CSE 326: Data Structures Splay Trees

Hal Perkins Winter 2008 Lecture 13

AVL Trees Revisited

- What extra info did we maintain in each node?
- Where were rotations performed?
- How did we locate this node?

AVL Trees Revisited

• Balance condition:

For every node x, $-1 \le \text{balance}(x) \le 1$

- 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?
 - buildTree ?

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Other Possibilities?

- Could use different balance conditions, different ways to maintain balance, different guarantees on running time, ...
- Why aren't AVL trees perfect?
- Many other balanced BST data structures
 - Red-Black trees
 - AA trees
 - Splay Trees
 - 2-3 Trees
 - B-Trees
 - ...

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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

Insert/Find always rotate node to the root!

SAT/GRE Analogy question:

AVL is to Splay trees as ______ is to ______

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Recall: Amortized Complexity

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

Recall: Amortized Complexity

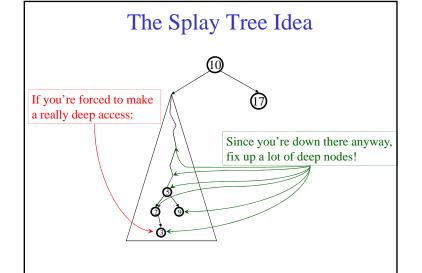
If a sequence of M operations takes O(M f(n)) time, we say the amortized runtime is O(f(n)).

- 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.

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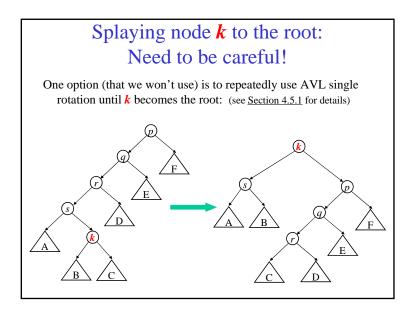
Find/Insert in Splay Trees

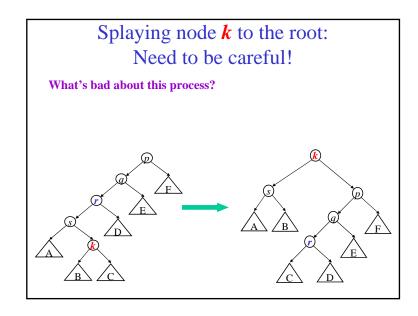
- 1. Find or insert a node *k*
- **2. Splay** *k* **to the root using:** zig-zag, zig-zig, or plain old zig rotation

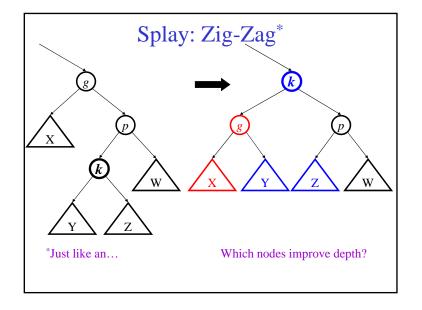
Why could this be good??

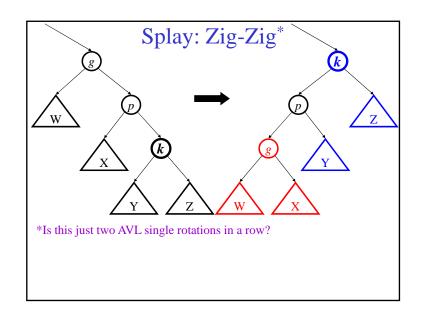
- 1. Helps the new root, k
 - o Great if k is accessed again
- 2. And helps many others!
 - o Great if many others on the path are accessed

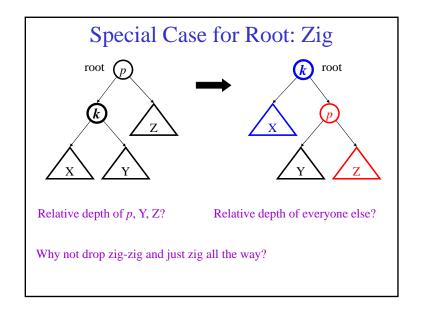
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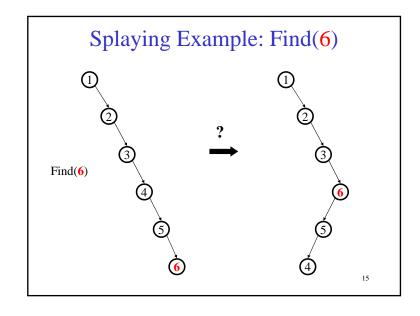


