted tree but input is given in form of a graph. This is answered by the fact that the given graph is a tree and a DFS from any vertex will give us a tree rooted at that vertex.

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**Algorithm 6**  $DFS(G, s)$

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- 1: **for**  $v \in G.V$  **do**
  - 2:    $visited[v] \leftarrow false$
  - 3: **end for**
  - 4: **for** some  $s \in G.V$ , **do**  $DFS\_visit(G, s)$
  - 5: **Return**  $\max breadth[v], \forall v$  as breadth of  $G$
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**Algorithm 7** *DFS\_visit*( $G, v$ )

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INPUT: Graph  $G$  and node to be visited  $v$   
OUTPUT: 1) *breadth* of tree rooted at  $v$   
          2) max path from some leaf to  $v$

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1:  $max_1 \leftarrow 0, max_2 \leftarrow 0, breadth_v \leftarrow 0$ 
2:  $visited[v] \leftarrow true$ 
3: for  $u \in G.V$  s.t.  $(v, u) \in G.E$  do
4:   if  $visited[u] = false$  then
5:      $(breadth_u, maxlen_u) = DFS\_visit(G, u)$ 
6:      $breadth_v = \max(breadth_u, breadth_v)$ 
7:     if  $max_1 < maxlen_u + 1$  then
8:        $max_2 \leftarrow max_1$ 
9:        $max_1 \leftarrow maxlen_u + 1$ 
10:    else if  $max_2 < maxlen_u + 1$  then
11:       $max_2 \leftarrow maxlen_u + 1$ 
12:    end if
13:  end if
14: end for
15:  $breadth_v \leftarrow \max(breadth_v, max_1 + max_2)$ 
16: return  $(breadth_v, max_1)$ 
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**Analysis**

1. **Sketch of correctness:** Argue that the above recurrence is correct. Then, starting from the base case, inductively show that *DFS\_visit*( $G, v$ ) computes correct *breadth<sub>v</sub>* for a sub-tree rooted at  $v$  and max path from leaf to  $v$ .
2. **Time Complexity:** Same as DFS,  $O(m + n) = O(n)$  for tree graph
3. **Space Complexity:** Same as DFS,  $O(n)$