#### trees

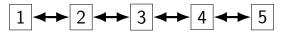
## are lists enough?

for correctness — sure

want to efficiently access items better than linear time to find something

want to represent relationships more naturally

## inter-item relationships in lists



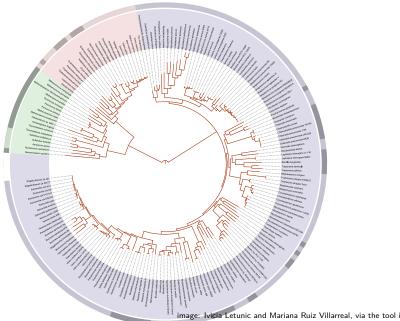
List: nodes related to predecessor/successor

#### trees

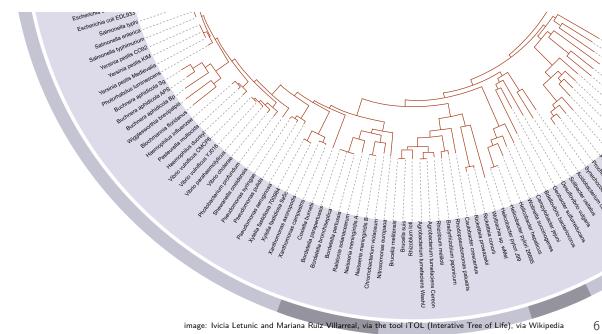
trees: allow representing more relationships (but not arbitrary relationships — see graphs later in semester)

restriction: single path from *root* to every node implies single path from every node to every other node (possibly through root)

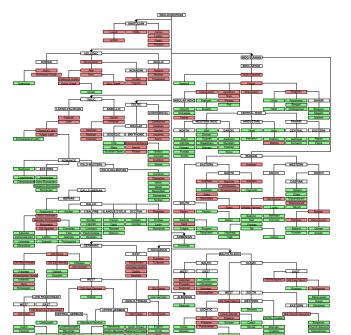
## natural trees: phylogenetic tree



## natural trees: phylogenetic tree (zoom)



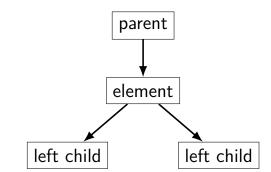
## natural trees: Indo-European languages



#### list to tree

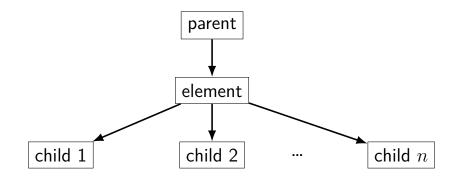
*list* — up to 2 related nodes

binary tree — up to 3 related nodes (list is special-case)

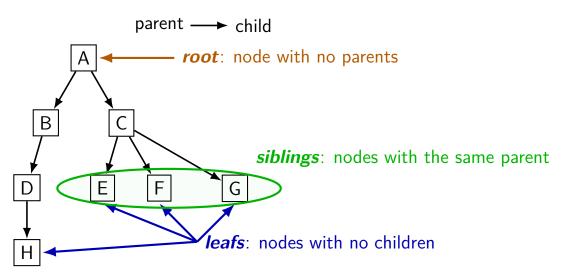


#### more general trees

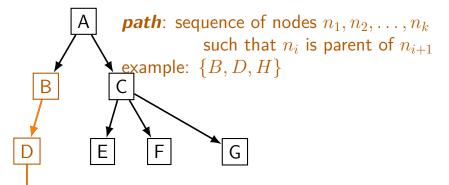
*tree* — any number of relationships (binary tree is special case) at most one parent



## tree terms (1)



## paths and path lengths



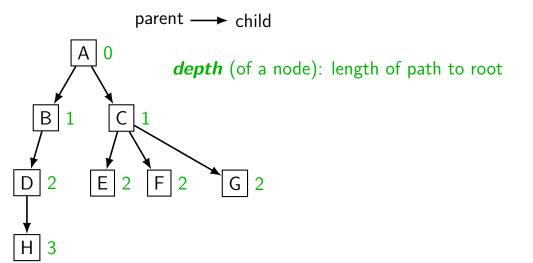
*length* (of path): number of *edges* in path example: 2 ( $B \rightarrow D$  and  $D \rightarrow H$ )

*internal path length*: sum of depth of nodes example: 6 = 1 + 2 + 3

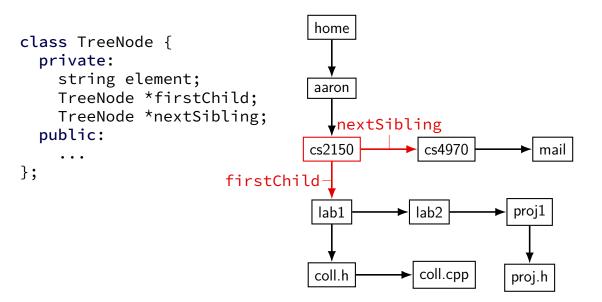
## tree/node height

parent  $\longrightarrow$  child 3 *height* (of a node): length of longest path to leaf В 2 height (of a tree): height of tree's root (this example: 3) G 0 Ε F 0 1 0

## tree/node depth



# first child/next sibling

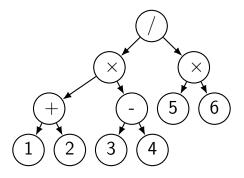


### another tree representations

```
class TreeNode {
   private:
     string element;
     vector<TreeNode *> children;
   public:
     ...
};
```

// and more --- see when we talk about graphs

#### tree traversal



pre-order: / \* + 1 2 - 3 4 \* 5 6in-order: (((1+2) \* (3-4)) / (5\*6)) (parenthesis optional?) post-order: 1 2 + 3 4 - \* 5 6 \* /