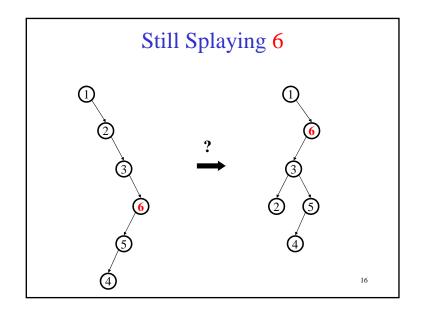


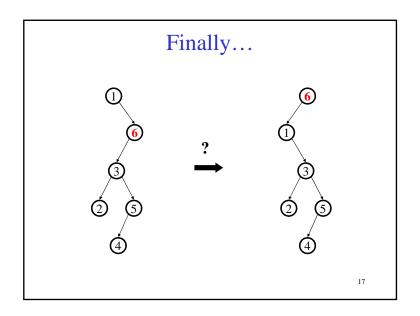


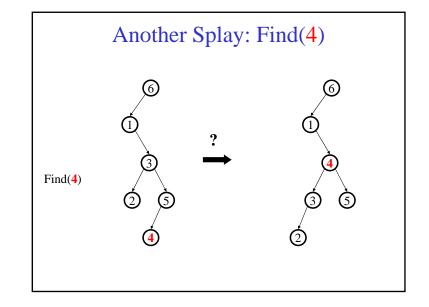


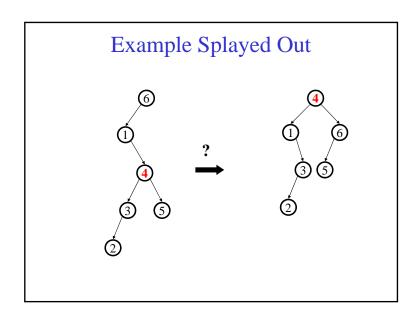












But Wait...

What happened here?

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

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

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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
- 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.)

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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?

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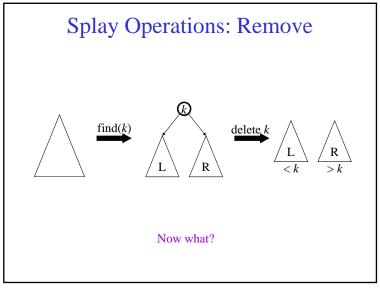
Practical Benefit of Splaying

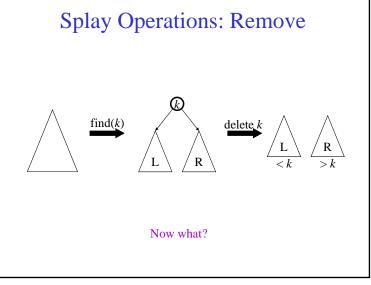
- No heights to maintain, no imbalance to check for
 - Less storage per node, easier to code
- Often data that is accessed once, is soon accessed again!
 - Splaying does implicit caching by bringing it to the root

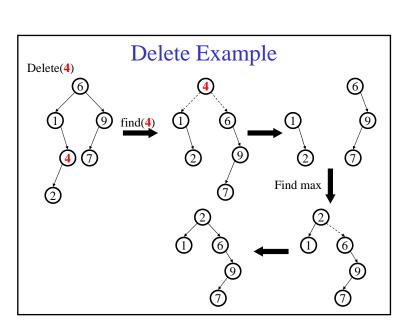
Splay Operations: Insert

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

What if we didn't splay?



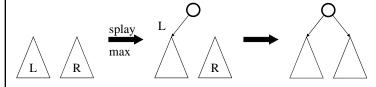




Join

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?

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 28

