#### COM1009 Introduction to Algorithms and Data Structures

Lecture #9



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Reading: Lecture notes on MOLE

#### Aims of this lecture

- To see a class of **self-balancing trees** guaranteeing operations in time  $O(\log n)$ .
- To show that the depth of AVL trees is  $O(\log n)$ .
- To show how to perform insertions and deletions,
   rebalancing the tree through rotations whenever it becomes unbalanced.

# Self-balancing trees

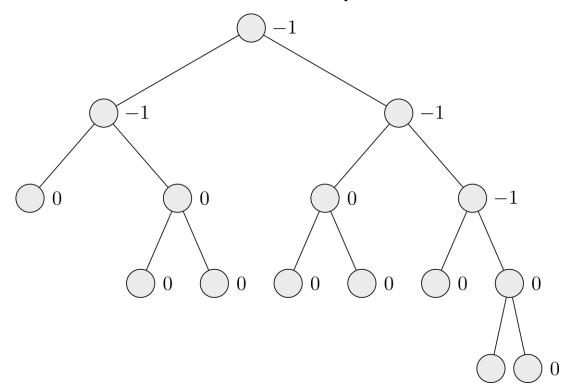
- There are various types of binary search trees that are guaranteed to have depth  $O(\log n)$ .
  - AVL Trees
  - 2-3 Trees
  - B-Trees
  - Red-black Trees
  - Splay Trees
  - Van Emde Boas Trees
  - **—** ...

#### **AVL Trees**

- Invented by and named after Adelson-Velskii and Landis.
- Invariant: all nodes are locally balanced.
- A binary tree is called AVL tree if for every node the following holds: the height of the left subtree and the height of the right subtree only differ by at most 1.
- Let v be a node and  $T_l, T_r$  be its left and right subtrees, respectively. Then  $bal(v) \coloneqq h(T_l) h(T_r)$  is the **balance** factor of v, h() denoting the height of a tree.
- In an AVL tree hence for every node v we have  $bal(v) \in \{-1, 0, +1\}$ .

#### Balance properties

 The local property does not mean that all leaves are on two levels. AVL trees can be lopsided, see this example:



However, overall the tree is still pretty balanced.

#### Estimating the depth of an AVL tree

**Theorem**: the height of an AVL tree with *n* nodes is at most

$$h \le \frac{1}{\log((\sqrt{5}+1)/2)} \log n \approx 1.44 \log n.$$

#### **Proof outline:**

- Consider the minimum number of nodes in any AVL tree of height h and call it A(h).
  - This means that any AVL tree of height h will have  $n \ge A(h)$  nodes.
- Show that A(h) (and thus n) is exponentially large in h.
  - Will show that A(h) is similar to Fibonacci numbers.
- Take logarithms (+maths) to get the claimed bound.

#### Minimum number of nodes in an AVL tree

- Let A(h) be the minimum number of nodes in any AVL tree of height h.
  - An AVL tree with height 0 consists of the root only, hence A(0) = 1.
  - The smallest AVL tree of height 1 has two nodes, hence A(1) = 2.
  - An AVL tree of height h has to have a root with one subtree of height h-1, and the other subtree of height at least h-2. Hence A(h) = 1 + A(h-1) + A(h-2).
- This is similar to the Fibonacci numbers (bar the "1 +"):
  - Fib(0) = Fib(1) = 1 and
  - Fib(h) = Fib(h-1) + Fib(h-2).
  - Handy closed form:  $\operatorname{Fib}(k) \geq \frac{1}{\sqrt{5}} \left[ \left( \frac{\sqrt{5}+1}{2} \right)^{k+1} 1 \right]$

#### Link to Fibonacci numbers

- We prove by induction that A(h) = Fib(h + 2) 1.
- Base case: A(0) = 1 = 2 1 = Fib(2) 1and A(1) = 2 = 3 - 1 = Fib(3) - 1.
- Assume that the claim holds for A(h-1) and A(h-2), then

$$A(h) = 1 + A(h-1) + A(h-2)$$
 (by recurrence)  
 $= 1 + \text{Fib}(h+1) - 1 + \text{Fib}(h) - 1$  (2x induction hypothesis)  
 $= \text{Fib}(h+1) + \text{Fib}(h) - 1$   
 $= \text{Fib}(h+2) - 1$  (by definition of Fib $(h+2)$ ).