## pre/post-order traversal printing

```
(this is pseudocode)
```

```
TreeNode::printPreOrder() {
    this->print();
    for each child c of this:
        c->printPreOrder()
}
```

```
TreeNode::printPostOrder() {
    for each child c of this:
        c->printPostOrder()
    this->print();
```

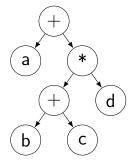
## in-order traversal printing

```
(this is pseudocode)
BinaryTreeNode::printInOrder() {
    if (this->left)
        this->left->printInOrder();
    cout << this->element << "_";
    if (this->right)
        this->right->printInOrder();
```

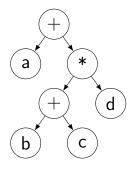
## post-order traversal counting

```
(this is pseudocode)
int numNodes(TreeNode *tnode) {
  if ( tnode == NULL )
      return 0;
  else {
      sum=0;
      for each child c of thode
          sum += numNodes(c);
      return 1 + sum;
  }
```

#### expression tree and traversals



#### expression tree and traversals



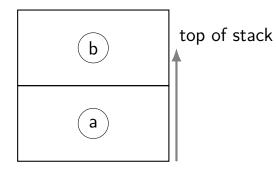
infix: (a + ((b + c) \* d))postfix: a b c + d \* + prefix: + a + b c d

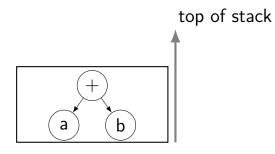
### postfix expression to tree

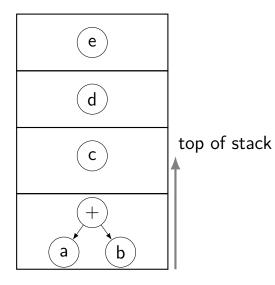
use a stack of trees

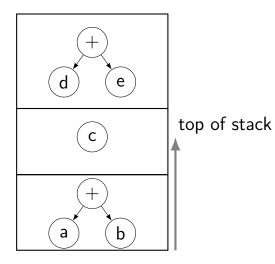
```
number n \to push(n)
```

```
operator OP \rightarrow
pop into A, B; then
push OP
B A
```

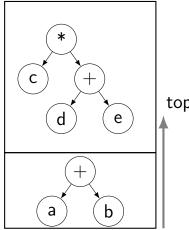




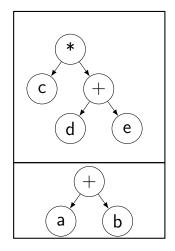


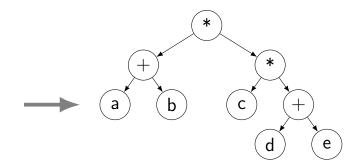


a b + c d e + \* \*

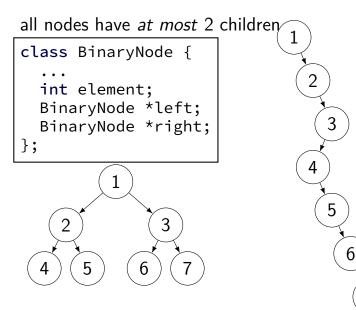


top of stack

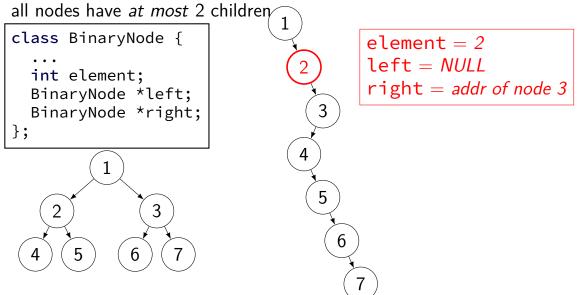




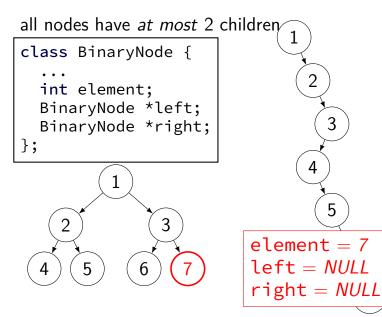
## binary trees



## binary trees



## binary trees



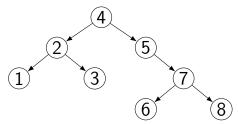
## binary search trees

binary tree and ...

each node has a key

for each node:

keys in node's left subtree are less than node's keys in node's right subtree are greater than node's



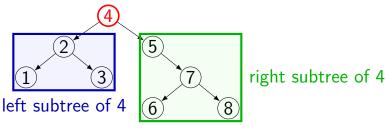
## binary search trees

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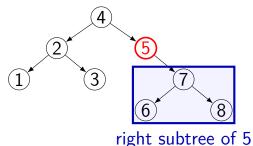
## binary search trees

binary tree and ...

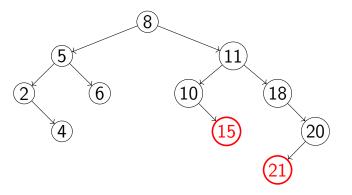
each node has a key

for each node:

keys in node's left subtree are less than node's keys in node's right subtree are greater than node's



## not a binary search tree



#### binary search tree versus binary tree

binary search trees are a kind of binary tree

...but — often people say "binary tree" to mean "binary search tree"

# **BST:** find

```
(pseudocode)
find(node, key) {
    if (node == NULL)
        return NULL;
    else if (key < node->key)
        return find(node->left, key)
    else if (key > node->key)
        return find(node->right, key)
    else // if (key == node->key)
        return node;
```

