# CSE 326: Data Structures Splay Trees

James Fogarty Autumn 2007 Lecture 10

### AVL Trees Revisited

• Balance condition:

Left and right subtrees of *every node* have *heights* **differing by at most 1** 

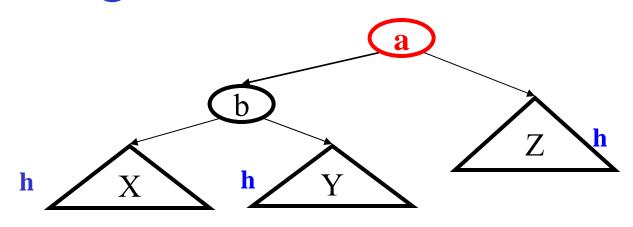
- Strong enough : Worst case depth is  $O(\log n)$ 

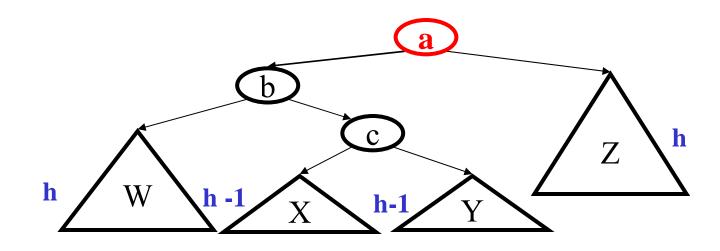
- Easy to maintain : one single or double rotation

- Guaranteed O(log *n*) running time for
  - Find ?
  - Insert ?
  - Delete ?
  - buildTree ?

 $\Theta(n \log n)$ 

#### Single and Double Rotations





#### AVL Trees Revisited

• What extra info did we maintain in each node?

• Where were rotations performed?

• How did we locate this node?

## 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 imbalance, recursive bottom-up implementation

- 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!
- <u>Amortized time</u> 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 Analo AVL is to Spla		is to	
Leftish heap : Skew heap		6	

### 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 <u>any</u> 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? Yes, it is only for sequences
- Is amortized guarantee any stronger than averagecase?

Yes, guarantees *no* bad sequences

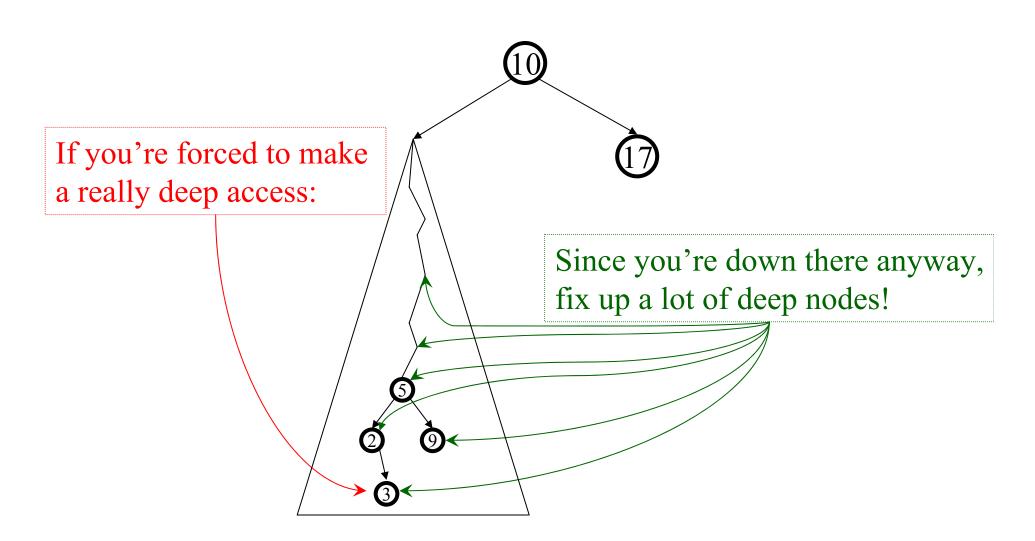
• Is average case guarantee good enough in practice?

