CSE 373: Data Structures & Algorithms Wrap up Amortized Analysis; AVL Trees

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Course Logistics

Symposium offered by CSE department today

 HW2 released, Big-O, Heaps (lecture slides have pseudocode that will help a lot)

Weekly summaries out soon

Review: Amortized Complexity

- For a Stack implemented with an array can we claim **push** is O(1) time if resizing is O(n) time?
 - We can't, but we can claim it's an O(1) amortized operation
- Why is this good? Why do we care?
 - If amortized is good enough for our problem,
 then it's great to be able to say O(1) instead of O(n)

Review: Amortized Complexity

Like an Average.

```
N inserts at 1 unit cost + 1 insert at N unit cost
```

```
= N * 1 + 1 * N
```

2N overall cost for N + 1 insertions

```
= 2N \cos t / (N + 1) insertions
```

$$= O(2N/(N+1))$$

= O(1) amortized cost

Not-Amortized Complexity

What if we only add 3 slots instead of N slots?

3 inserts at 1 unit cost + 1 insert at N unit cost

$$= 3 * 1 + 1 * N$$

= N + 3 overall cost for N + 1 insertions

$$= N + 3 \cos t / (3 + 1) insertions$$

$$= O((N+3)/(3+1))$$

$$= O(N)$$

Not Amortized! Didn't build up enough "credit"

Example #1: Resizing stack

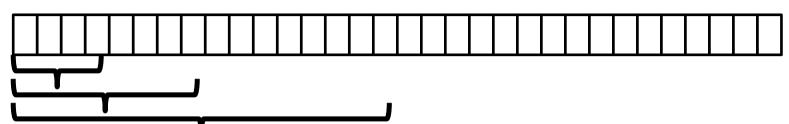
A stack implemented with an array where we double the size of the array if it becomes full

Claim: Any sequence of push/pop/isEmpty is amortized O(1)

Need to show any sequence of M operations takes time O(M)

- Recall the non-resizing work is $O(\mathbf{M})$ (i.e., $\mathbf{M}^*O(\mathbf{1})$)
- The resizing work is proportional to the total number of element copies we do for the resizing
- So it suffices to show that:
 After M operations, we have done < 2M total element copies
 (So average number of copies per operation is bounded by a constant)

Amount of copying



After **M** operations, we have done < **2M** total element copies

Let **n** be the size of the array after **M** operations

- Then we have done a total of:

Because we must have done at least enough push operations to cause resizing up to size n:

$$M \ge n/2$$

- So

 $2M \ge n > number of element copies$

Have to be careful

- If array grows by a constant amount (say 1000), operations are not amortized O(1)
 - After $O(\mathbf{M})$ operations, you may have done $\Theta(\mathbf{M}^2)$ copies
- If array doubles when full and shrinks when 1/2 empty, operations are not amortized O(1)
 - Terrible case: pop once and shrink, push once and grow, pop once and shrink, ...
- If array doubles when full and shrinks when 3/4 empty,
 it is amortized O(1)
 - Proof is more complicated, but basic idea remains: by the time an expensive operation occurs, many cheap ones occurred

Amortized Complexity Summary

- Like an average
- You're building up "credit" with N cheap tasks proportional to 1 expensive N task
- Sometimes hard to prove, but useful if your application doesn't require every single operation to be cheap.

There's another example on slides from Wednesday involving Queues, if you're curious

Review: Balanced BST

Observation

- BST: the shallower the better!
- For a BST with n nodes inserted in arbitrary order
 - Average height is $O(\log n)$ see text for proof
 - Worst case height is O(n)
- Simple cases, such as inserting in key order, lead to the worst-case scenario

Solution: Require a Balance Condition that

- 1. Ensures depth is always $O(\log n)$ strong enough!
- 2. Is efficient to maintain not too strong!