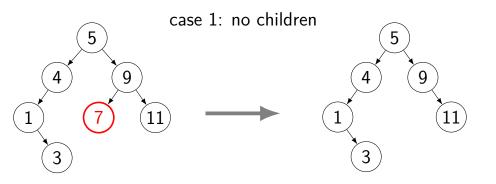
### **BST:** insert

```
(pseudocode)
insert(Node *&node, key) {
    if (node == NULL)
        node = new BinaryNode(key);
    else if (key < node->key)
        insert(node->left, key);
    else if (key < root->key)
        insert(node->right, key);
    else // if (key > root->key)
        ; // duplicate -- no new node needed
```

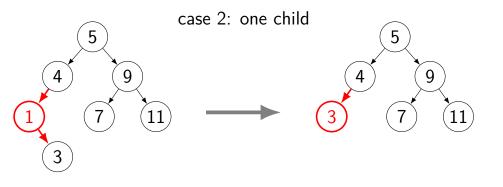
# **BST:** findMin

```
(pseudocode)
findMin(Node *node, key) {
    if (node->left == NULL)
        return node;
    else
        insert(node->left, key);
}
```

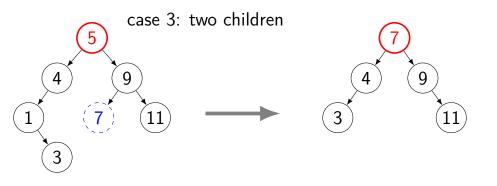
# **BST**: remove (1)



# **BST:** remove (2)



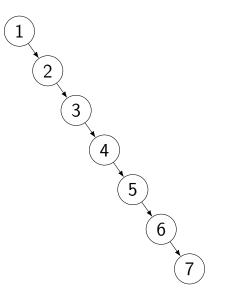
# BST: remove (3)



replace with minimum of right subtree (alternately: maximum of left subtree, ...)

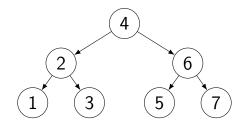
#### binary tree: worst-case height

*n*-node BST: worst-case height/depth n-1



### binary tree: best-case height

height h: at most  $2^{h+1} - 1$  nodes



### binary tree: proof best-case height is possible

proof **by induction**: can have  $2^{h+1} - 1$  nodes in *h*-height tree

 $\mathbf{h} = \mathbf{0}$ : h = 0: exactly one node;  $2^{h+1} - 1 = 1$  nodes

 $\mathbf{h} = \mathbf{k} 
ightarrow \mathbf{h} = \mathbf{k} + \mathbf{1}$ :

start with two copies of a maximum tree of height  $\boldsymbol{k}$ 

create a new tree as follows:

create a new root node

add edges from the root node to the roots of the copies

```
the height of this new tree is k + 1 path of length k in old tree + either new edge
```

the number of nodes is a(k+1) + 1 + a(k+1) + a(k

#### binary tree: best-case height is best

(informally)

property of trees in root:

except for the leaves, every node in tree has 2 children

no way to add nodes without increasing height add below leaf — longer path to root — longer height add above root — every old node has longer path to root

### binary tree height formula

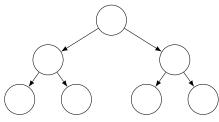
n: number of nodes

h: height

$$\begin{array}{rcl}
n+1 &\leq 2^{h+1} \\
\log_2(n+1) &\leq \log_2\left(2^{h+1}\right) \\
\log(n+1) &\leq h+1 \\
h &\geq \log_2\left(n+1\right) - 1
\end{array}$$

shortest tree of n nodes:  $\sim \log_2(n)$  height

# perfect binary trees



a binary tree is perfect if

all leaves have same depth all nodes have zero children (leaf) or two children

exactly the trees that achieve  $2^{h+1} - 1$  nodes

### **AVL** animation tool

http://webdiis.unizar.es/asignaturas/EDA/
AVLTree/avltree.html

#### AVL tree idea

AVL trees: one of many balanced trees — search tree balanced to keep height  $\Theta(\log n)$  avoid "tree is just a long linked list" scenarios

gaurentees  $\Theta(\log n)$  for find, insert, remove

AVL = Adelson-Velskii and Landis

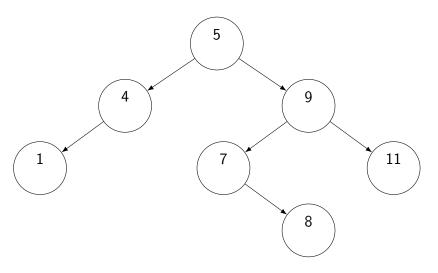
### **AVL** gaurentee

the height of the left and right subtrees of *every node* differs by at most one

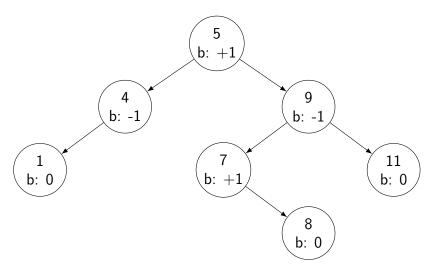
#### **AVL** state

```
normal binary search tree stuff:
data; and left, right, parent pointers
additional AVL stuff:
height of right subtree minus height of left subtree
called "balance factor"
-1, 0, +1
(kept up to date on insert/delete — computing on demand is too slow)
```

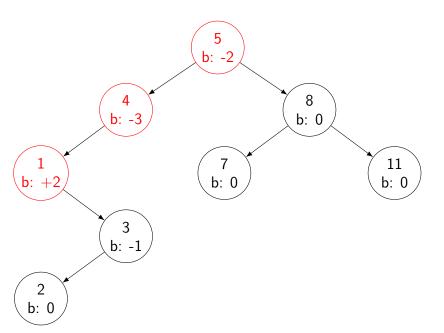
# example AVL tree



### example AVL tree



#### example non-AVL tree



### **AVL** tree algorithms

find — exactly the same as binary search tree just ignore balance factors

insert — two extra steps: update balance factors "fix" tree if it became unbalanced

