



INF421, Lecture 5

Balanced Trees

Leo Liberti

LIX, École Polytechnique, France



Course

- **Objective:** teach notions AND develop intelligence
- **Evaluation:** TP noté en salle info, Contrôle à la fin. Note:
 $\max(CC, \frac{3}{4}CC + \frac{1}{4}TP)$
- **Organization:** fri 31/8, 7/9, 14/9, 21/9, 28/9, 5/10, 12/10, 19/10, 26/10,
amphi 1030-12 (Arago), TD 1330-1530, 1545-1745 (SI:30-34)
- **Books:**
 1. K. Mehlhorn & P. Sanders, *Algorithms and Data Structures*, Springer, 2008
 2. D. Knuth, *The Art of Computer Programming*, Addison-Wesley, 1997
 3. G. Dowek, *Les principes des langages de programmation*, Editions de l'X, 2008
 4. Ph. Baptiste & L. Maranget, *Programmation et Algorithmique*, Ecole Polytechnique (Polycopié), 2006
- **Website:** www.enseignement.polytechnique.fr/informatique/INF421
- **Blog:** inf421.wordpress.com
- **Contact:** liberti@lix.polytechnique.fr (e-mail subject: INF421)

Lecture summary



- Binary search trees
- AVL trees
- Heaps and priority queues
- Tries



Notation

Tree T	node v	root node $r(T)$
$L(T)$: left subtree of $r(T)$	$R(T)$: right subtree of $r(T)$	depth $D(T)$
$L(v)$: left subnode of v	$R(v)$: right subnode of v	
$L(T) = R(T) = \emptyset$: leaf	$T = \langle L(T), r(T), R(T) \rangle$	$P(v)$: parent of v
$p(v)$: unique path $r(T) \rightarrow v$	path length: $\sum_v p(v) $	$D(T) = \max_v p(v) $



Binary search trees (BST)

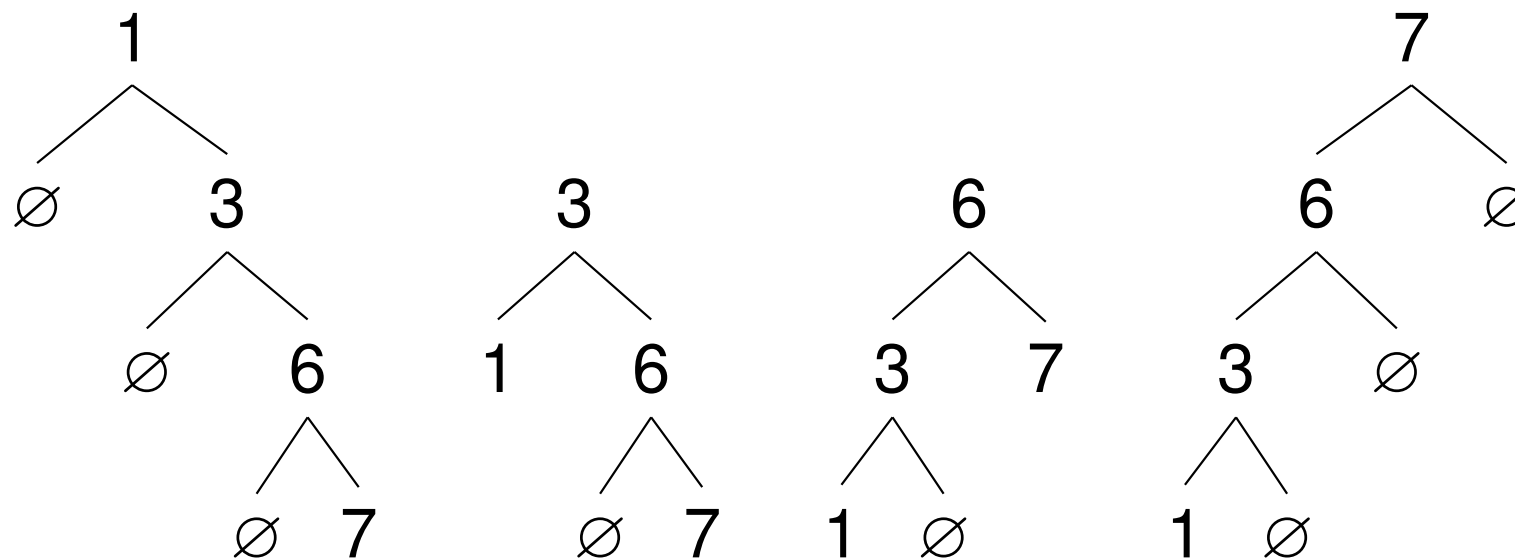


Sorted sequences

- Store a set V as a **sorted sequence**
- Answer the question $v \in V$ efficiently
- **Invariant** :

$$\boxed{L(v) < v < R(v)} \quad (*)$$

- Example: $V = \{1, 3, 6, 7\}$

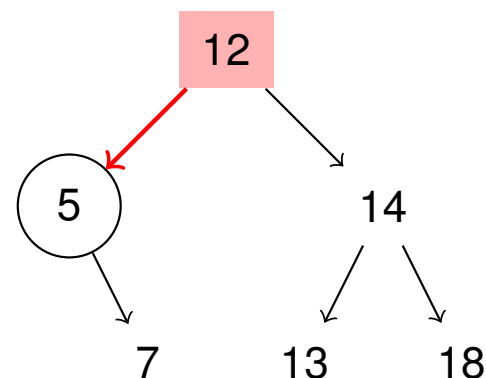




BST min/max

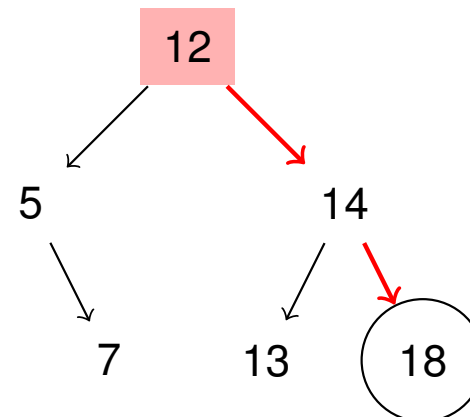
● $\text{min}(v)$:

- 1: **if** $L(v) = \emptyset$ **then**
- 2: **return** v ;
- 3: **else**
- 4: **return** $\text{min}(L(v))$;
- 5: **end if**



● $\text{max}(v)$:

- 1: **if** $R(v) = \emptyset$ **then**
- 2: **return** v ;
- 3: **else**
- 4: **return** $\text{max}(R(v))$;
- 5: **end if**





Base cases for recursion

All other BST functions $f(k, v)$:

$f(k, \emptyset)$ returns without doing anything



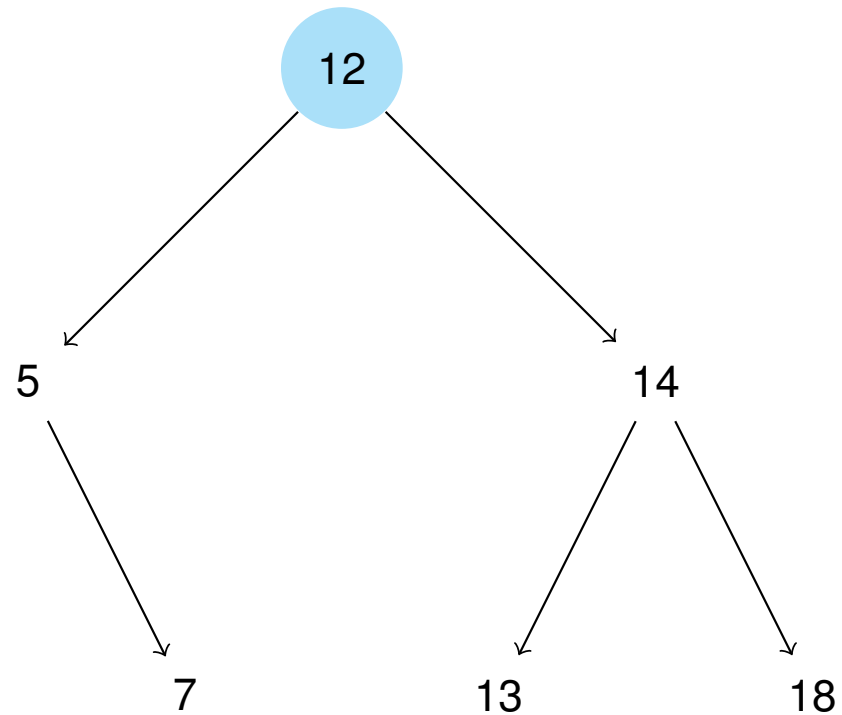
BST find

```
● find( $k, v$ ):  
  1: ret = not_found;  
  2: if  $v = k$  ( $\Rightarrow$  " $v$  stores  $k$ ") then  
  3:   ret =  $v$ ;  
  4: else if  $k < v$  then  
  5:   ret = find( $k, L(v)$ );  
  6: else  
  7:   ret = find( $k, R(v)$ );  
  8: end if  
  9: return ret;
```