The AVL Balance Condition

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

```
Definition: balance(node) = height(node.left) -
height(node.right)
```

AVL property: for every node x, $-1 \le balance(x) \le 1$

- Ensures small depth
 - Will prove this by showing that an AVL tree of height h must have a number of nodes exponential in h
- Efficient to maintain using single and double rotations

The AVL Tree Data Structure

Structural properties

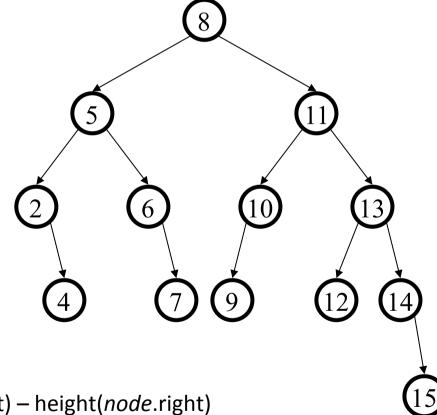
- 1. Binary tree property
- 2. Balance property: balance of every node is between -1 and 1

Result:

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

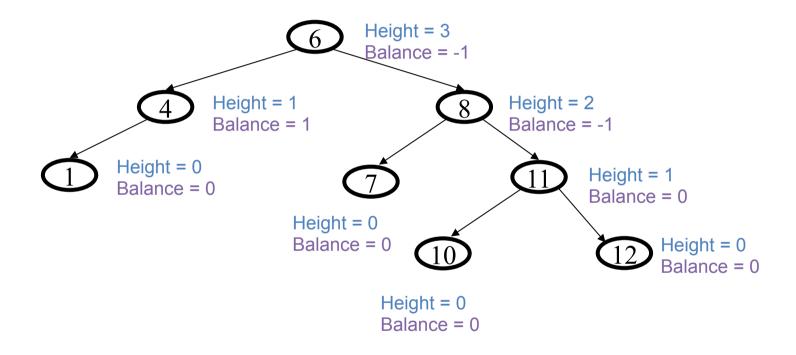
Ordering property

Same as for BST

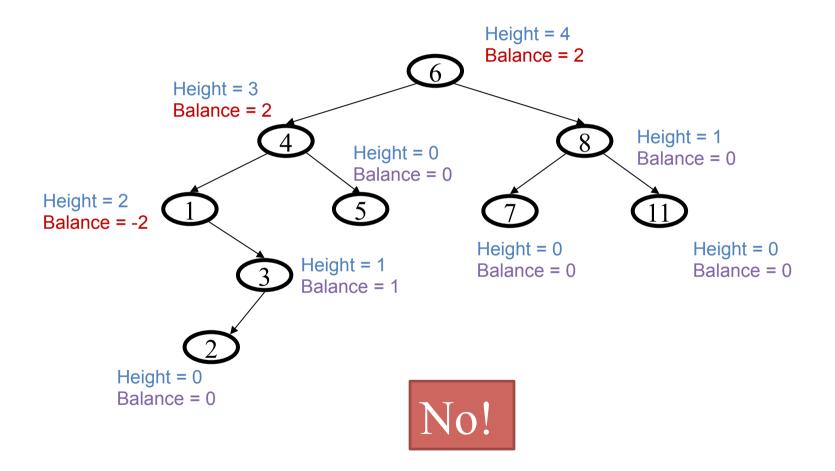


Definition: balance(node) = height(node.left) - height(node.right)

An AVL tree?



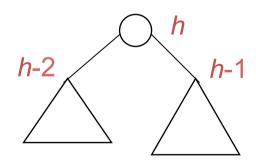
An AVL tree?



Intuition: compactness

- If the heights differ by at most 1, your two subtrees are roughly the same size
- If this is true at every node, it's true all the way down
- If this is true all the way down, your tree winds up compact.
- Height is O(logN)

We'll revisit the formal proof of this soon



AVL Operations

If we have an AVL tree, the height is $O(\log n)$, so find is $O(\log n)$

But as we insert and delete elements, we need to:

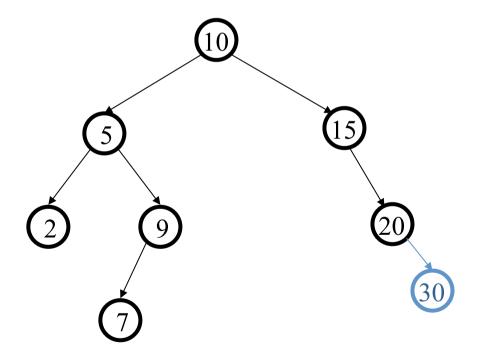
- 1. Track balance
- 2. Detect imbalance
- 3. Restore balance

Is this AVL tree balanced?

