

## Implementation and Analysis of Multi-Splay Trees

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Instructor:
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Teaching Assistant: Sravanthi Chede Summary: This project aims at the implementation of Multi-Splay Trees in C++ and analysing the run-time on some test cases. We also analyse the amortized cost of Multi-Splay Trees theoretically and study the different properties such as Sequential Access property, Static Finger property, and various other Lemmas associated with it. We have used the concepts of Binary Search Trees and Splay Trees. Multi-Splay Trees are conjectured to be Dynamically Optimal.

## 1. Introduction

A Splay Tree is a Binary Search Tree which performs an extra operation called splaying. Splaying is an algorithm which along moves a node to the root upon access, such that the BST property is still maintained. Splay Tree have various useful properties which help us reduce the amortized cost of accessing elements in the BST. Some of these properties are Sequential Access Property, Dynamic Finger Property, Working Set Property, Dynamic Optimality, etc. A Multi-Splay Tree is a tree made up of Splay trees, wherein the nodes have several different types of properties. The root of each individual splay trees is darkened.

Multi-Splay tree has O(log log n) competitive bound and O(log n) amortized complexity for access in a BST, which is proved by an access Lemma. In the Multi-splay tree data structure, we assume a reference tree, denoted by P, which is a perfectly balanced tree consisting n nodes. Each node has a preferred child. A path consisting of preferred child of various node is called preferred path. The preference tree is only for understanding purpose and is not part of our data structure.

Our Multi-Splay tree denoted by T, has either solid or dashed edges. Set of nodes connected by solid edges is our Splay Tree. The set of nodes in a Splay Tree is exactly the same as the nodes in its corresponding preferred path. That is, at any point in time the Multi-Splay Trees can be obtained from the reference tree by viewing each preferred edge as solid, and doing rotations on the solid edges.

Functions used for various operations in our implementation:

explore() - auxiliary function to help display the whole tree.

treeFor() - function to create the whole tree.

rotate() - function to perform the standard rotation operation.

splay() - function to perform the standard splaying operation.

switchPath() - function to adjust the depths and minDepths values of the nodes

refParent() - function to return the first child whose minDepth value is greater than the depth value

expose() - function to bring the current node to the root of the whole tree.

query() - function to access an element in the tree (multi-splay).

# 2. Figures, Tables and Algorithms

An example of a Multi-Splay Tree along with the reference tree for explanation and better understanding.

## 2.1. Figures

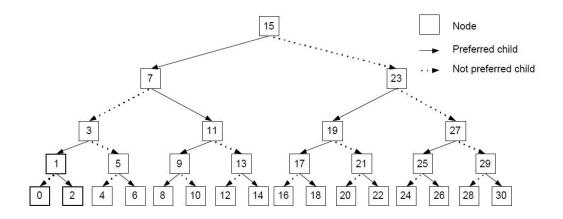


Figure 1: Reference Tree P

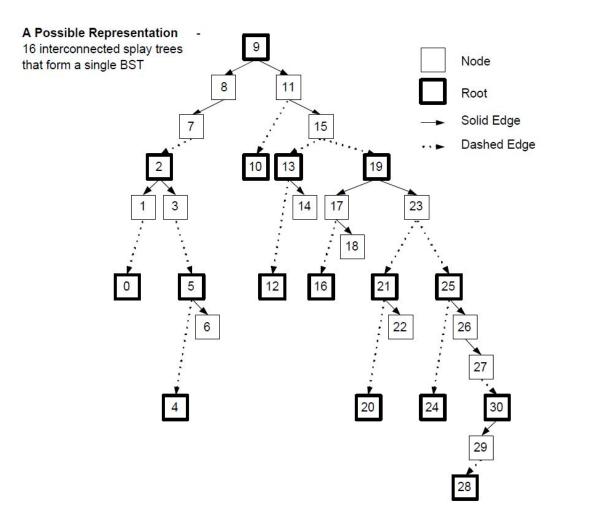


Figure 2: 16 inter-connected Splay Trees to form a single BST

#### 2.2. Tables

This table contains the approximate Build Time and Query Time taken by the Multi-Splay Tree in the different test cases used. For the different test cases, different access sequences were used on the tree having size ranging from 30 to 300000 elements.

	Build Time	Query Time
Sequential		
1. Size: 30	$0 \mathrm{ns}$	$0 \mathrm{ns}$
2. Size: 3000	$900 \mathrm{ns}$	$3500 \mathrm{ns}$
3. Size: 300000	$200000 \mathrm{ns}$	650000 ns
Reverse		
1. Size: 30	$0 \mathrm{ns}$	$0 \mathrm{ns}$
2. Size: 3000	$900 \mathrm{ns}$	$2500 \mathrm{ns}$
3. Size: 300000	$35000 \mathrm{ns}$	600000 ns
Random		
1. Size: 30	$0 \mathrm{ns}$	$0 \mathrm{ns}$
2. Size: 3000	$900 \mathrm{ns}$	1000ns-10000ns
3. Size: 300000	$40000 \mathrm{ns}$	500000ns-1000000ns

Table 1: Runtime Analysis

## 2.3. Algorithms

We call the query() function to access any node in the Multi-Splay Tree. Its functionality is that it splays the element until it reaches to the root of the entire BST. So, when splayed for the first time, the node is moved to the root of the splay tree in which it belongs by performing necessary rotations. Following, using refParent() and switchPath(), the further splaying procedure is taken place by adjusting the depths and minDepths of the node. It ends when the parent of the accessed node is NULL, i.e. the accessed node reaches to the root of the BST. query() uses expose() function, which in turn uses splay(), rotate(), refParent() and switchPath() function.

Pseudo code of the query() algorithm:

### Algorithm 1 bool query(key)

- 1: Search until we find either the key or NULL
- 2: The search operation is similar to Binary Search
- 3: curr = current node
- 4: prev = parent of curr
- 5: **if** curr == NULL **then**
- 6: expose(prev)
- 7: return false
- 8: end if
- 9: expose(curr)
- 10: return true

Pseudo code of the expose() algorithm:

#### Algorithm 2 void expose(node)

- 1: while node->parent != NULL do
- 2: splay(node)
- 3: update depths and minDepths
- 4: end while
- 5: return

## 3. Some further useful suggestions

**Theorem 3.1.** Multi-Splaying is amortized  $O(\log n)$ .

The amortized time taken to access elements in a Multi-Splay Tree is O(log n).

**Theorem 3.2.** Multi-Splaying is competitive  $O(\log \log n)$ .

Multi-Splay Tree is a competitive BST having a time complexity of O(log log n).

**Theorem 3.3.** For any query in a multi-splay tree, the worst-case cost is  $O(\log^2 n)$ .

This follows from the fact that to query a node, we visit at most O(height(P)) splay trees. Because the size of each splay tree is  $O(\log n)$ , the total number of nodes we can possibly touch is  $O(\log^2 n)$ .

### 4. Conclusions

It is the only algorithm with O(log log n) competitiveness in a BST. In addition, the above theorems and results show that Multi-Splay Trees satisfy many of the important properties of a Dynamically Optimal BST algorithm.

## 5. Bibliography and citations

[1] [2] [3] [4] [5]

# Acknowledgements

Research Papers published by researchers at School of Computer Science, Carnegie Mellon University in Pittsburgh.

#### References

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Hyperlink:

Research Papers: [2] [4] [5]