Successful find

$\text{find}(13, r(T))$

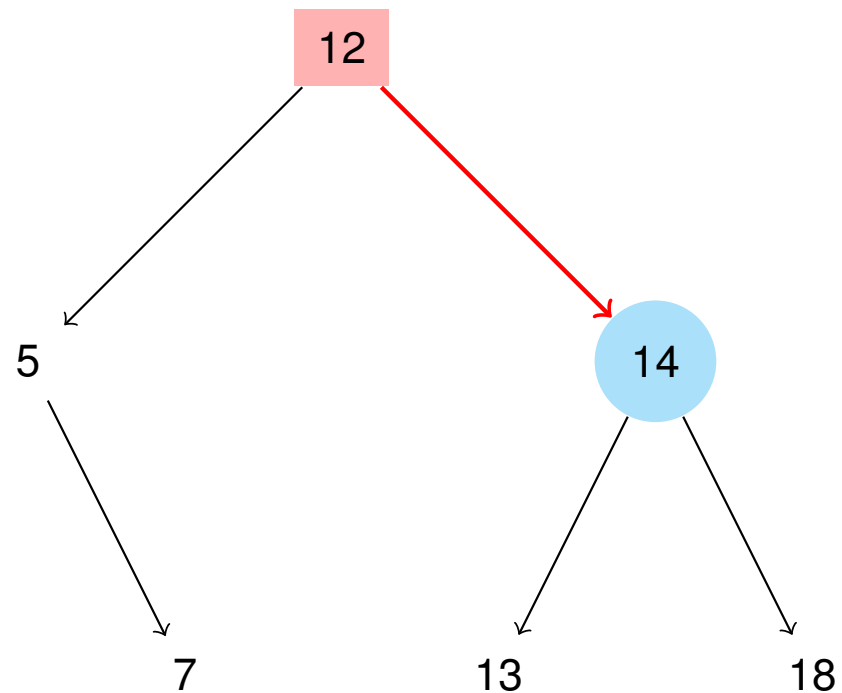


$13 > 12$, take right branch



Successful find

$\text{find}(13, r(T))$

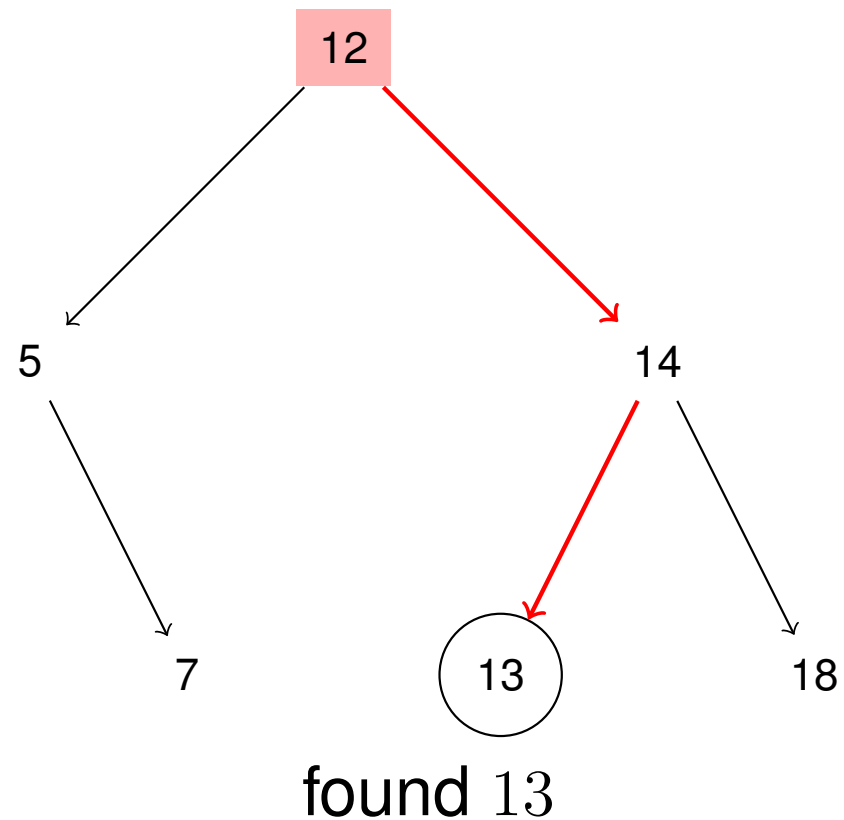


$13 < 14$, take left branch



Successful find

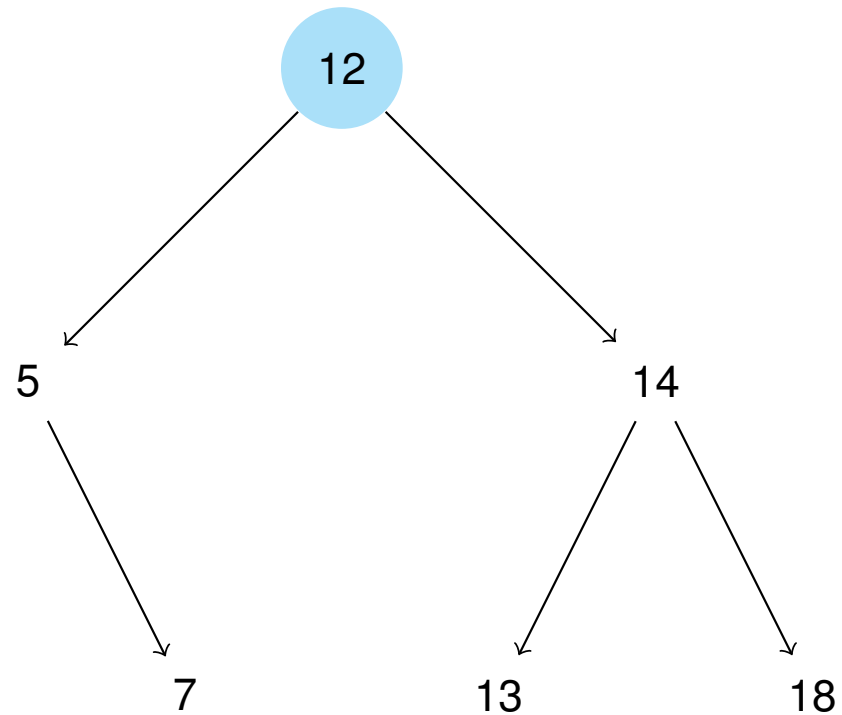
$\text{find}(13, r(T))$





Unsuccessful find

$\text{find}(1, r(T))$

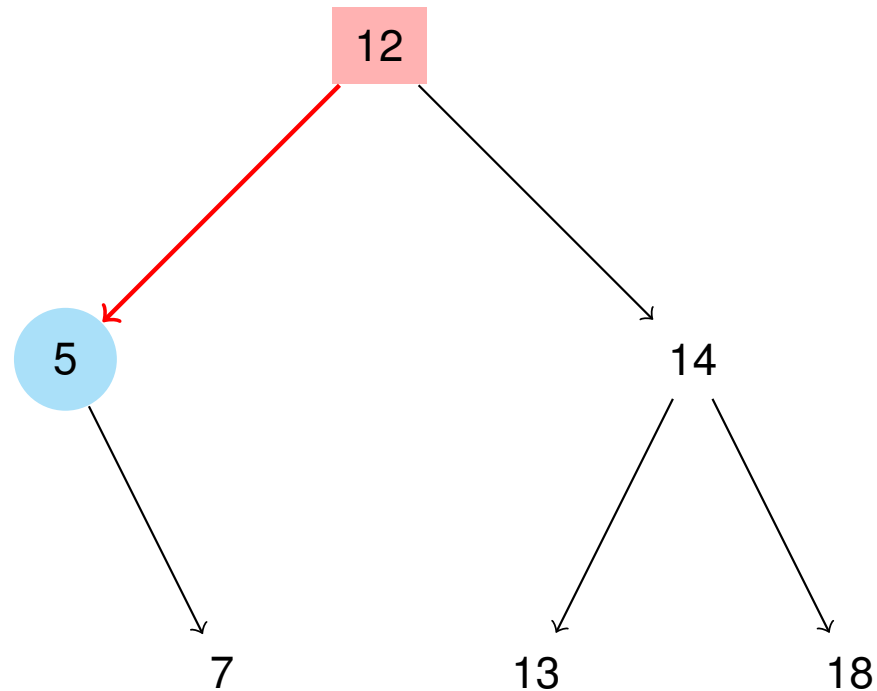


$1 < 12$, take left branch



Unsuccessful find

$\text{find}(1, r(T))$



$1 < 5$, should take left branch but $L(5) = \emptyset$, not found



BST insert



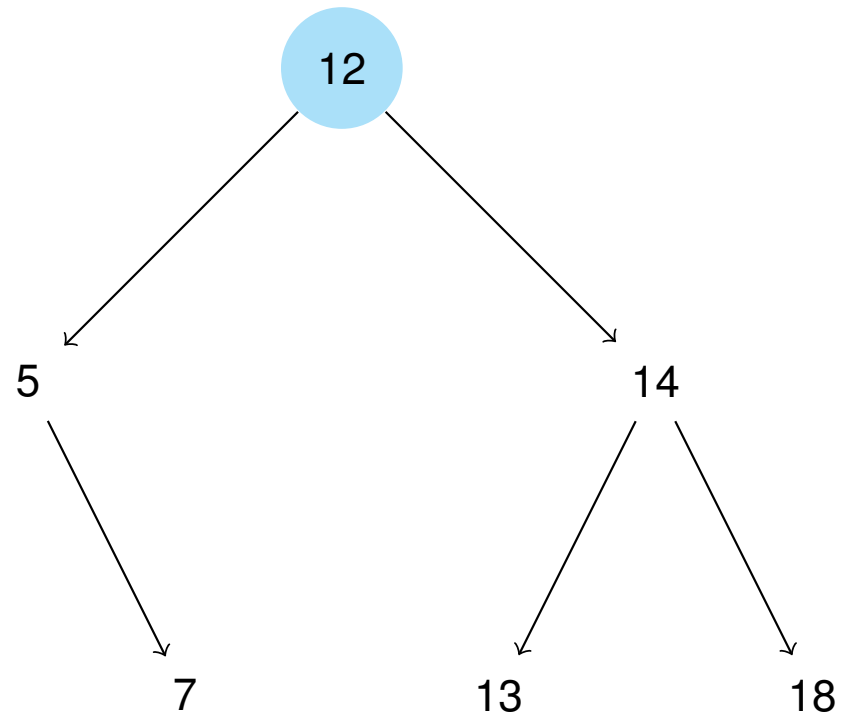
insert(k, v):

```
1: if  $k = v$  then
2:   return already_in_set;
3: else if  $k < v$  then
4:   if  $L(v) = \emptyset$  then