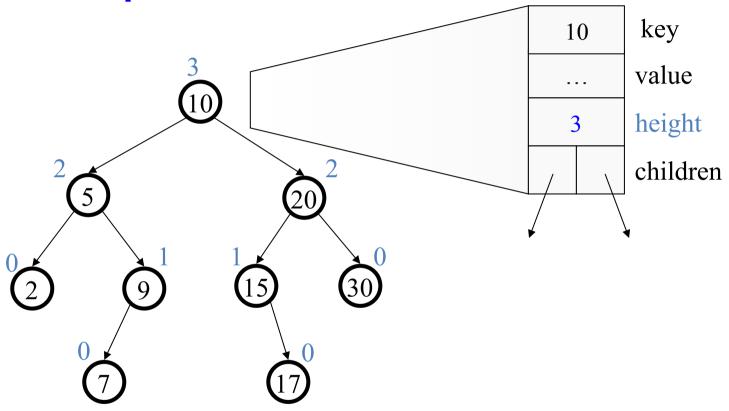
Yep!

How about after insert (30)?

No, now the Balance of 15 is off



Keep the tree balanced



Track height at all times!

AVL tree Operations

• AVL find:

Same as BST find

• AVL insert:

- First BST insert, *then* check balance and potentially "fix" the AVL tree
- Four different imbalance cases

• AVL delete:

- The "easy way" is lazy deletion
- Otherwise, do the deletion and then have several imbalance cases

Insert: detect potential imbalance

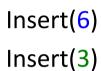
- 1. Insert the new node as in a BST (a new leaf)
- 2. For each node on the path from the root to the new leaf, the insertion may (or may not) have changed the node's height
- 3. So after recursive insertion in a subtree, detect height imbalance and perform a *rotation* to restore balance at that node

Type of rotation will depend on the location of the imbalance (if any)

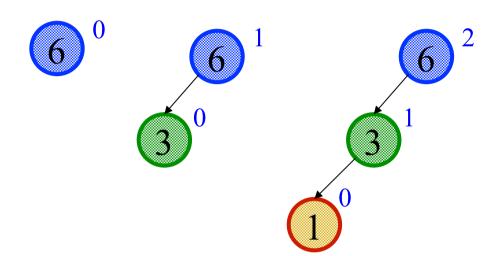
Facts about insert imbalances:

- If there's an imbalance, there must be a deepest element that is imbalanced after the insert
- After rebalancing this deepest node, every node is balanced
- So at most one node needs to be rebalanced

Case #1: Example



Insert(1)



Third insertion violates balance property

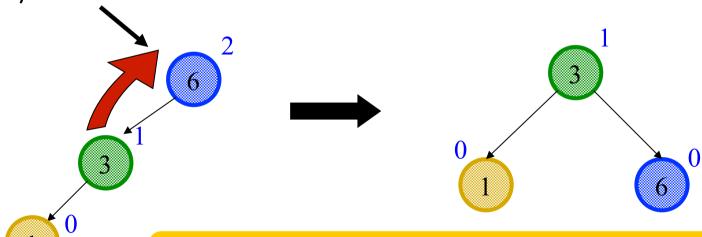
happens to be at the root

What is the only way to fix this (the only valid AVL tree with these nodes?

