

CSE 326 Data Structures

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Priority Queues : ~~AVL~~ ^{SPLAY} Trees

Logistics

- Hand in Homework #3
- Midterm this Friday, Friday, Friday, in class
 - Review this Wednesday
 - Sample Midterm on website
 - Homework solutions on website

AVL Trees Revisited

- Balance condition:
 - For every node x , $-1 \leq \text{balance}(x) \leq 1$
 - Strong enough : Worst case depth is $O(\log n)$
 - Easy to maintain : *one* single or double rotation
- Guaranteed $O(\log n)$ running time for
 - Find ←
 - Insert ←
 - Delete ←

AVL Trees Revisited

- What **extra info** did we maintain in each node?

Height

- **Where** were rotations performed?

destroyed balance \rightarrow fix

- How did we **locate** this node?

Recursion.

Other Possibilities?

- Could use different balance conditions, different ways to maintain balance, different guarantees on running time, ...

- Why aren't AVL trees perfect?

Extra info
complex logic to
detect & fix balance

- Many other balanced BST data structures

- Red-Black trees

- AA trees

- **Splay Trees** ←

- 2-3 Trees

- **B-Trees** ←

- ...

skip heap
leftist

Splay Trees

- Blind adjusting version of AVL trees
 - Why worry about balances? Just rotate anyway!
- Amortized time per operations is $O(\log n)$
- Worst case time per operation is $O(n)$
 - But guaranteed to happen rarely $O(M \log n)$

Insert/Find always rotate node to the root!

I am dd.

SAT/GRE Analogy question:

AVL is to Splay trees as Leftist is to Skip

Recall: Amortized Complexity

If a sequence of M operations takes $O(M f(n))$ time, we say the amortized runtime is $O(f(n))$.

- Worst case time *per operation* can still be large, say $O(n)$
- Worst case time for any sequence of M operations is $O(M f(n))$

Average time per operation for any sequence is $O(f(n))$

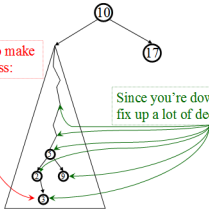
Amortized complexity is *worst-case* guarantee over *sequences* of operations.

Recall: Amortized Complexity

- Is amortized guarantee any weaker than worstcase?
- Is amortized guarantee any stronger than averagecase?
- Is average case guarantee good enough in practice?
- Is amortized guarantee good enough in practice?

The Splay Tree Idea

If you're forced to make a really deep access:



Since you're down there anyway, fix up a lot of deep nodes!

Find/Insert in Splay Trees

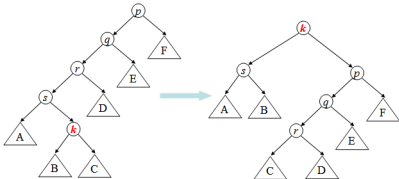
1. Find or insert a node k
2. **Splay k to the root using:**
 - zig-zag, zig-zig, or plain old zig rotation

Why could this be good??

1. Helps the new root, k
 - o *Great if k is accessed again*
2. And helps many others!
 - o *Great if many others on the path are accessed*

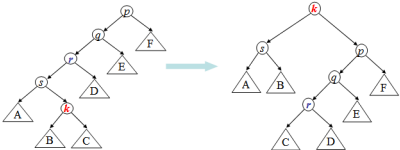
Splaying node k to the root: Need to be careful!

One option (that we won't use) is to repeatedly use AVL single rotation until k becomes the root:
(see [Section 4.5.1](#) for details)

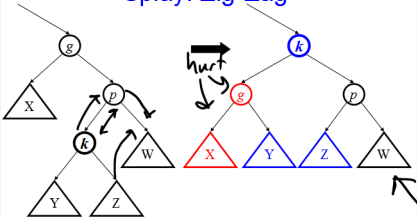


Splaying node k to the root: Need to be careful!

What's bad about this process?



Splay: Zig-Zag*



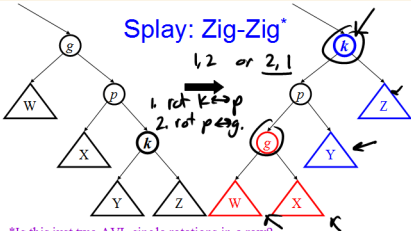
*Just like an...