No, adversarial input,

• Is amortized guarantee good enough in practice?

Yes, again, no bad sequences

## The Splay Tree Idea



#### Find/Insert in Splay Trees

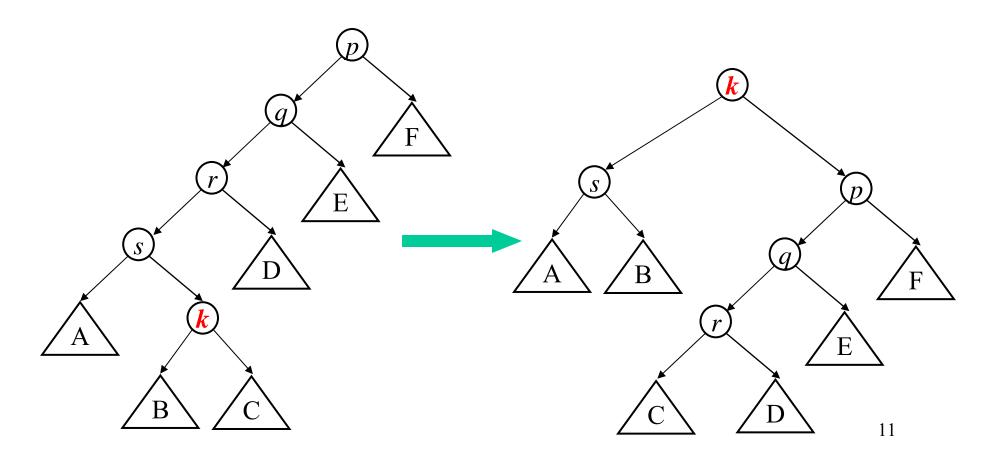
- 1. Find or insert a node k
- 2. Splay *k* to the root using: zig-zag, zig-zig, or plain old zig rotation

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!

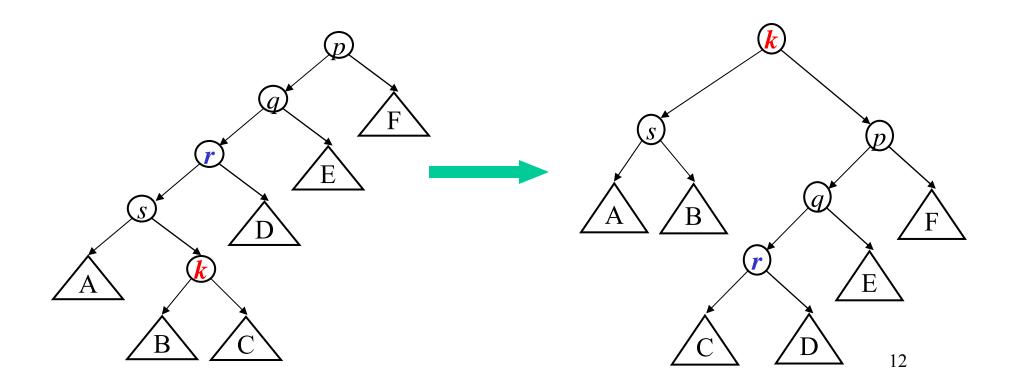
One option (that we won't use) is to repeatedly use AVL single rotation until k becomes the root: (see Section 4.5.1 for details)