5:      $L(v) = k$ ;           // store  $k$  in  $L(v)$ 
6:   else
7:     insert( $k, v$ );
8:   end if
9: else
10:  if  $R(v) = \emptyset$  then
11:     $R(v) = k$ ;           // store  $k$  in  $R(v)$ 
12:  else
13:    insert( $k, R(v)$ );
14:  end if
15: end if
```



Insert example

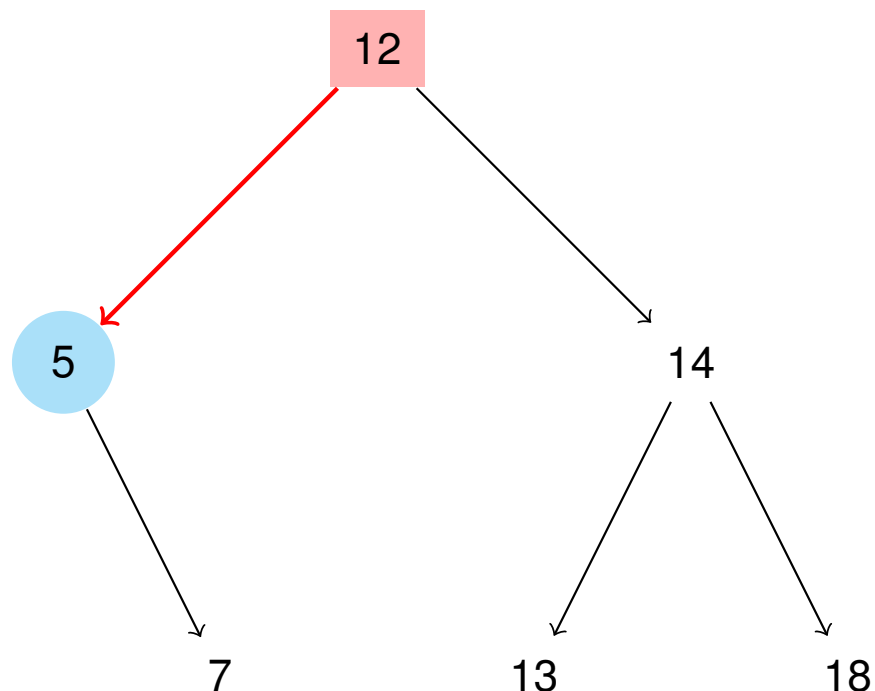
$\text{insert}(1, r(T))$



$1 < 12$, take left branch

Insert example

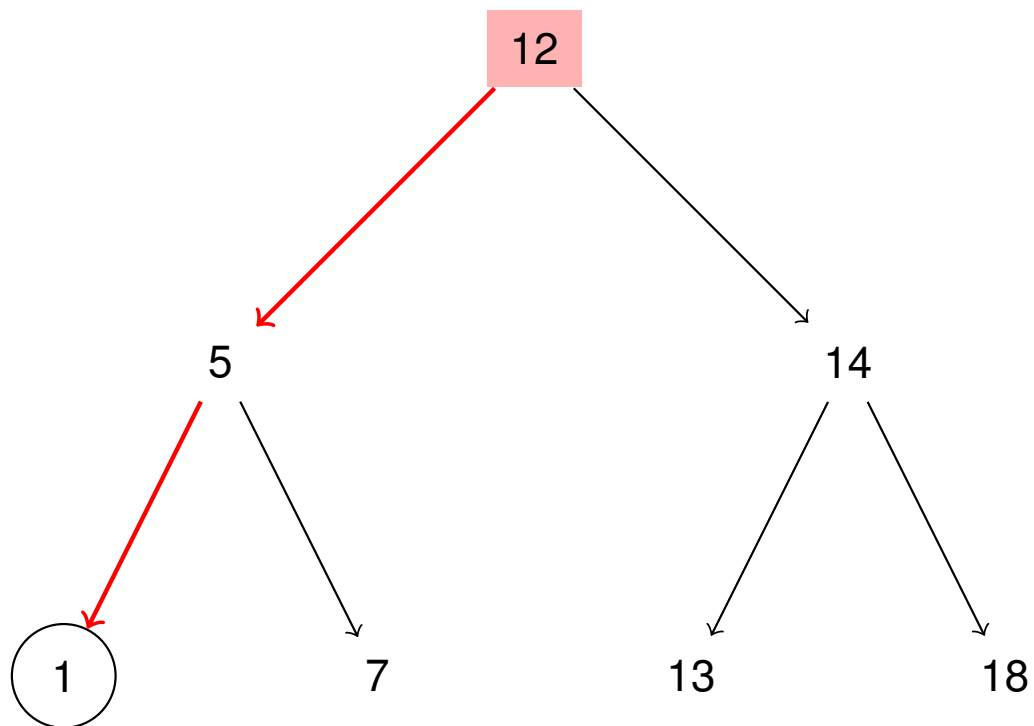
$\text{insert}(1, r(T))$



$1 < 5$, should take left branch but $L(5) = \emptyset$

Insert example

$\text{insert}(1, r(T))$

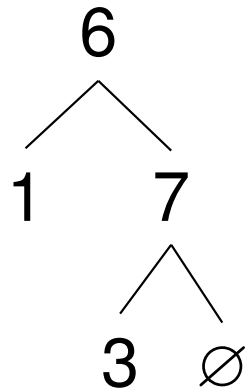


Add $k = 1$ as $L(5)$



A global invariant

- $L(v) \leq v \leq R(v)$ only involves direct subnodes of v
- \Rightarrow it is local
- Is this tree possible?