Fix: Apply "Single Rotation"

- Single rotation: The basic operation we'll use to rebalance
 - Move child of unbalanced node into parent position
 - Parent becomes the "other" child (always okay in a BST!)
 - Other subtrees move in only way BST allows (next slide)



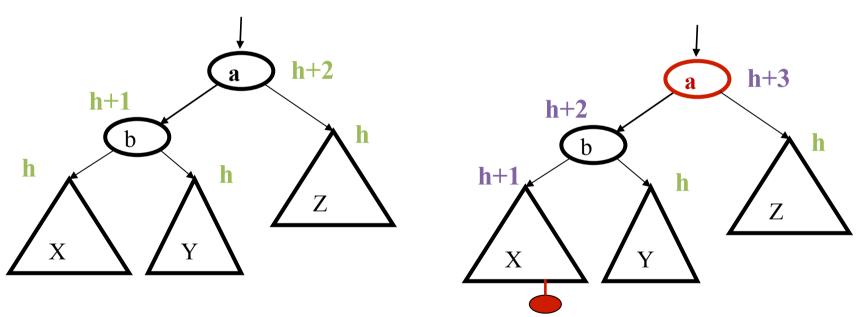


Intuition: 3 must become root

New parent height is now the old parent's height before insert

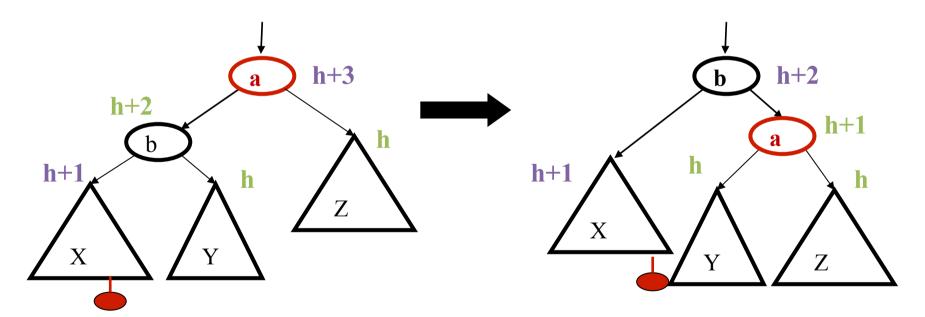
The example generalized

- Node imbalanced due to insertion somewhere in left-left grandchild that causes an increasing height
 - 1 of 4 possible imbalance causes (other three coming)
- First we did the insertion, which would make *a* imbalanced



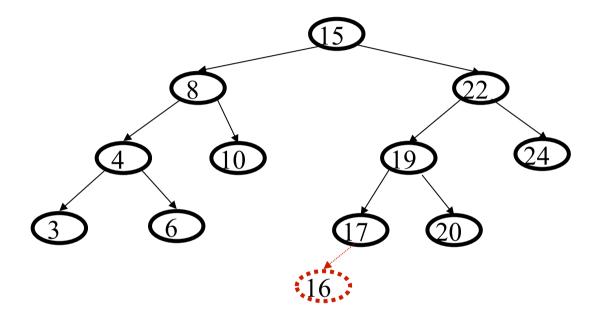
The general left-left case

- Node imbalanced due to insertion somewhere in left-left grandchild
 - 1 of 4 possible imbalance causes (other three coming)
- So we rotate at **a**, using BST facts: X < b < Y < a < Z



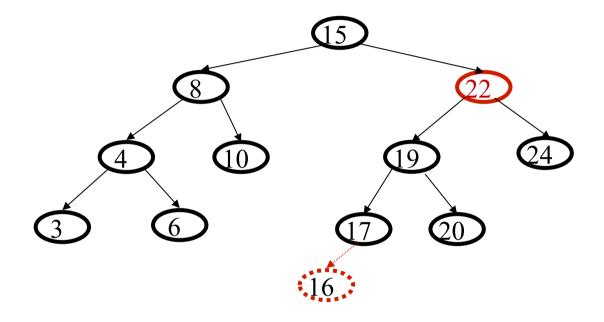
- A single rotation restores balance at the node
 - To same height as before insertion, so ancestors now balanced CSE373: Data Structures &

Another example: insert (16)



Where is the imbalance?