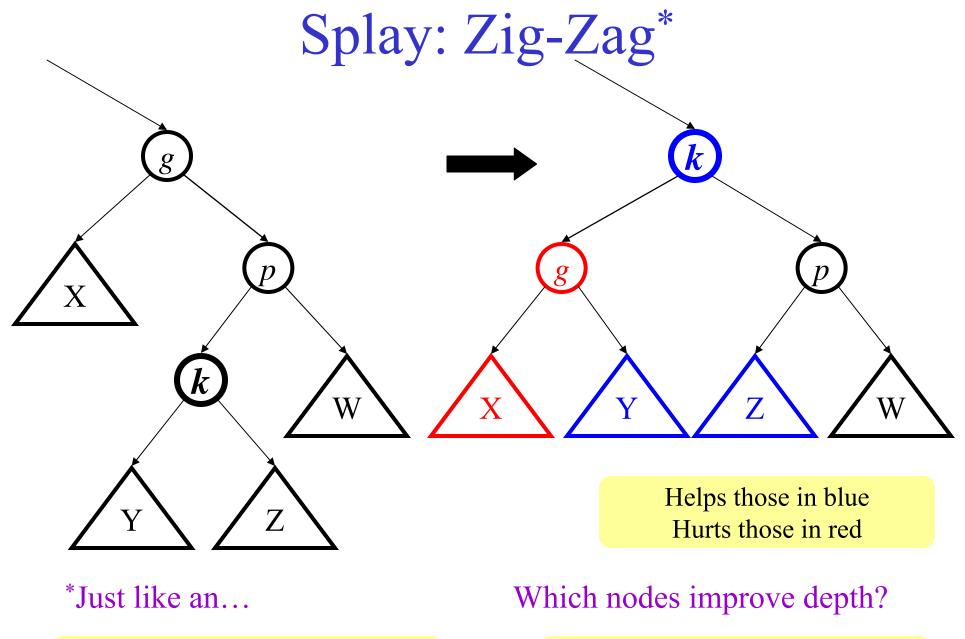


#### Splaying node *k* to the root: Need to be careful!

What's bad about this process?

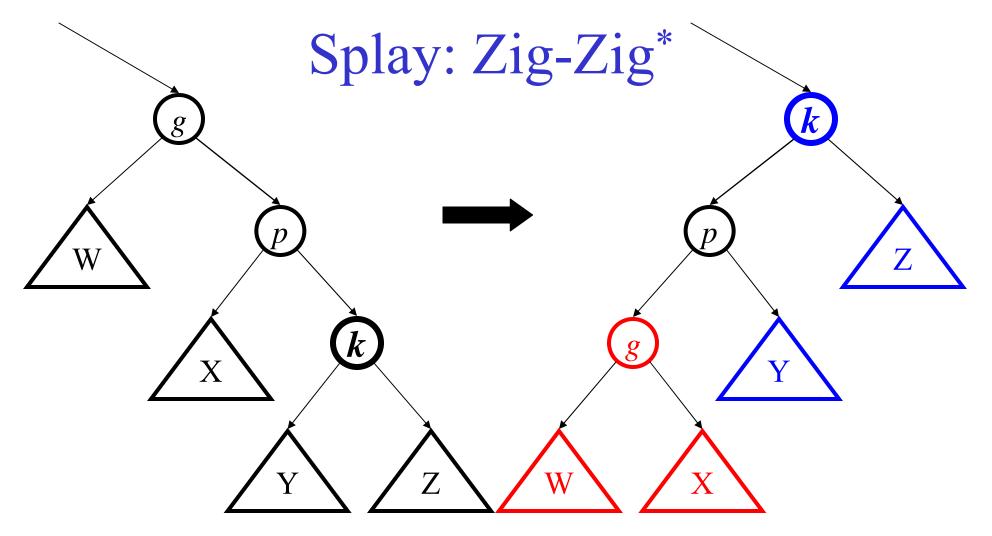
*r* is pushed almost as low as *k* was Bad seq: find(k), find(r), find(...), ...