### **AVL** tree algorithms

find — exactly the same as binary search tree just ignore balance factors

insert — two extra steps: update balance factors "fix" tree if it became unbalanced

runtime for both  $\Theta(d)$  where d is depth of node found/inserted max balance factor  $\pm 1$  at root max depth of node is  $\Theta(\log_2 n+1)=\Theta(\log n)$ 

#### **AVL** insertion cases

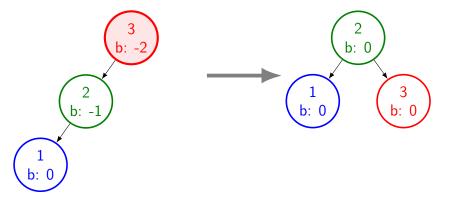
simple case: tree remains balanced

otherwise:

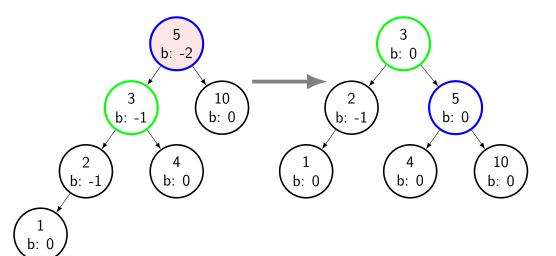
let x be deepest imbalanced node (+2/-2 balance factor) insert in left subtree of left child of x: single rotation right insert in right subtree of right child of x: single rotation left insert in right subtree of left child of x: double left-right rotation insert in left subtree of right child of x: double right-left rotation

#### **AVL: simple right rotation**

just inserted 0 unbalanced root becomes new left child

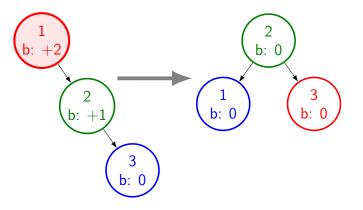


just inserted 0 unbalanced root becomes new left child



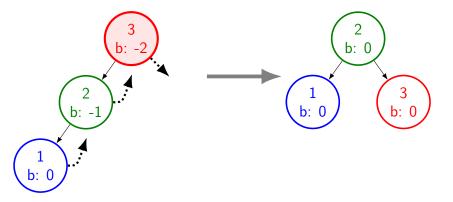
### **AVL: simple left rotation**

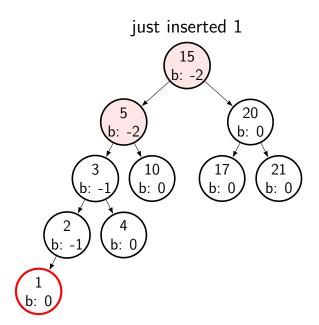
just inserted 1 deepest unbalanced node is 3

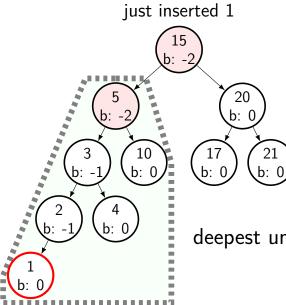


#### AVL rotation: up and down

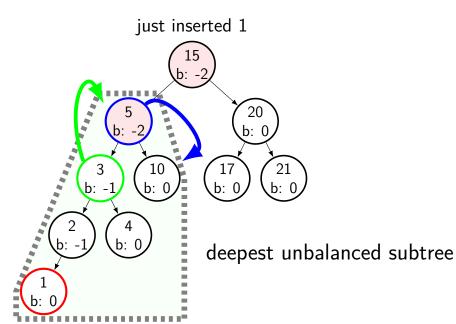
*at least* one node moves up (this case: 1 and 2) *at least* one node moves down (this case: 3)