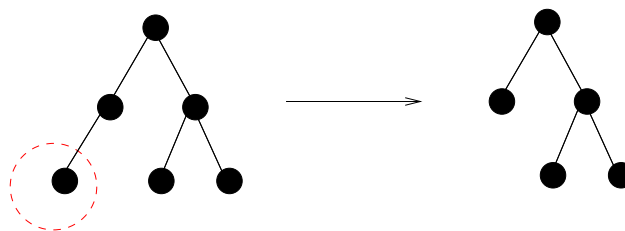


- It satisfies invariant
- By `insert`, 3 would be stored in $R(1)$
- \Rightarrow Invariant is global:

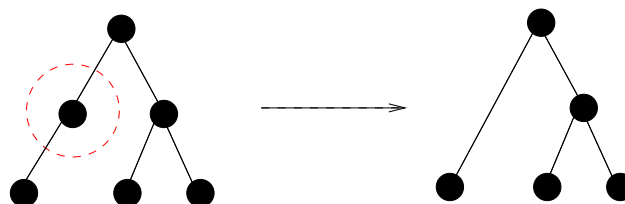
$$\forall u \in \text{tree}(L(v)), w \in \text{tree}(R(v)) \quad u < v < w$$

Deletion

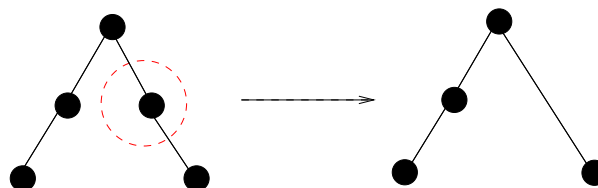
- If node v to delete is a leaf, easy: “cut” it (`unlink`)



- If $R(v) = \emptyset$ and $L(v) \neq \emptyset$, replace with $L(v)$

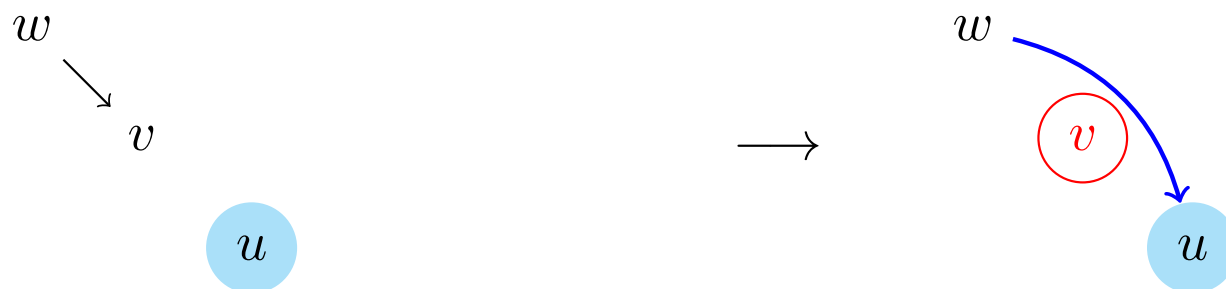


- If $L(v) = \emptyset$ and $R(v) \neq \emptyset$, replace with $R(v)$



- If v has both subtrees, nontrivial

Replacing a node



Replace link $\{P(v), v\}$ with $\{P(v), u\}$, then unlink v

● `replace(v, u) // replace v with u`

1: **if** $R(P(v)) = v$ (i.e. u is a right subnode) **then**

2: $R(P(v)) \leftarrow u;$

3: **else**

4: $L(P(v)) \leftarrow u;$

5: **end if**

6: **if** $u \neq \emptyset$ **then**

7: $P(u) \leftarrow P(v);$

8: **end if**

9: `unlink v;`

● `unlink: set $L(v) = R(v) = P(v) = \emptyset$`



Deleting $v : L(v) \neq \emptyset \wedge R(v) \neq \emptyset$

Idea: swap v with $u = \min R(v)$ then delete it

Thm.

Invariant $L(v) \leq v < R(v)$ holds after swap

- Min of a BST: **leftmost node without left subtree**
- \Rightarrow Can delete u (case $L(\cdot) = \emptyset$ above)
- After swap (u, v) , $v = \min(R(v))$; hence $v < R(v)$
- Before swap $u \in \text{tree}(R(v)) \Rightarrow$ after swap $v > L(v)$
- \Rightarrow Thm. holds



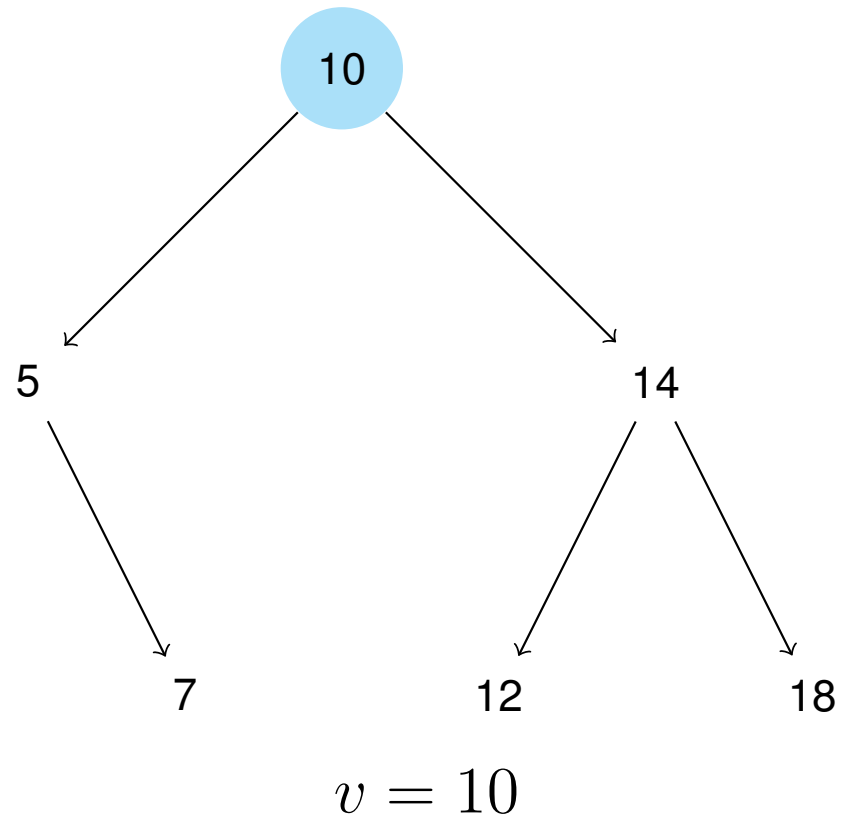
BST delete

```
● delete( $k, v$ ):  
  1: if  $k < v$  then  
  2:   delete( $k, L(v)$ );  
  3: else if  $k > v$  then  
  4:   delete( $k, R(v)$ );  
  5: else  
  6:   if  $L(v) = \emptyset \vee R(v) = \emptyset$  then  
  7:     delete  $v$ ; // one of the easy cases  
  8:   else  
  9:      $u = \min(R(v))$ ;  
 10:    swap_values( $u, v$ );  
 11:    delete  $u$ ; // easy case:  $L(u) = \emptyset$   
 12:  end if  
 13: end if
```