AVL double rotation

k and its original children



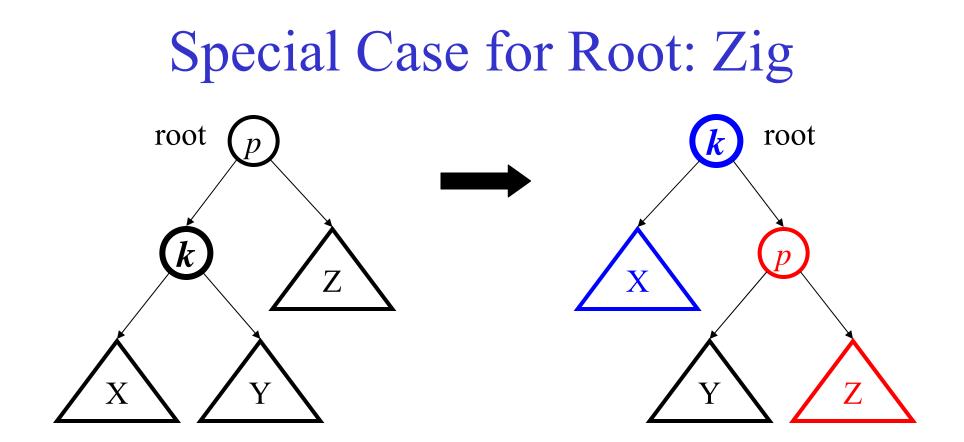
\*Is this just two AVL single rotations in a row?

Not quite – we rotate g and p, then p and k

Why does this help?

Same number of nodes helped as hurt. But later rotations help the whole subtree.

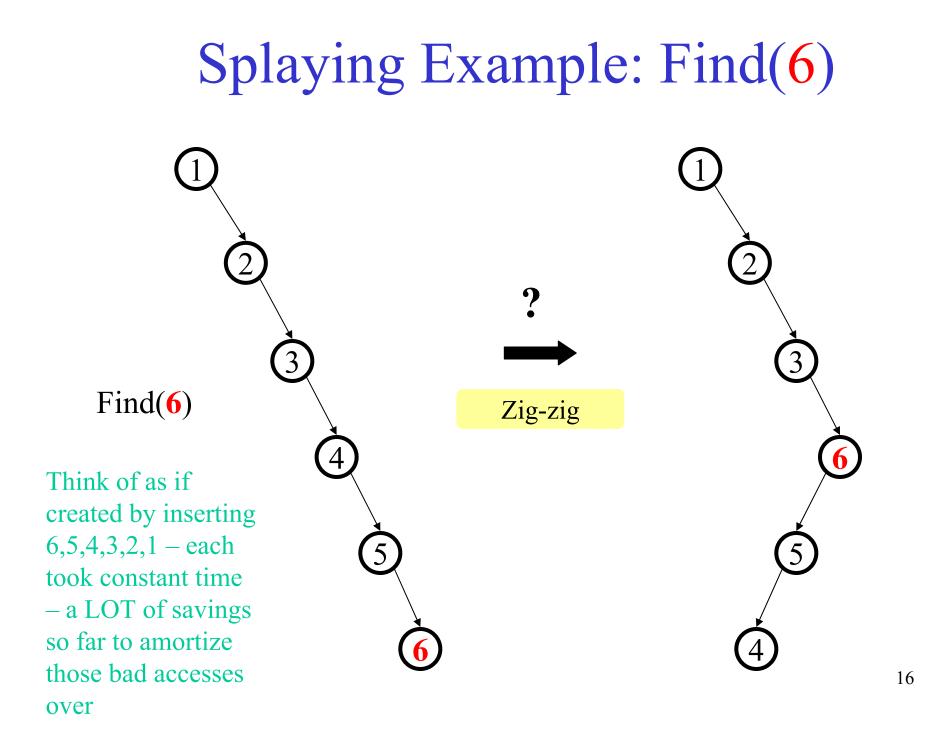
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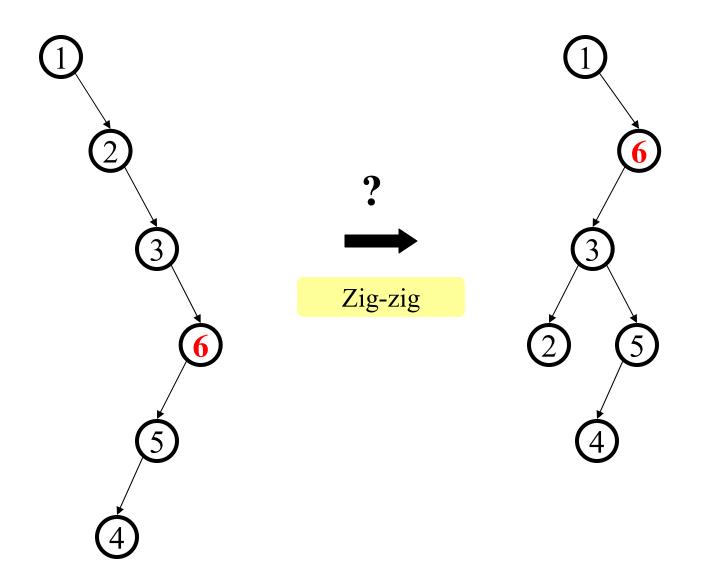
Relative depth of *p*, Y, Z?