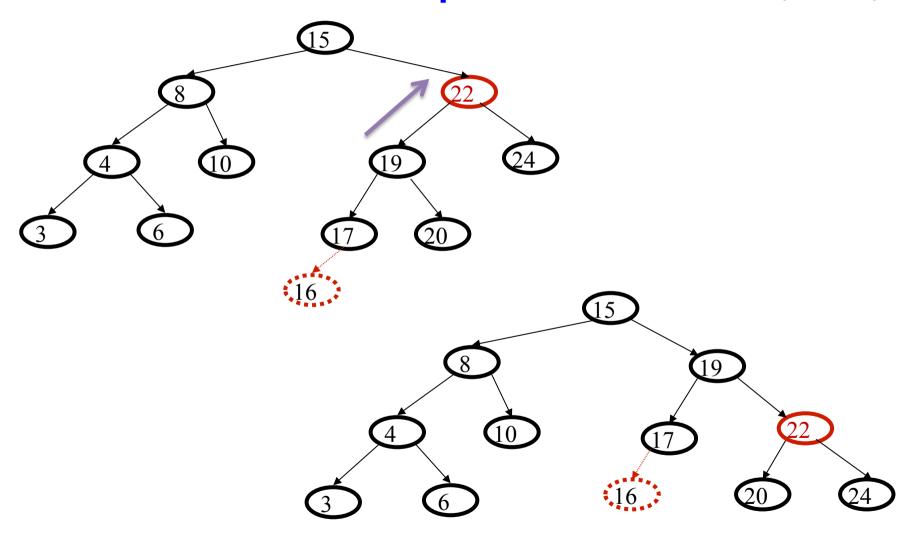
Another example: insert (16)



Where is the imbalance?

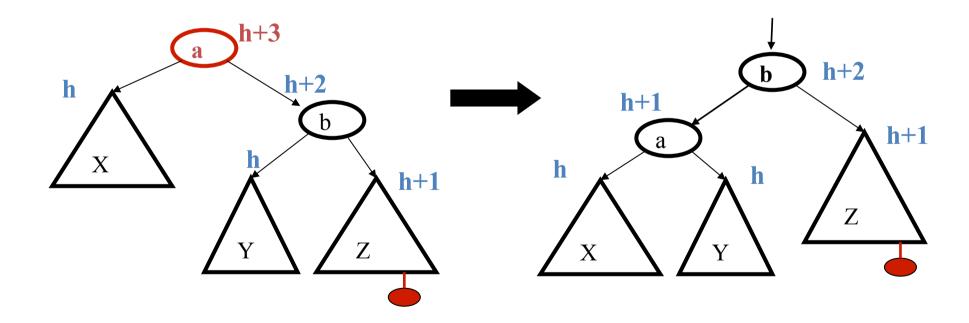
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Another example: insert (16)

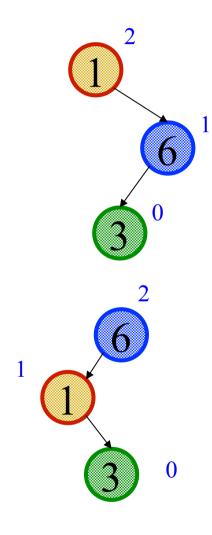


The general right-right case

- Mirror image to left-left case, so you rotate the other way
 - Exact same concept, but need different code



Case 3 & 4: left-right and right-left



Insert(1)

Insert(6)

Insert(3)

Is there a single rotation that can fix either tree?

Insert(6)

Insert(1)

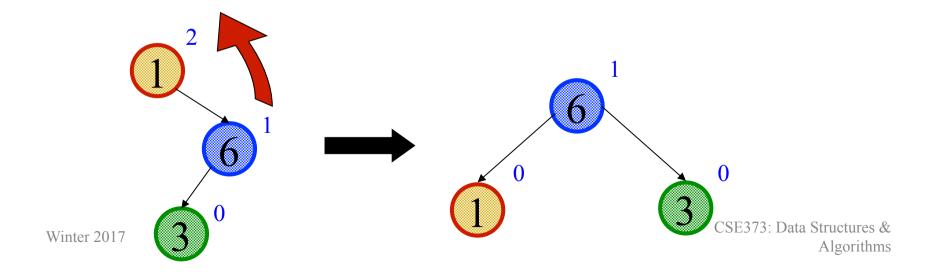
Insert(3)

Wrong rotation #1:

Unfortunately, single rotations are not enough for insertions in the **left-right** subtree or the **right-left** subtree

Simple example: insert(1), insert(6), insert(3)