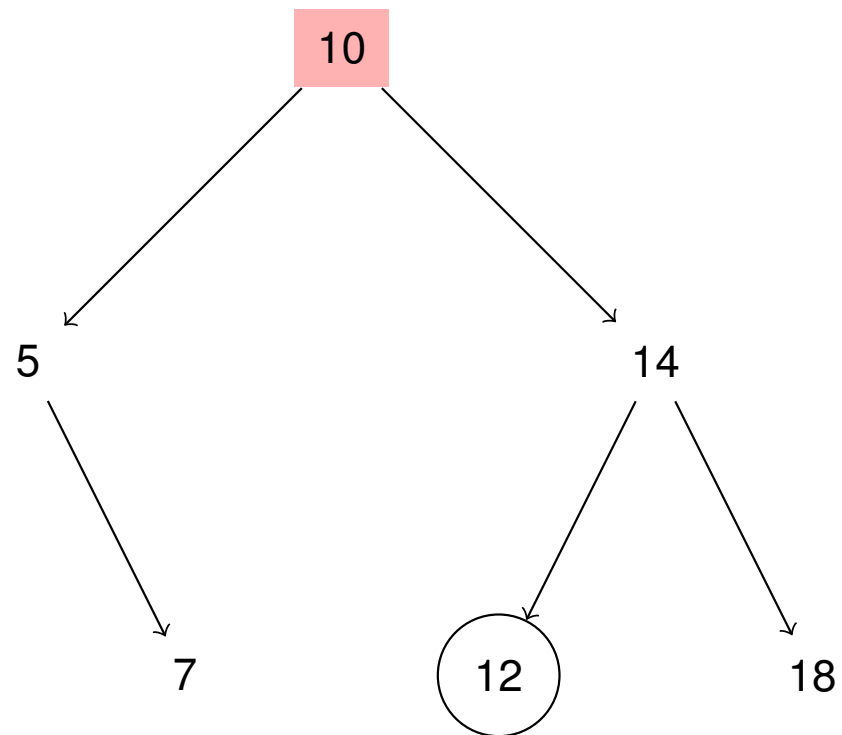
Delete example

`delete(10, r(T))`



Delete example

$\text{delete}(10, r(T))$

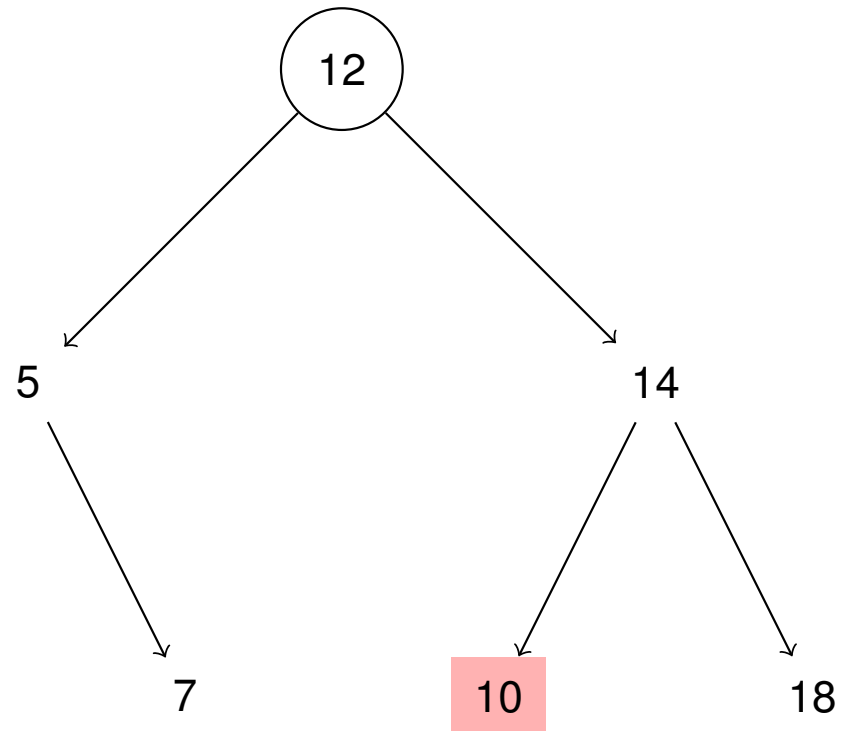


$$u = \min T(14) = 12$$



Delete example

`delete(10, r(T))`

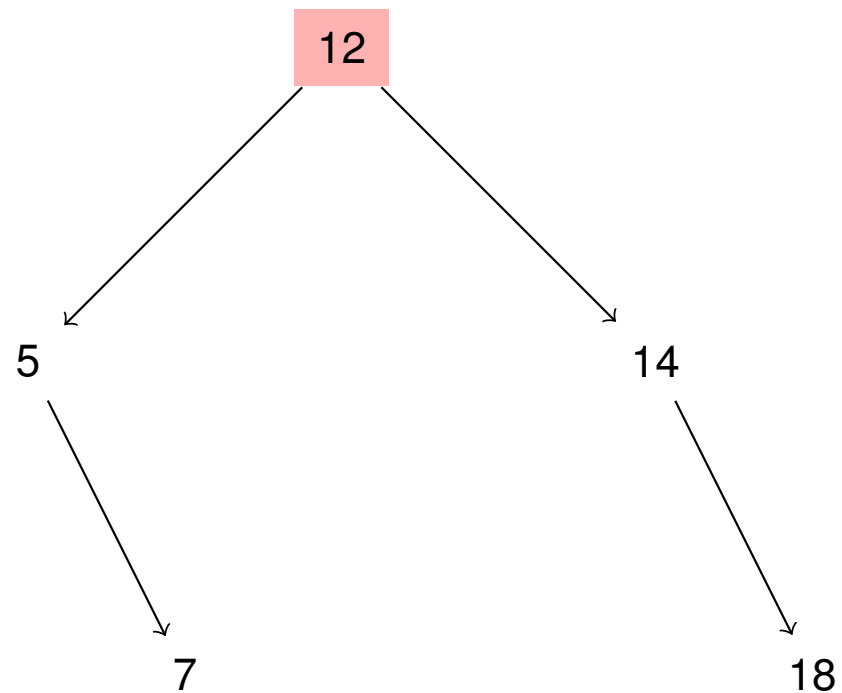


swap values of 10 and 12



Delete example

`delete(10, r(T))`



delete 10

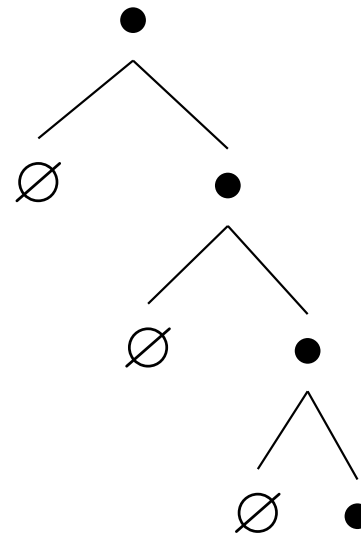
Tree balance

- Balance: $B(T) = D(L(T)) - D(R(T))$
- Tree is **balanced**: $B(T) \in \{-1, 0, 1\}$
- In a balanced tree, $D(T)$ is $O(\log n)$
- **Intuition**: if a BST has $n = 2^k$ nodes at level k , then $k = \log n$
- Intuitively, balance \approx all leaves have same depth
- **Not actually true**, but close enough
- If T is balanced, $D(T) < \log_{\phi}(n + 2) - 1$ with ϕ golden ratio



Complexity

- Every call involves at most one recursion
- \Rightarrow Recurse along one path only, no backtracking
- Worst-case complexity proportional to depth $D(T)$
- Tree balanced: $D(T)$ is $O(\log n)$
- Otherwise: $D(T)$ is $O(n)$



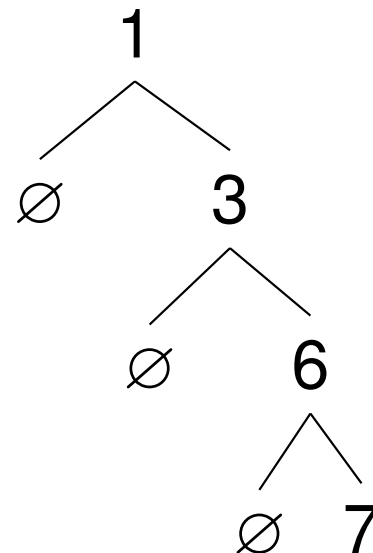


Adelson-Velskii & Landis (AVL) trees



AVL Trees

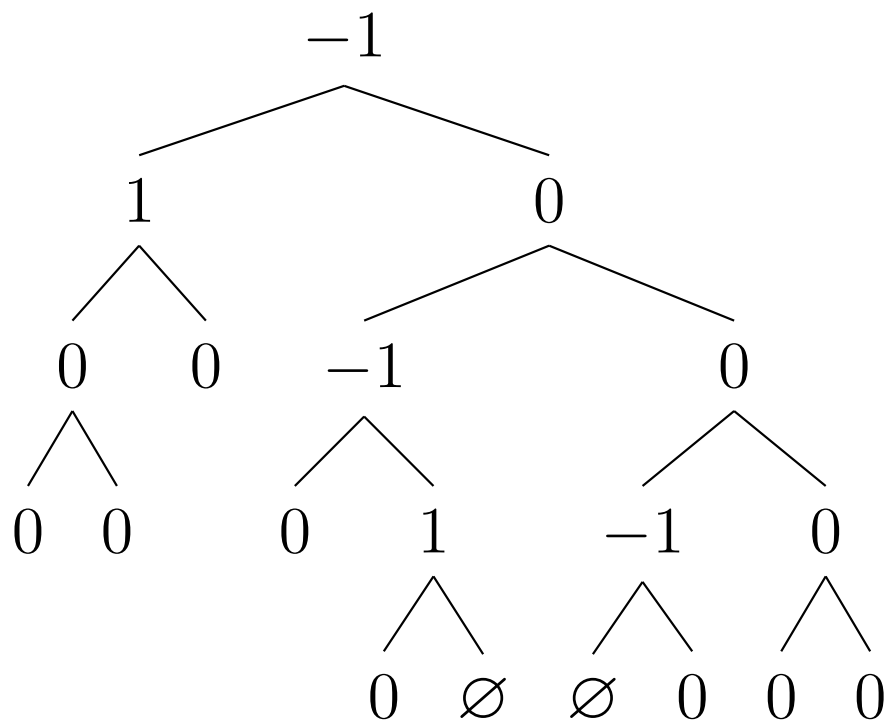
- Try inserting 1, 3, 6, 7 in this order: get **unbalanced tree**