Relative depth of everyone else?

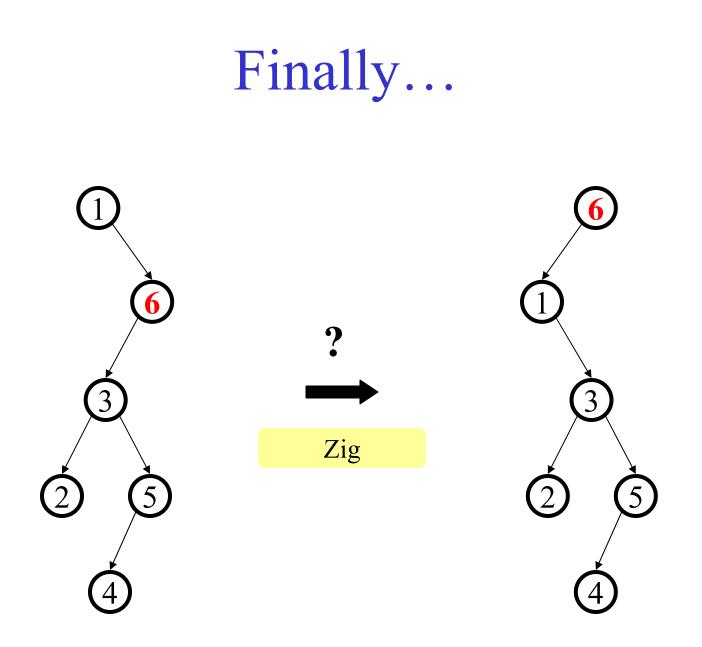
Down 1 level Much better Why not drop zig-zig and just zig all the way? Zig only helps one child!