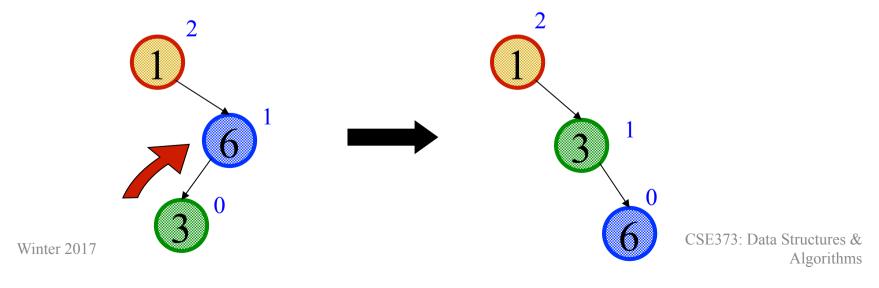
First wrong idea: single rotation like we did for left-left



Wrong rotation #2: Unfortunately, single rotations are not enough for insertions in the left-right subtree or the right-left subtree

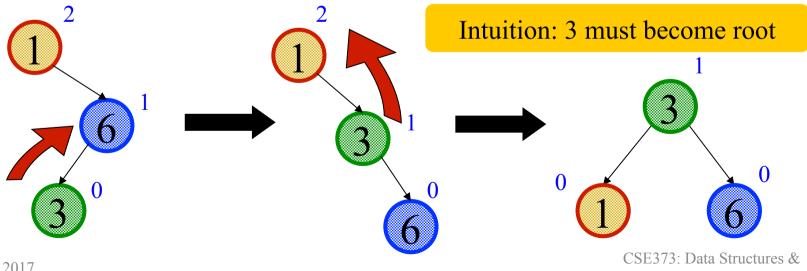
Simple example: insert(1), insert(6), insert(3)

 Second wrong idea: single rotation on the child of the unbalanced node



Sometimes two wrongs make a right

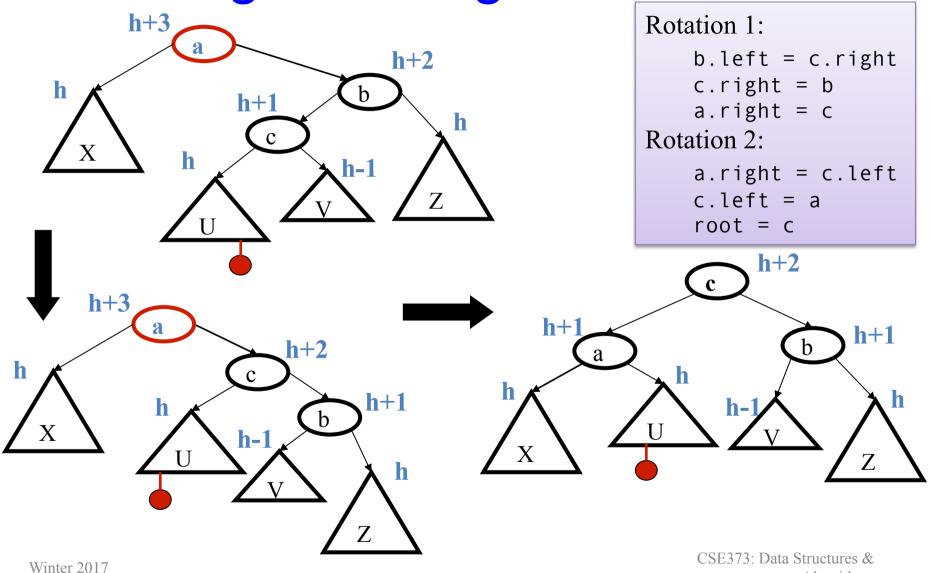
- First idea violated the BST property
- Second idea didn't fix balance
- But if we do both single rotations, starting with the second, it works! (And not just for this example.)
- Double rotation:
 - Rotate problematic child and grandchild
 - Then rotate between self and new child