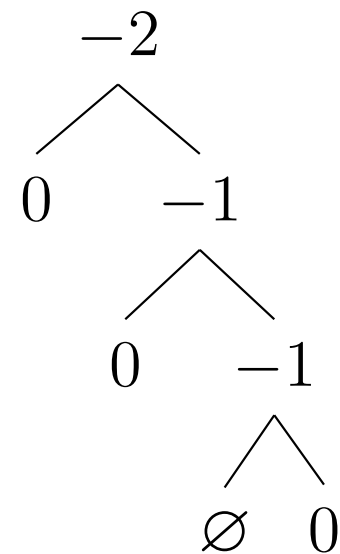
- Worst case find (i.e., find the key 7) is $O(n)$
- Need to *rebalance* the tree to be more efficient
- **AVL trees invariant:** $B(T) \in \{-1, 0, 1\}$

Examples

AVL tree:



Non-AVL tree:



Nodes indicate $B(\text{tree}(v))$



Insertion example

insert 1

1

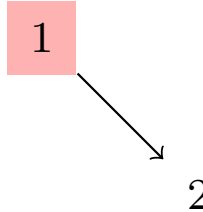
$$v_1 = 1;$$

$$r(T) = v_1;$$



Insertion example

insert 2



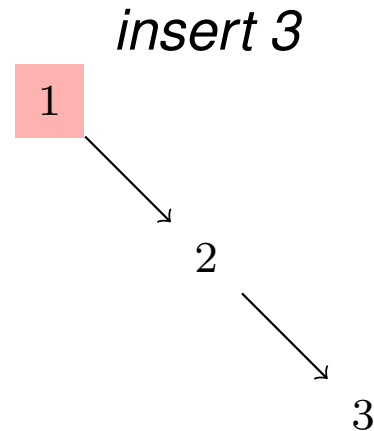
$$v_2 = 2;$$

$$R(v_1) = v_2;$$

$$P(v_2) = v_1;$$



Insertion example



$$v_3 = 3;$$

$$R(v_2) = v_3;$$

$$P(v_3) = v_2;$$

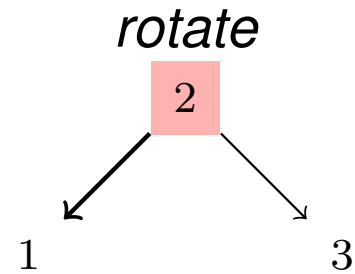
$$D(L(v_1)) = 0,$$

$$D(R(v_1)) = 2:$$

$B(T) = -2$: out of balance



Insertion example



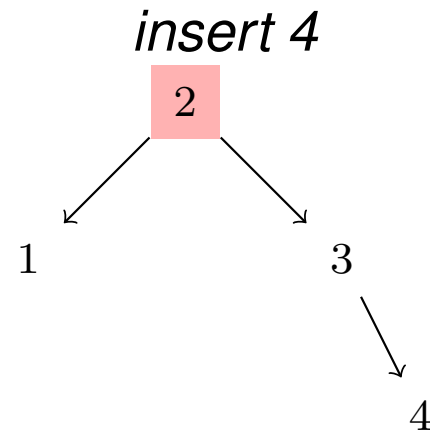
$$r(T) = v_2;$$

$$L(v_2) = v_1;$$

$$P(v_1) = v_2;$$



Insertion example



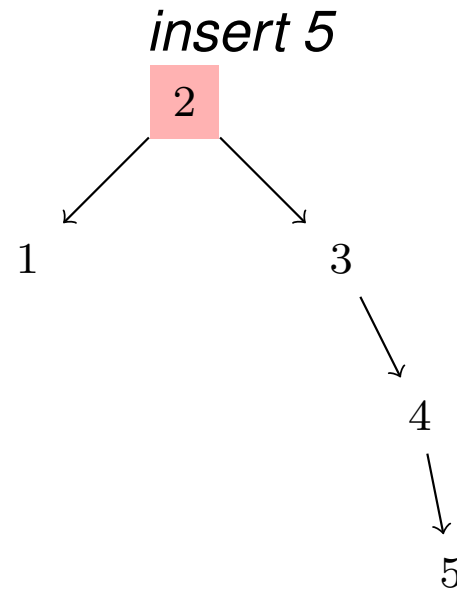
$$v_4 = 4;$$

$$R(v_3) = v_4;$$

$$P(v_4) = v_3;$$



Insertion example



$$v_5 = 5;$$

$$R(v_4) = v_5;$$

$$P(v_5) = v_4;$$

$$H(L(v_3)) = 0,$$

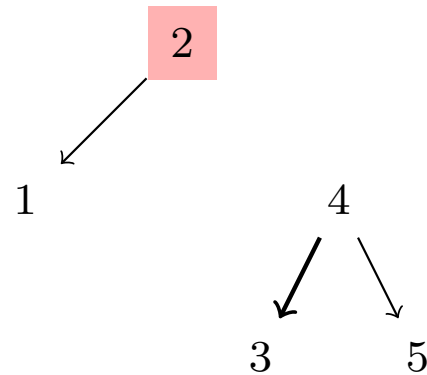
$$H(R(v_3)) = 2:$$

$B(T) = -2$: out of balance



Insertion example

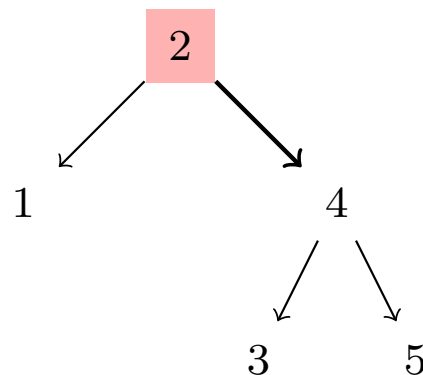
rotate 1/2



$$L(v_4) = v_3;$$
$$P(v_3) = v_4;$$

Insertion example

rotate 2/2

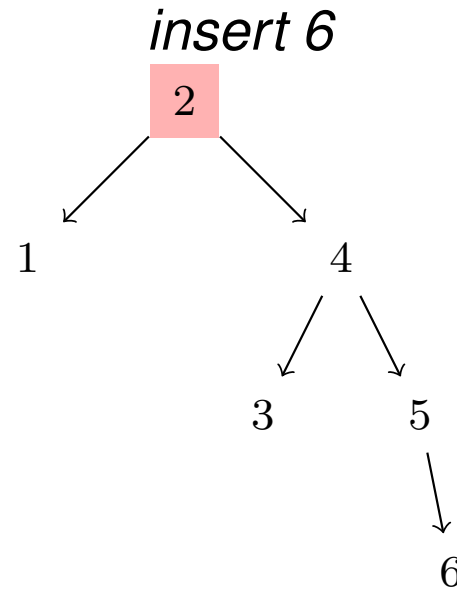


$$R(v_2) = v_4;$$

$$P(v_4) = v_2;$$



Insertion example



$$v_6 = 6;$$

$$R(v_5) = v_6;$$

$$P(v_6) = v_5;$$

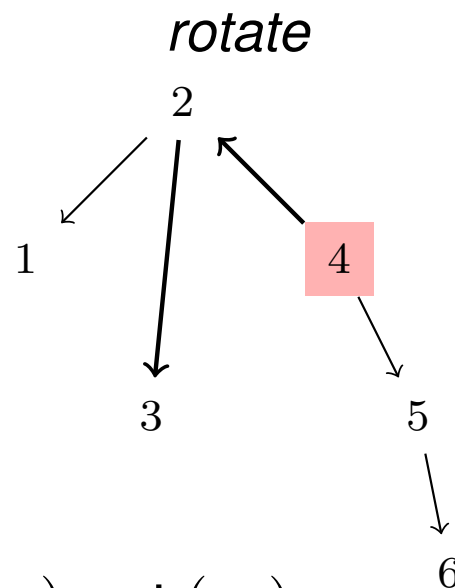
$$H(L(v_2)) = 1,$$

$$H(R(v_2)) = 3:$$

$B(T) = -2$: out of balance



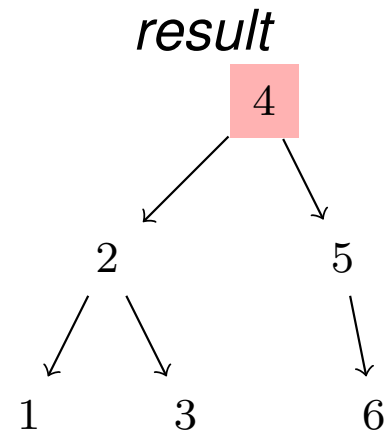
Insertion example



$$\begin{aligned} R(v_2) &= L(v_4); \\ P(L(v_4)) &= v_2; \\ L(v_4) &= v_2; \\ P(v_2) &= v_4; \\ r(T) &= v_4; \end{aligned}$$