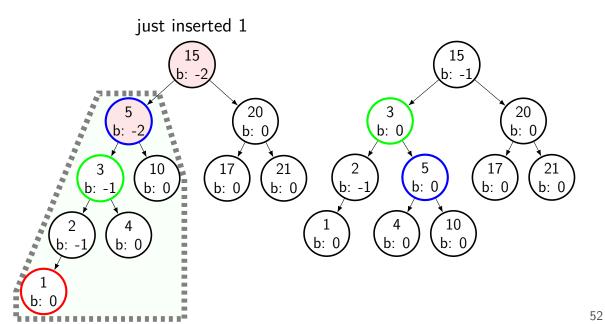




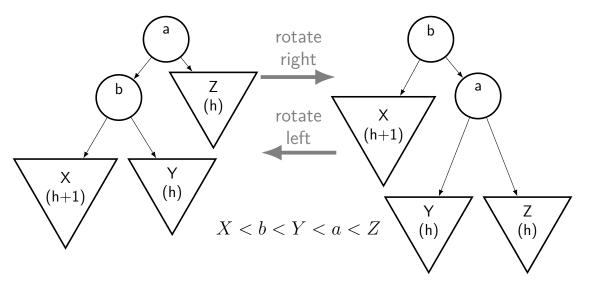


deepest unbalanced subtree

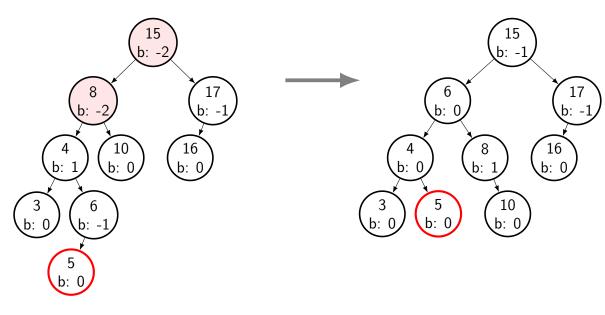


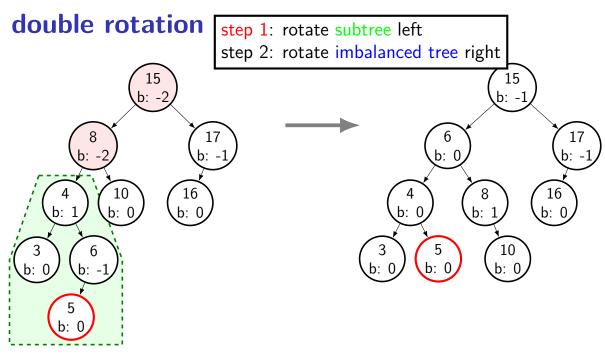


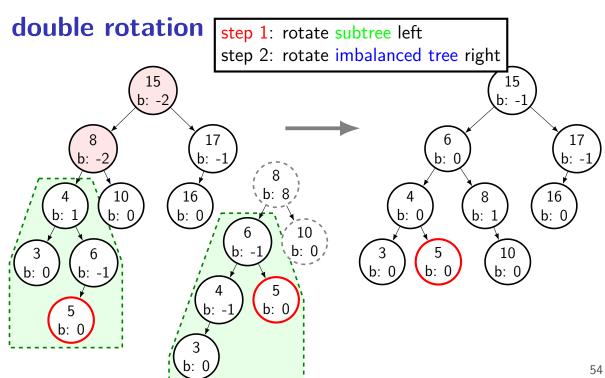
# general single rotation

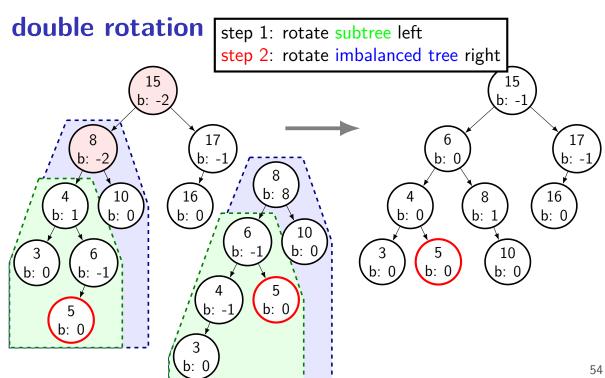


### double rotation

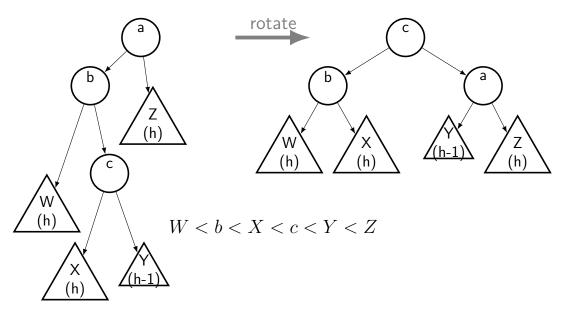




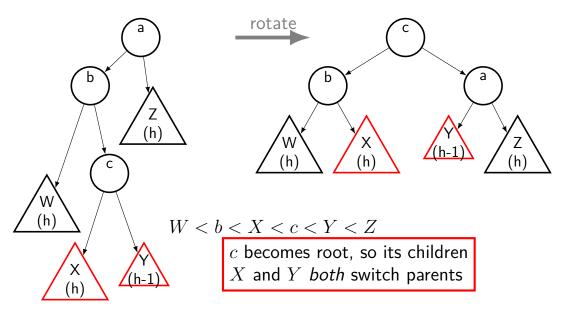




# general double rotation



# general double rotation



# double rotation names

- sometimes "double left" first rotation left, or second?
- us: "double left-right" rotate child tree left rotate parent tree right
- "double right-left" rotate child tree right rotate parent tree left

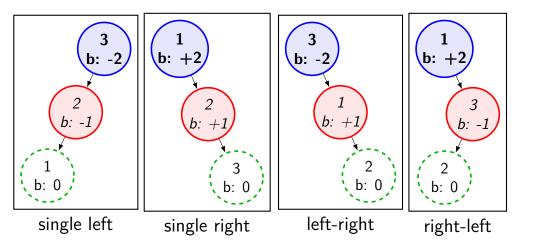
## **AVL** insertion cases

simple case: tree remains balanced

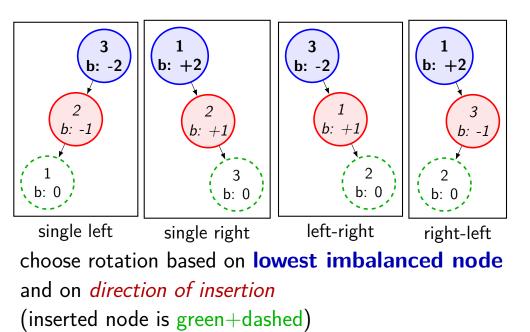
otherwise:

let x be deepest imbalanced node (+2/-2 balance factor) insert in left subtree of left child of x: single rotation right insert in right subtree of right child of x: single rotation left insert in right subtree of left child of x: double left-right rotation insert in left subtree of right child of x: double right-left rotation

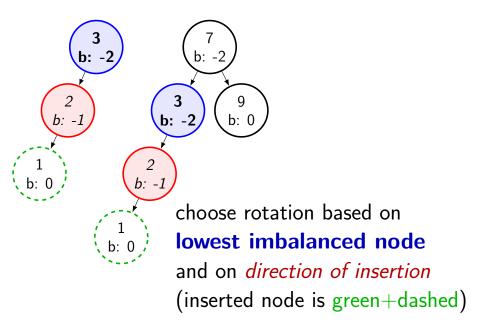
## AVL insert cases (revisited)