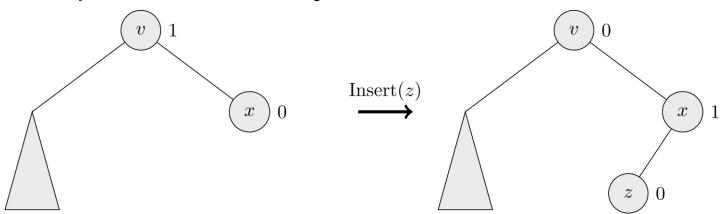
- Every AVL tree with n nodes and height h has  $n \ge A(h) \ge Fib(h+2) 1$ .
- Plugging in closed form for Fib gives  $\left(\frac{\sqrt{5}+1}{2}\right)^{n+3} \le \sqrt{5}n+1$
- Taking logarithm of base  $\frac{\sqrt{5}+1}{2}$ :  $h+3 \leq \log_{(\sqrt{5}+1)/2}(\sqrt{5}n+1)$ Converting to  $\log_2$  completes proof.  $\Rightarrow h \leq \log_{(\sqrt{5}+1)/2}(n)$

#### Search in an AVL Tree

• Works like in an ordinary binary search tree.

#### Inserting in an AVL Tree

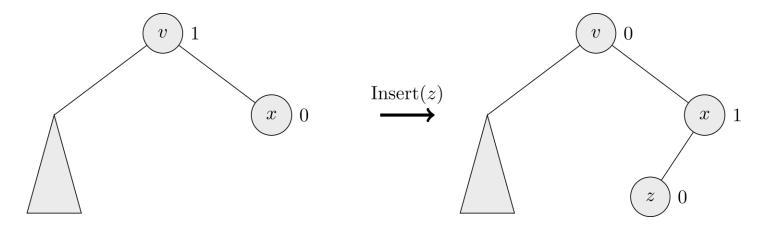
- Works like in an ordinary binary search tree.
- But the tree may become unbalanced, hence we need to rebalance. (We focus on ideas here; writing code is easy.)
- We record the search path to new element, and then work back up the search path to rebalance so long as the height of the current subtree has increased.
- Let v be the current node and its right child x be on the search path (left child is symmetric)



# **►**Insert (1)

**Case 1**: bal(v) = 1.

- Left subtree of v was higher than right subtree before insertion.
- After inserting z, the right subtree has increased its height, hence the subtree at v is now balanced.
- The height of v has not changed, hence rebalancing is done.



# **►Insert (2)**

**Case 2**: bal(v) = 0.

- Both subtrees of v were balanced before insertion.
- After inserting z, the right subtree has increased its height, hence now bal(v) = -1.
- The height of the subtree at v has increased, hence we need to continue rebalancing at v's parent to check for imbalances further up the tree.
- If *v* was the root, we stop.

# **►Insert (3)**

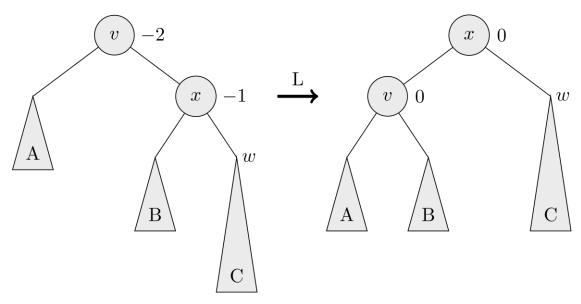
**Case 3**: bal(v) = -1.

- After insertion, the tree has become unbalanced: bal(v) = -2.
- Search path contains nodes *v, x, w* whose subtrees increased in height.
- We distinguish two sub-cases, depending on whether w is the right child or the left child of x.

# **►Insert (4)**

**Sub-case 3-1**: w is the right child of x.

- The tree is lopsided because of an "outside" problem.
- Now rotate the tree to the left: x becomes the parent of v, and x's left subtree B becomes a subtree of v.

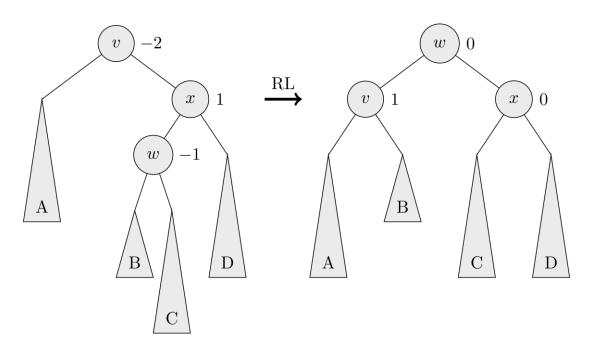


Height of whole subtree is the same as before insert. Done.