Insertion example

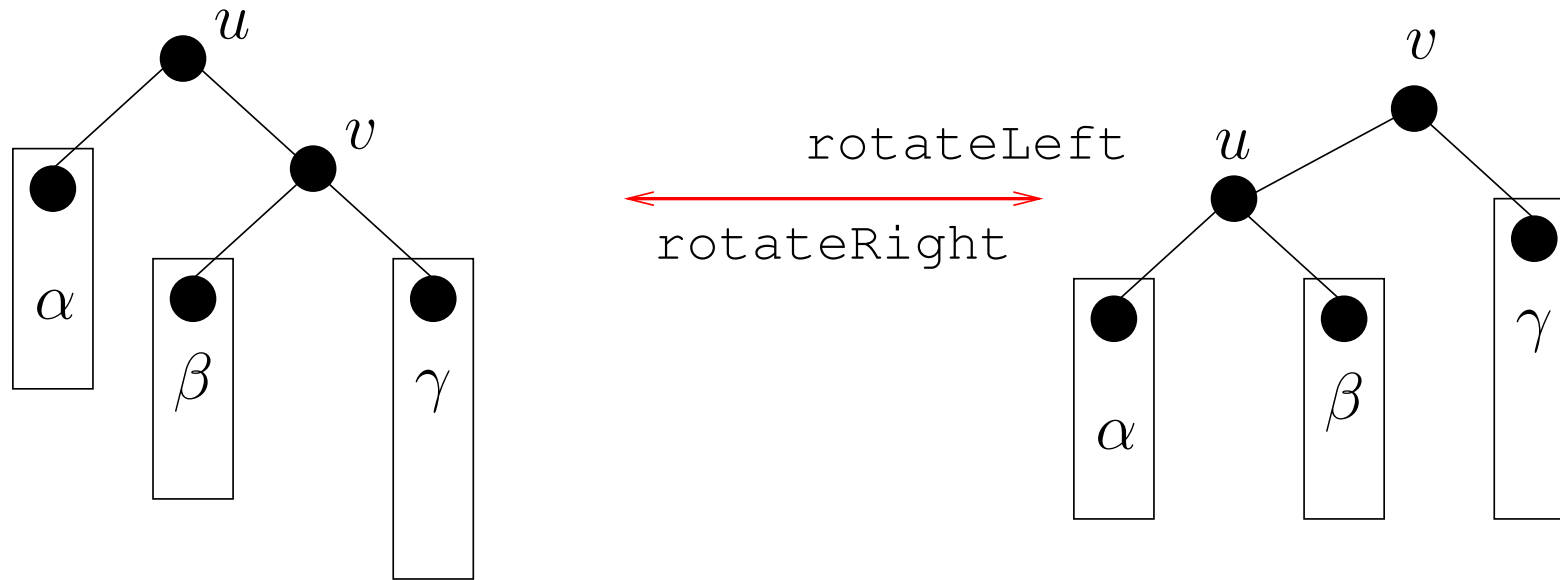




In general

- Decompose balanced trees operations into:
 - the operation itself
 - some *rebalancing operations* called **rotations**
- `min/max, find`: as in BSTs
- Unbalancing can occur on insertion and deletion
- Insert/delete one node at a time \Rightarrow unbalance offset ≤ 1
- I.e., $B(T) \in \{-2, -1, 0, 1, 2\}$
- `insert, delete`: as in BST with rotations

Left and right rotation





Algebraic interpretation

- Let α, β, γ be trees, u, v be nodes not in α, β, γ
- Define:
 - $\text{rotateLeft}(\langle \alpha, u, \langle \beta, v, \gamma \rangle \rangle) = \langle \langle \alpha, u, \beta \rangle, v, \gamma \rangle$
 - $\text{rotateRight}(\langle \langle \alpha, u, \beta \rangle, v, \gamma \rangle) = \langle \alpha, u, \langle \beta, v, \gamma \rangle \rangle$
- A sort of “associativity of trees”
- Remark: $\text{rotateLeft}, \text{rotateRight}$ are inverses

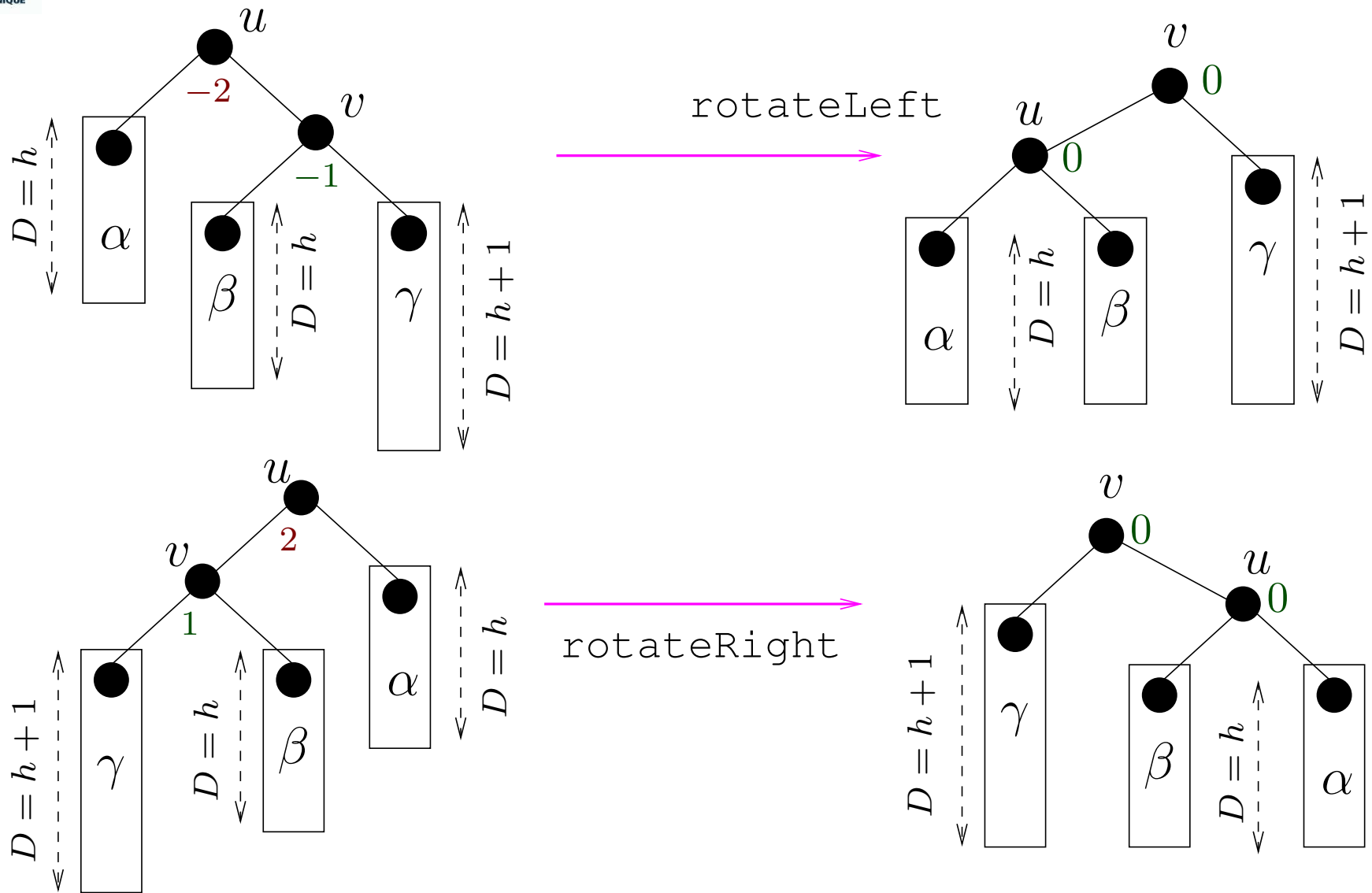
Thm.

$$\begin{aligned} \text{rotateRight}(\text{rotateLeft}(T)) &= \\ \text{rotateLeft}(\text{rotateRight}(T)) &= T \end{aligned}$$

Proof

Directly from the definition

Rotating and rebalancing



Properties of rotation

Thm.