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Algorithms

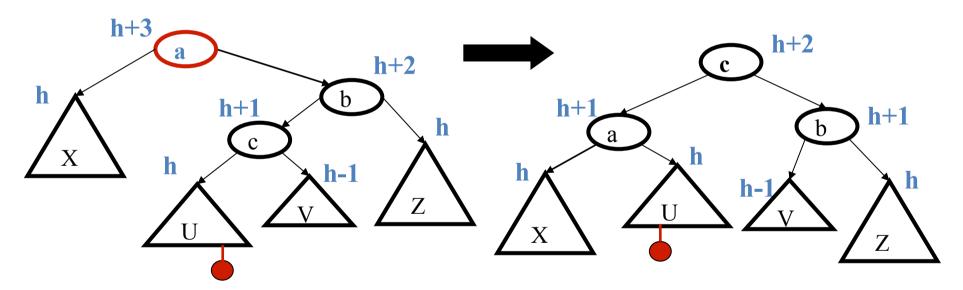
The general right-left case



Algorithms

Comments

- Like in the left-left and right-right cases, the height of the subtree after rebalancing is the same as before the insert
 - So no ancestor in the tree will need rebalancing
- Does not have to be implemented as two rotations; can just do:



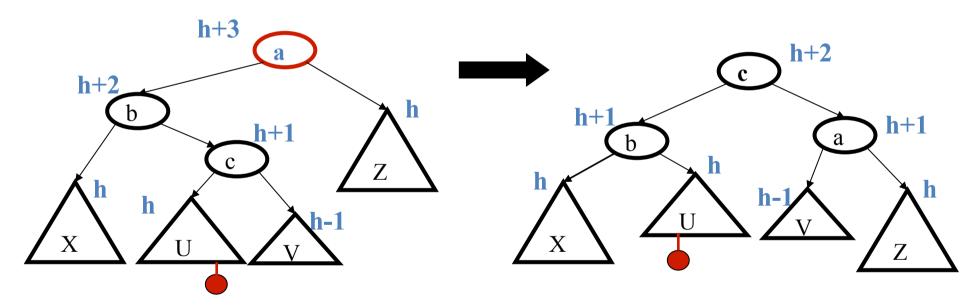
Easier to remember than you may think:

- 1) Move c to grandparent's position
- 2) Put a, b, X, U, V, and Z in the only legal positions for a BST

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The last case: left-right

- Mirror image of right-left
 - Again, no new concepts, only new code to write



Insert, summarized

- Insert as in a BST
- Check back up path for imbalance, which will be 1 of 4 cases:
 - Node's left-left grandchild is too tall (left-left single rotation)
 - Node's left-right grandchild is too tall (left-right double rotation)
 - Node's right-left grandchild is too tall (right-left double rotation)
 - Node's right-right grandchild is too tall (right-right double rotation)
- Only one case occurs because tree was balanced before insert
- After the appropriate single or double rotation, the smallestunbalanced subtree has the same height as before the insertion
 - So all ancestors are now balanced

Efficiency

- Worst-case complexity of find: $O(\log n)$
 - Tree is balanced
- Worst-case complexity of insert: $O(\log n)$
 - Tree starts balanced
 - A rotation is O(1) and there's an $O(\log n)$ path to root
 - (Same complexity even without one-rotation-is-enough fact)
 - Tree ends balanced
- Worst-case complexity of buildTree: $O(n \log n)$

Takes some more rotation action to handle **delete...**

Pros and Cons of AVL Trees

Arguments for AVL trees:

- 1. All operations logarithmic worst-case because trees are *always* balanced
- 2. Height balancing adds no more than a constant factor to the speed of insert and delete

Arguments against AVL trees:

- 1. Difficult to program & debug [but done once in a library!]
- 2. More space for height field
- 3. Asymptotically faster but rebalancing takes a little time
- 4. Most large searches are done in database-like systems on disk and use other structures (e.g., *B*-trees, a data structure in the text)
- 5. If *amortized* (later, I promise) logarithmic time is enough, use splay trees (also in text)

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Today's Takeaways

- Review of Amortized Analysis:
 - feel comfortable proving a runtime's amortized cost
- AVL trees:
 - understand the AVL balance condition
 - be able to identify AVL trees
 - intuition on why the height is O(logN)
 - understand AVL inserts and rotations