# **►Insert (5)**

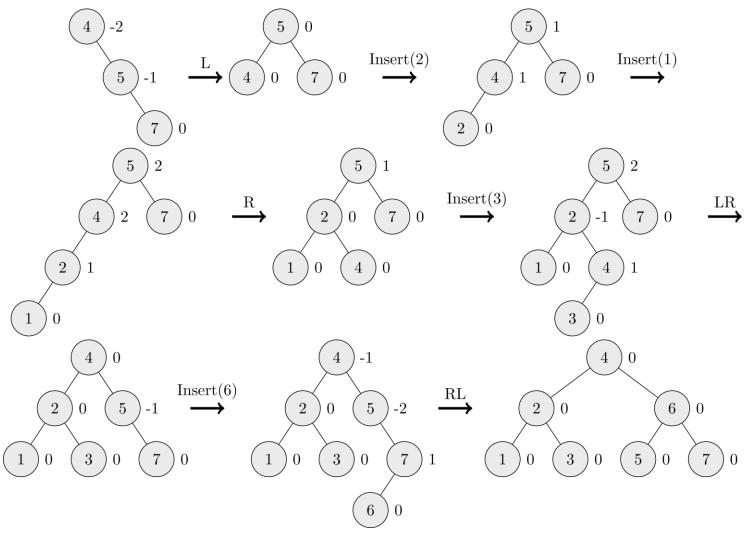
Sub-case 3-2: w is the left child of x.

- The tree is lopsided because of an "inside" problem.
- Now need a double rotation to rebalance the tree: a right rotation at x, followed by an immediate left rotation at v.



NB: heights of B and C could be the other way round.

# ►Insert: Example



#### **►** Runtime of Insert

- Inserting an element takes time O(h).
- Rebalancing:
  - Insert finishes with the first rotation/double rotation.
  - All rotations (L/R/LR/RL) take time O(1).
  - Backing up the search path takes time O(1) for each node on the search path, hence time O(h) overall.
  - This includes the time to update balance factors.
- Total runtime of Insert:  $O(h) = O(\log n)$ .

#### Deleting in an AVL Tree

- Like for Insert, we work backwards up the search path to rebalance so long as the height of the current subtree has decreased.
- Assume without loss of generality that delete decreased the height of the *left* subtree.
- Case 1: bal(v) = 1. Here deletion decreased the height of the higher subtree, leading to bal(v) = 0.
   However, the height of v has decreased, so we need to iterate the rebalance procedure with v's parent.
- Case 2: bal(v) = 0. Then we update bal(v) = -1 and note that the height of v's subtree has not decreased, so the rebalancing is complete.

# Delete (2)

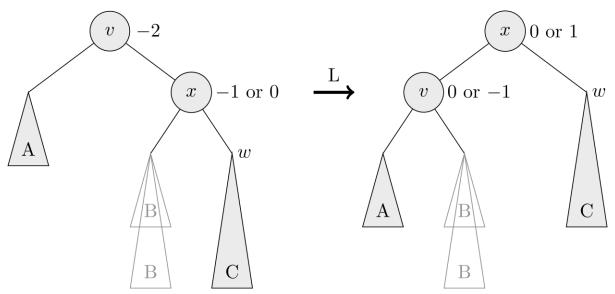
**Case 3**: bal(v) = -1.

- After deletion, the shallower subtree has become even more shallow: bal(v) = -2.
- Consider path of nodes v, x, w whose subtrees are now too high.
- We distinguish two sub-cases, depending on whether w is the right child or the left child of x.

# Delete (3)

**Sub-case 3-1**:  $bal(x) \in \{-1, 0\}$ .

- The tree is lopsided because of an "outside" problem.
- Now rotate the tree to the left.
- Two possibilities for the height of B.

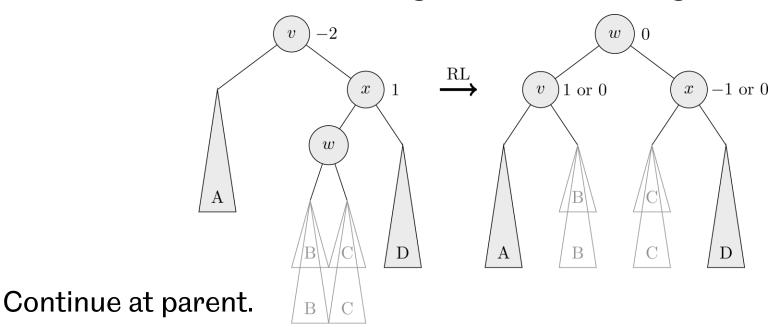


If B was high, we're done. Otherwise, iterated at x's parent.

# Delete (4)

**Sub-case 3-2**: bal(x) = 1.

- The tree is lopsided because of an "inside" problem.
- Again need a double rotation.
- B and C can have one of two heights; one must be high.



#### Runtime of Delete

- Delete may not finish with the first rotation/double rotation.
- Still, the time spent at each node on the search path is O(1), so we still get a time of  $O(h) = O(\log n)$ .

# **▶**Summary

- AVL trees with n elements have height  $O(\log n)$ .
- AVL trees with n nodes execute the following operations in time  $O(\log n)$ 
  - Searching, Minimum, Maximum, Successor
    - Follows since AVL trees are binary search trees whose height is always  $h = O(\log n)$ .
  - Insertion
  - Deletion
- Greater efficiency from a simple idea: rotating nodes.