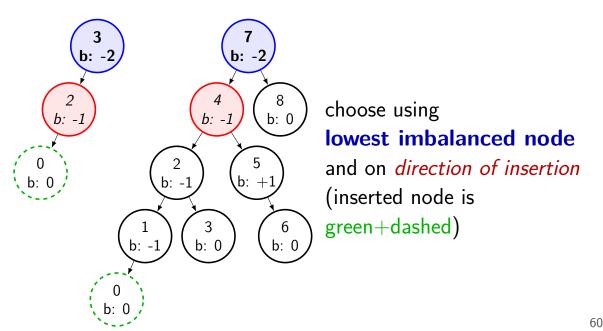
## AVL insert cases (revisited)



# AVL insert case: detail (1)



# AVL insert case: detail (2)



There are 4 cases in all, choosing which Left Left Case **Right Right Case** Left Right Case **Right Left Case** one is made by seeing the direction Root of the first 2 nodes Root from the unbalanced node to the newly Pivot Pivot inserted node and /Α Root Root matching them to the top most row. Pivot Pivot Root is the initial parent before a rotation and Pivot is the child to take the root's place. Right Rotation Left Left Right Rotation Rotation Rotation Root Root Pivot Pivot /A Left Right Rotation Rotation

## **AVL tree: runtime**

worst depth of node:  $\Theta(\log_2 n + 2) = \Theta(\log n)$ 

find:  $\Theta(\log n)$ 

worst case: traverse from root to worst depth leaf

insert:  $\Theta(\log n)$ 

worst case: traverse from root to worst depth leaf then back up (update balance factors) then perform constant time rotation

remove:  $\Theta(\log n)$ 

left as exercise (similar to insert)

 $\begin{array}{l} \text{print: } \Theta(n) \\ \text{visit each of } n \text{ nodes} \end{array}$ 

# other types of trees

many kinds of balanced trees

not all binary trees

different ways of tracking balance factors, etc.

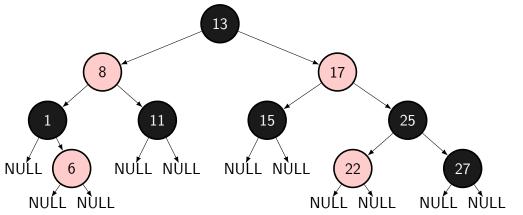
different ways of doing tree rotations or equivalent

#### red-black trees

each node is **red** or **black** 

null leafs considered nodes to aid analysis (still null pointers...)

rules about when nodes can be red/black gaurentee maximum depth



#### red-black tree rules

root is **black** 

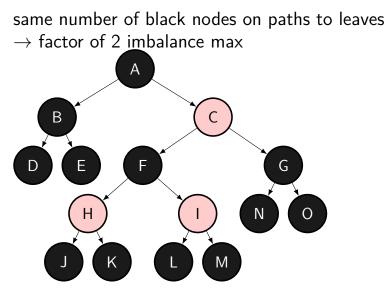
counting null pointers as nodes, leaves are **black** 

- a **red** node's children are **black** 
  - $\rightarrow$  a red node's parents are black

every simple path from node to leaf under it contains same number of black nodes

(property holds regardless of whether null pointers are considered nodes)

### worst red-black tree imbalance

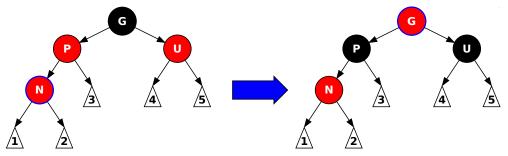


- (1) if new node is root: color **black**
- (2) if parent is black: keep child red
- (3) if parent and uncle is red: adjust several colors
- (4) if parent is **red**, uncle is **black**, new node is right child perform a rotation, then go to case 5
- (5) if parent is **red**, uncle is **black**, new node is left child perform a rotation

- (1) if new node is root: color **black**
- (2) if parent is black: keep child red
- (3) if parent and uncle is red: adjust several colors
- (4) if parent is **red**, uncle is **black**, new node is right child perform a rotation, then go to case 5
- (5) if parent is property: "children of **red** node are **black**" perform a no change in # of **black** nodes on paths

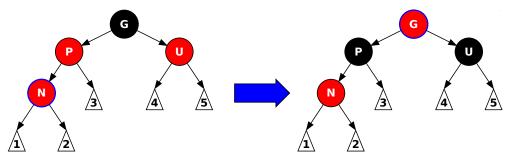
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### case 3: parent, uncle are red



make grandparent **red**, parent and uncle **black** (property: every path to leaf has same number of black nodes) just swapped grandparent and parent/uncle in those paths

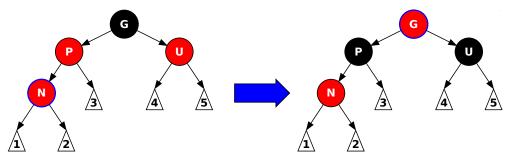
### case 3: parent, uncle are red



make grandparent **red**, parent and uncle **black** (property: every path to leaf has same number of black nodes) just swapped grandparent and parent/uncle in those paths

but...what if grandparent's parent is red? (property: children of red node are black) solution: recurse to the grandparent, as if it was just inserted

### case 3: parent, uncle are red

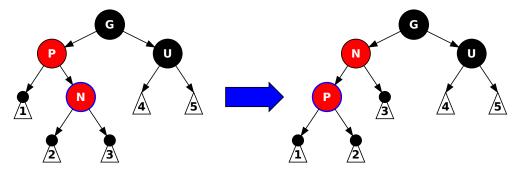


make grandparent **red**, parent and uncle **black** (property: every path to leaf has same number of black nodes) just swapped grandparent and parent/uncle in those paths

but...what if grandparent's parent is red? (property: children of red node are black) solution: recurse to the grandparent, as if it was just inserted