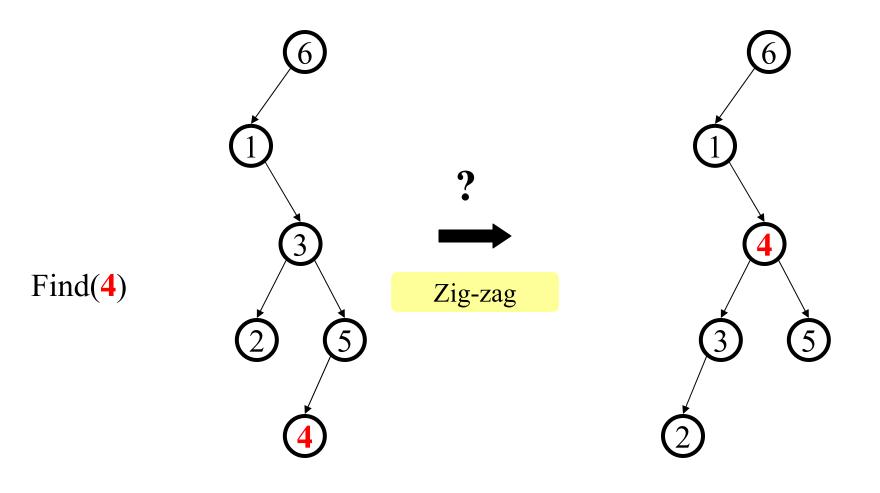
## Still Splaying 6



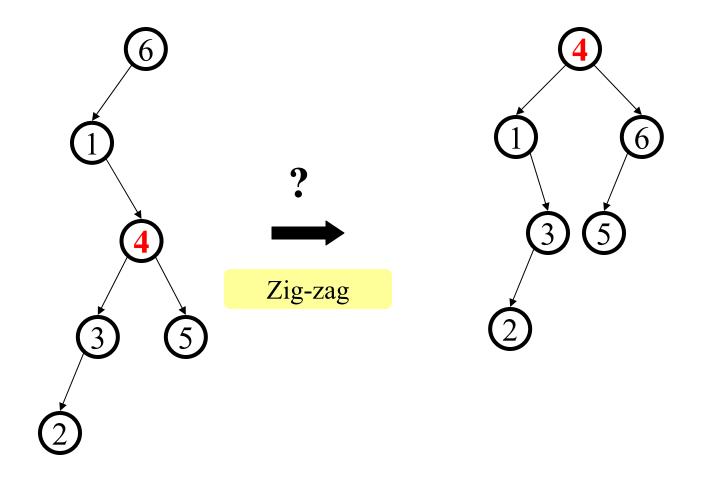
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#### Another Splay: Find(4)



#### Example Splayed Out



#### But Wait...

What happened here?

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

What about the amortized O(log *n*) guarantee?

That still holds, though we must take into account the previous steps used to create this tree. In fact, a splay tree, by construction, will *never* look like the example we started with!

## 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.)

## Practical Benefit of Splaying

- No heights to maintain, no imbalance to check for
  Less storage per node, easier to code
- Data accessed once, is often 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 <u>not</u> found, splay what would have been its parent

What if we didn't splay?

Amortized guarantee fails! Bad sequence: find(leaf *k*), find(*k*), find(*k*), ...

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## Splay Operations: Insert

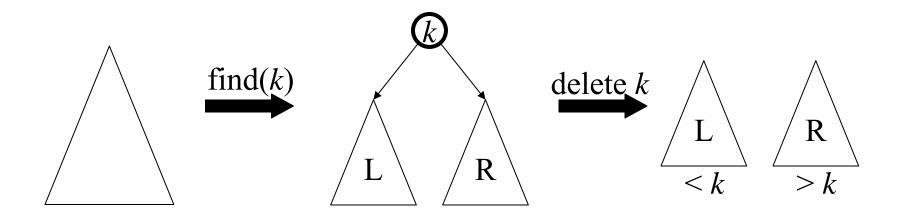
- Insert the node in normal BST manner
- Splay the node to the root

What if we didn't splay?

Amortized guarantee fails! Bad sequence: insert(*k*), find(*k*), find(*k*), ...

### Splay Operations: Remove

Everything else splayed, so we'd better do that for remove

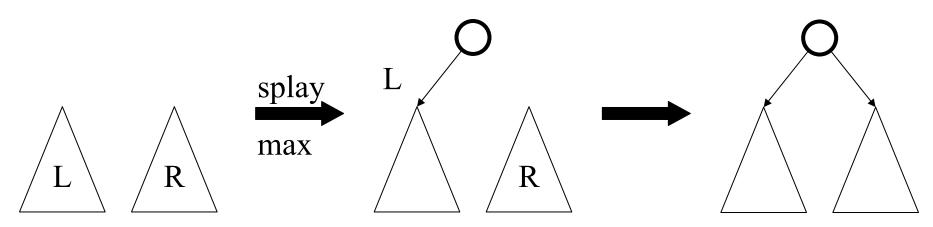


#### Now what?

### 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

Similar to BST delete – find max = find element with no right child

Does this work to join *any* two trees? No, need L < R

#### Delete Example Delete(4) (9) find(4) Find max

## 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
    Like what? Skew heaps! (don't need to wait)
  - 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

What happens to node that never get accessed? (tend to drop to the bottom)

 Excellent *locality* properties: frequently accessed keys are cheap to find