AVL double rotation

Which nodes improve depth?

Y, Z, K

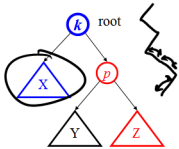
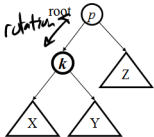
Splay: Zig-Zig*



Why does this help?

Not really for subtrees

Special Case for Root: Zig



Relative depth of p , X , Z ?

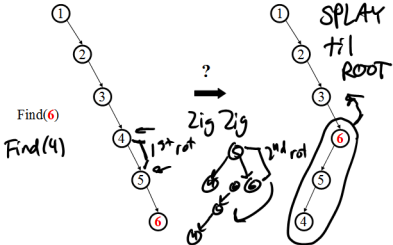
Down

Relative depth of everyone else?

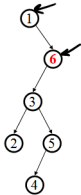
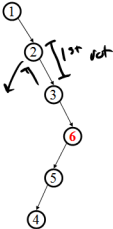
Why not drop zig-zig and just zig all the way?

Helps only one child

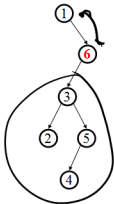
Splaying Example: Find(6)



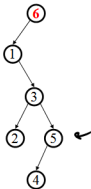
Still Splaying 6



Finally...

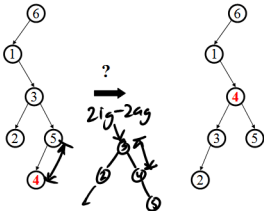


?
→
Zig

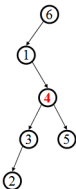


Another Splay: Find(4)

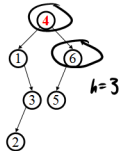
Find(4)



Example Splayed Out



?
→
Zig zag



But Wait... $M=2$

$O(M \log n)$

What happened here?

Didn't two find operations take linear time instead of logarithmic?

What about the amortized $O(\log n)$ guarantee?

Do every thing **SPLAY.**

Time in constructing

Why Splaying Helps

- If a node n on the access path is at depth d before the splay, it's at about depth $d/2$ after the splay

$$d \rightarrow \frac{d}{2} \quad \text{node} \rightarrow \text{depth} = l$$


- Overall, nodes which are low on the access path tend to move closer to the root

- Splaying gets amortized $O(\log n)$ performance. (Maybe not now, but soon, and for the rest of the operations.)

Practical Benefit of Splaying

- No heights to maintain, no imbalance to
check for memory
105 complex
 - Less storage per node, easier to code
- Often data that is accessed once,
is soon accessed again!
 - Splaying does implicit *caching* by bringing it to
the root

Splay Operations: Find

- Find the node in normal BST manner
- Splay the node to the root
 - if node not found, splay what would have been its parent 

What if we didn't splay?

Worst case



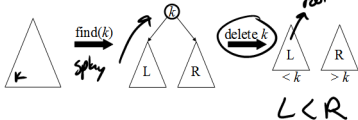
Splay Operations: Insert

- Insert the node in normal BST manner
- Splay the node to the root

What if we didn't splay?

Splay Operations: Remove

delete

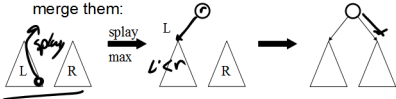


Now what?

Join

Join(L, R):

given two trees such that (stuff in L) < (stuff in R),
merge them:

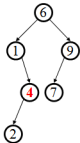


**Splay on the maximum element in L, then
attach R**

Does this work to join any two trees? *need LLR*

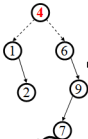
Delete Example

Delete(4)



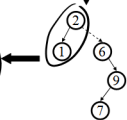
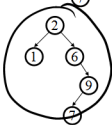
find(4)

zig zag



Find max

zig



Splay Tree Summary

- All operations are in amortized $O(\log n)$ time
- Splaying can be done top-down; this may be better because:
 - only one pass
 - no recursion or parent pointers necessary
 - *we didn't cover top-down in class*
- Splay trees are *very* effective search trees
 - Relatively simple ←
 - No extra fields required ←
 - Excellent locality properties: frequently accessed keys are cheap to find