- (1) if new node is root: color **black**
- (2) if parent is black: keep child red
- (3) if parent and uncle is red: adjust several colors
- (4) if parent is **red**, uncle is **black**, new node is right child perform a rotation, then go to case 5
- (5) if parent is **red**, uncle is **black**, new node is left child perform a rotation

## case 4: parent red, uncle black, right child

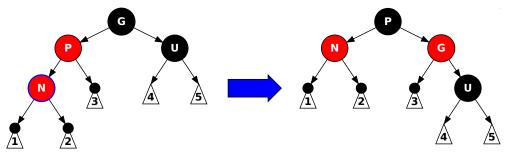


perform left rotation on parent subtree and new node

now case 5 (but new node is P, not N)

- (1) if new node is root: color **black**
- (2) if parent is black: keep child red
- (3) if parent and uncle is red: adjust several colors
- (4) if parent is **red**, uncle is **black**, new node is right child perform a rotation, then go to case 5
- (5) if parent is **red**, uncle is **black**, new node is left child perform a rotation

## case 5: parent red, uncle black, left child

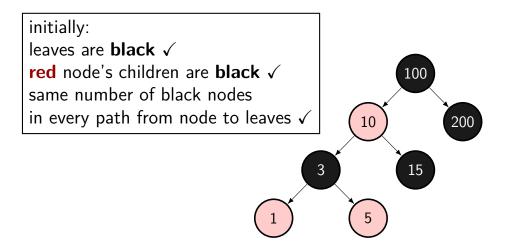


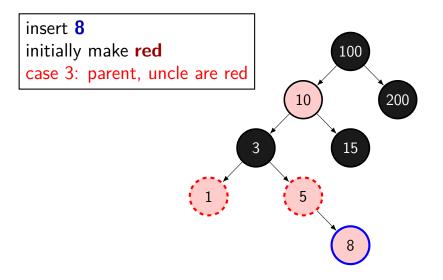
perform right rotation of grandparent and parent

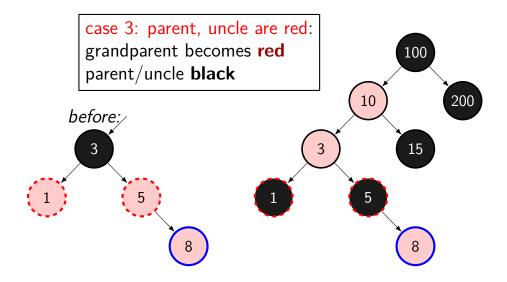
swap colors of parent and grandparent

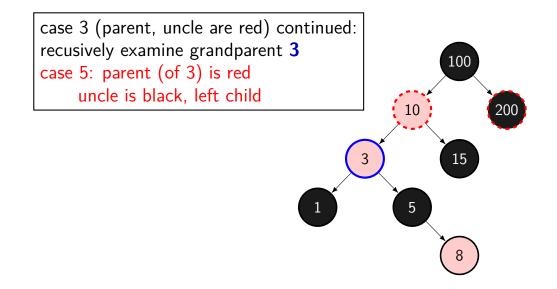
preserves properties:

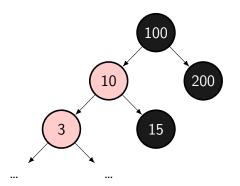
red parent's children are black every path to leaf has same number of black nodes

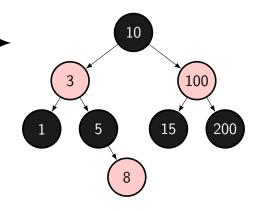


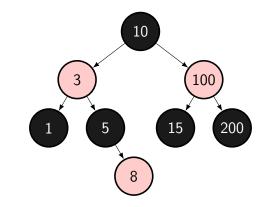












#### **RB-tree: removal**

start with normal BST remove of x, but...

instead find next highest/lowest node y can choose node with at most one child ("bottom" of a left or right subtree)

swap x and y's value, then replace y with its child

several cases for color maintainence/rotations

### **RB tree: removal cases**

N: node just replaced with child; S: its sibling; P: its parent

- (1): N is new root
- (2): S is red
- (3): P, S, and S's children are **black**
- (4): S and S's children are **black**

(5): S is **black**, S's left child is **red**, S's right child is **black**, N is left child of P

(6): S is **black**, S's right child is **red**, N is left child

 $A_{i} = A_{i} = A_{i$ 

## why red-black trees?

- a lot more cases...but
- a lot less rotations
- ...because tree is kept less rigidly balanced
- red-black trees end up being faster in practice

### more balanced trees

several other kinds of balanced trees

one notable kind: non-binary balanced trees

commonly used in databases

more efficient to store multiple nodes together on disk/SSD

### splay trees

tree that's fast for recently used nodes

self-balancing binary search tree

keeps recent nodes near the top

simpler to implement than AVL or RB trees

# 'splaying'

every time node is accessed (find, insert, delete)...

"splay" tree around that node

make the node the new tree root

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 $\Theta(h)$  time — where h is tree height

## 'splaying'

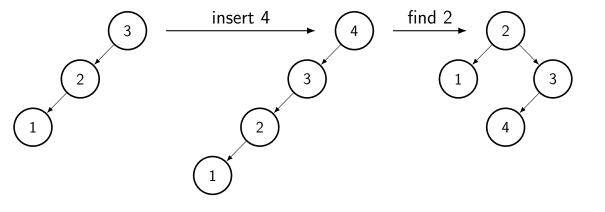
every time node is accessed (find, insert, delete)...

"splay" tree around that node

make the node the new tree root

 $\Theta(h)$  time — where h is tree height worst-case height:  $\Theta(n)$  — linked-list case

### splay tree operations



### amortized complexity

splay tree insert/find/delete is amortized  $O(\log n)$  time

informally: average insert/find/delete:  $O(\log n)$ 

more formally: m operations:  $O(m\log n)$  time (where n: max size of tree)

## splay tree pro/con

can be *faster* than AVL, RB-trees in practice take advantage of frequently accessed items

simpler to implement

but worst case find/insert is  $\Theta(n)$  time

### last time

red-black trees

less well-balanced than AVL trees track color instead of balance factor rules about colors to limit possible imbalance algorithm for insertion, etc. that makes tree always obey rules usually faster in practice — less rotations

splay trees

optimized for repeated accesses keep recently accessed items near top of tree find rearranges tree! *amortized* logarithmic time (but worst case is linear)

### amortized analysis: vector growth

vector insert algorithm:

if not big enough, double capacity write to end of vector

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vector insert algorithm:

if not big enough, double capacity write to end of vector

doubling size — requires copying! —  $\Theta(n)$  time

 $\Theta(n)$  worst case per insert

but average...?

## counting copies (1)

suppose initial capacity 100 + insert 1600 elements

 $\begin{array}{l} 100 \rightarrow 200 \text{: } 100 \text{ copies} \\ 200 \rightarrow 400 \text{: } 200 \text{ copies} \\ 400 \rightarrow 800 \text{: } 400 \text{ copies} \\ 800 \rightarrow 1600 \text{: } 800 \text{ copies} \\ \text{total: } 1500 \text{ copies} \end{array}$ 

total operations:  $1500 \ {\rm copies} + 1600 \ {\rm writes} \ {\rm of} \ {\rm new} \ {\rm elements}$  about  $2 \ {\rm operations} \ {\rm per} \ {\rm insert}$ 

# counting copies (2)

more generally: for N inserts

about N copies + N writes why? K to 2K elements: K copies N inserts:  $1 + 2 + 4 + \ldots + N/4 + N/2 = N - 1$  copies (and a bit better if initial capacity isn't 1)

# counting copies (2)

```
more generally: for N inserts
about N copies + N writes
why? K to 2K elements: K copies
N inserts: 1 + 2 + 4 + \ldots + N/4 + N/2 = N - 1 copies
(and a bit better if initial capacity isn't 1)
```

 $\Theta(n)$  worst case

but  $\Theta(n)$  time for n inserts

 $\rightarrow O(1)$  amortized time per insert

## other vector capacity increases? (1)

instead of doubling...add 1000

N inserts:  $1000 + 2000 + 3000 + \ldots + N \sim N^2$ 

 $\rightarrow \Theta(N^2)$  total — O(N) amortized time per insert

## other vector capacity increases? (1)

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increase by constant: linear worst-case and amortized

instead of doubling...multiply by k>1 e.g. k=1.1 — increase by 10%

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e.g. k = 1.1 — increase by 10%

N inserts: 
$$1 + k + k^2 + k^3 + \ldots + k^{\log_k N} = \frac{1 - k^{\log_k N}}{1 - k} \sim N$$

instead of doubling...multiply by k > 1

e.g. k = 1.1 — increase by 10%

N inserts: 
$$1 + k + k^2 + k^3 + \ldots + k^{\log_k N} = \frac{1 - k^{\log_k N}}{1 - k} \sim N$$

 $\rightarrow \Theta(N)$  total — O(1) amortized time per insert

amortized constant time for all k > 1

### trees are not great for...

- ordered, unsorted lists
- list of TODO tasks
- being easy/simple to implement

```
compare, e.g., stack/queue
```

```
\Theta(1) time
```

- compare vector
- compare hashtables (almost)

#### programs as trees

program int z; functions vars int foo (int x) { (foo() main() (int z)for (int y = 0; params vars body params vars v < x; y++) (int z) for (int z) cout << y << endl;</pre> int y  $\sqrt{v} < x$ body v+-=5 cout endl int main() { У int z = 5;body cout << "enter x" << endl; foo() cin cout cin >> z;foo(z); "enter z" end z 7

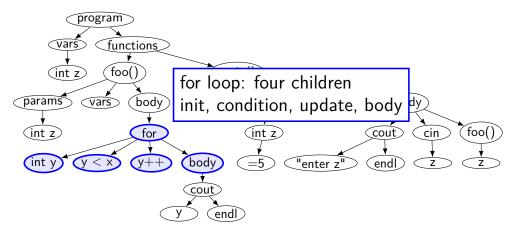
programs as trees

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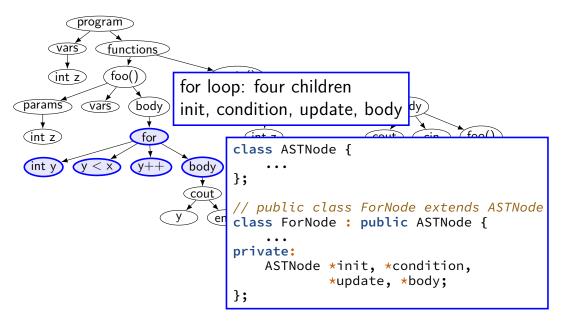
#### abstract syntax tree



#### abstract syntax tree



### abstract syntax tree



## **AST** applications

- "abstract syntax tree" = "parse tree"
- part of how compilers work
- do some tree traversal to do...
- code generation e.g. ASTNode::outputCode() method
- optimization
- type checking...

## using AST to compare programs

comparing trees is a good way to compare programs... while ignoring:

function/method order (e.g. sort function nodes by length)

variable names (e.g. ignore variable names when comparing)

comments

...

part of many software plagerism/copy+paste detection tools