$\forall T, \text{rotateLeft}(T), \text{rotateRight}(T')$ are BSTs

Proof

(Sketch): The tree order only changes locally for u, v . In T , $\text{tree}(v) = R(u) \Rightarrow u < v$. In $\text{rotateLeft}(T)$, $\text{tree}(u) = L(v)$, which is consistent with $u < v$. Similarly for T' .

- Suppose $D(\alpha) = D(\beta) = h$ and $D(\gamma) = h + 1$
- Let $T = \langle \alpha, u, \langle \beta, v, \gamma \rangle \rangle$: then $B(T) = -2$
- Let $T' = \langle \langle \gamma, u, \beta \rangle, v, \alpha \rangle$: then $B(T') = 2$

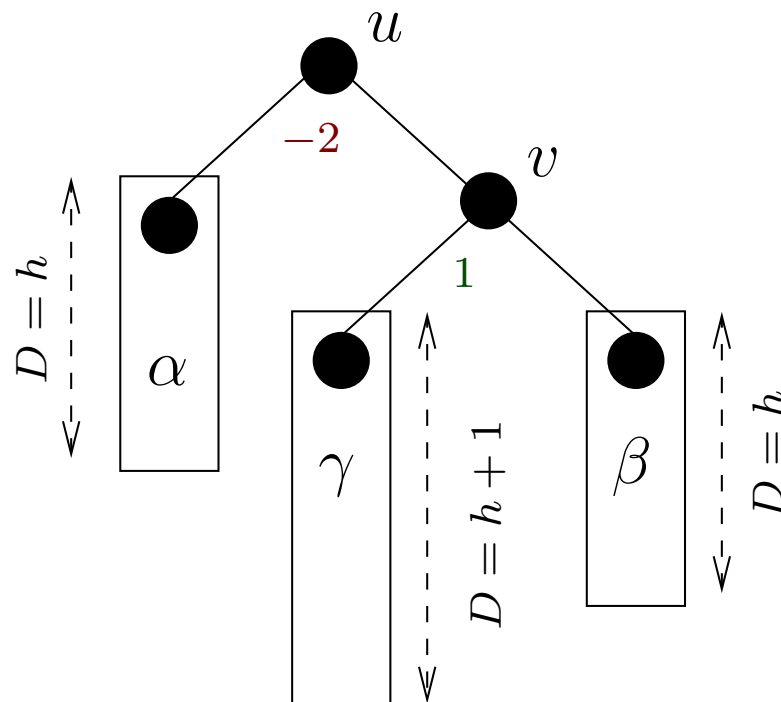
Thm.

T, T' as above $\Rightarrow B(\text{rotateLeft}(T)) = 0, B(\text{rotateRight}(T')) = 0$

Proof

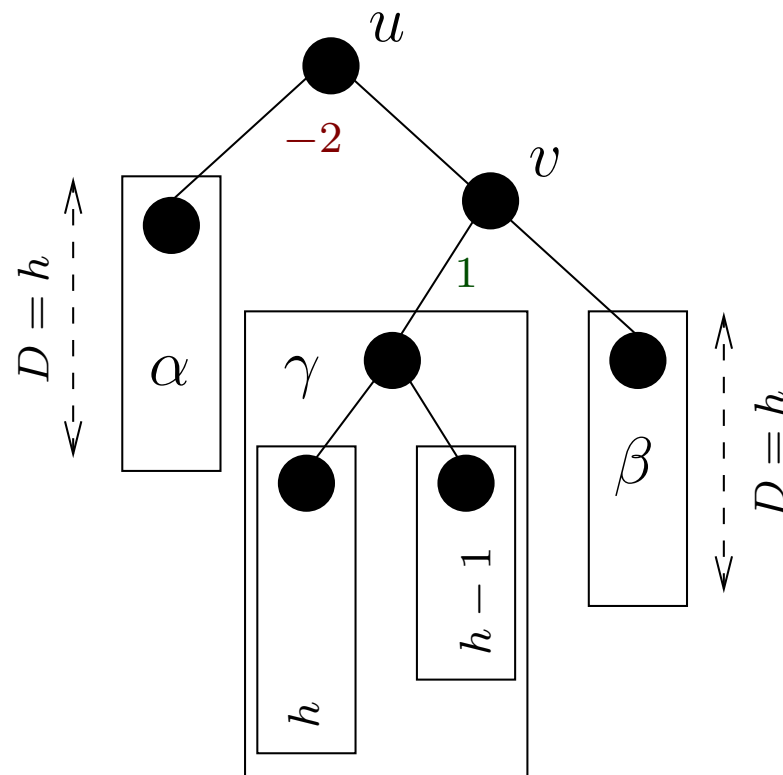
(Sketch): since subtrees α, γ are swapped, tree depth is $D = h$ for all subtrees

Is this enough?



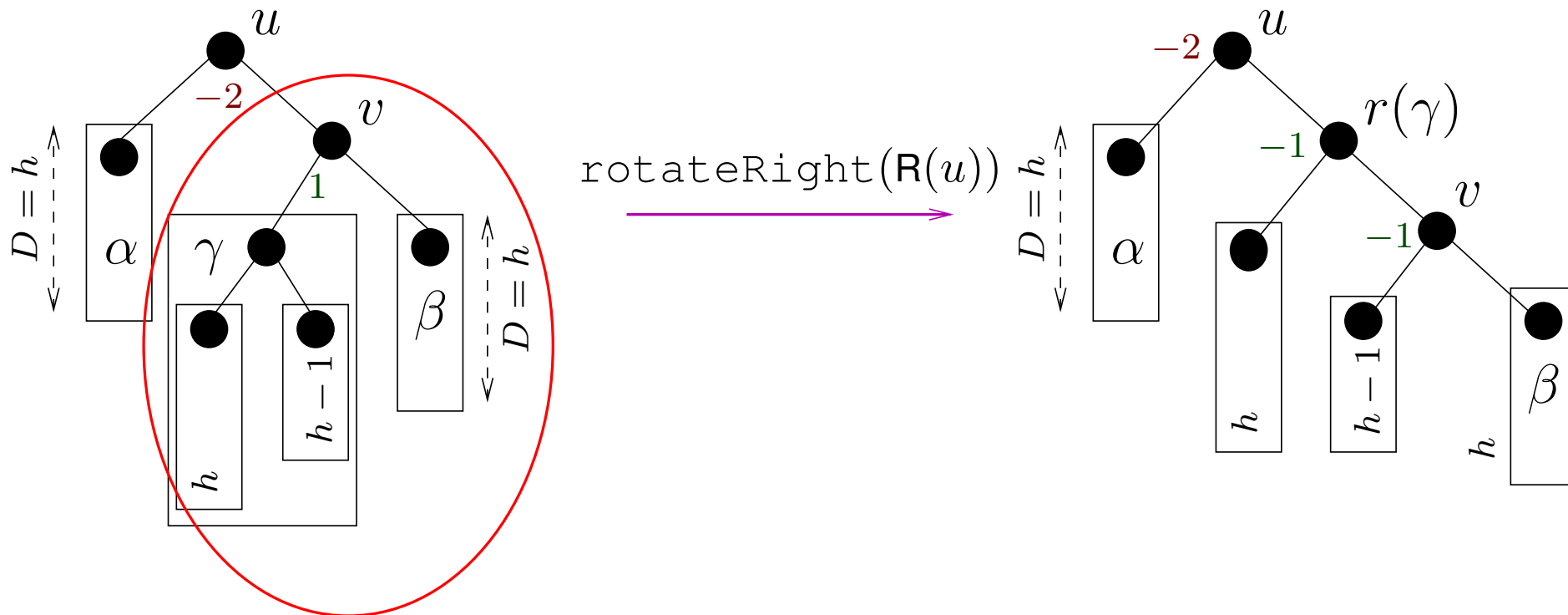
Rotating leaves γ at its place, doesn't work

Break γ up into subtrees



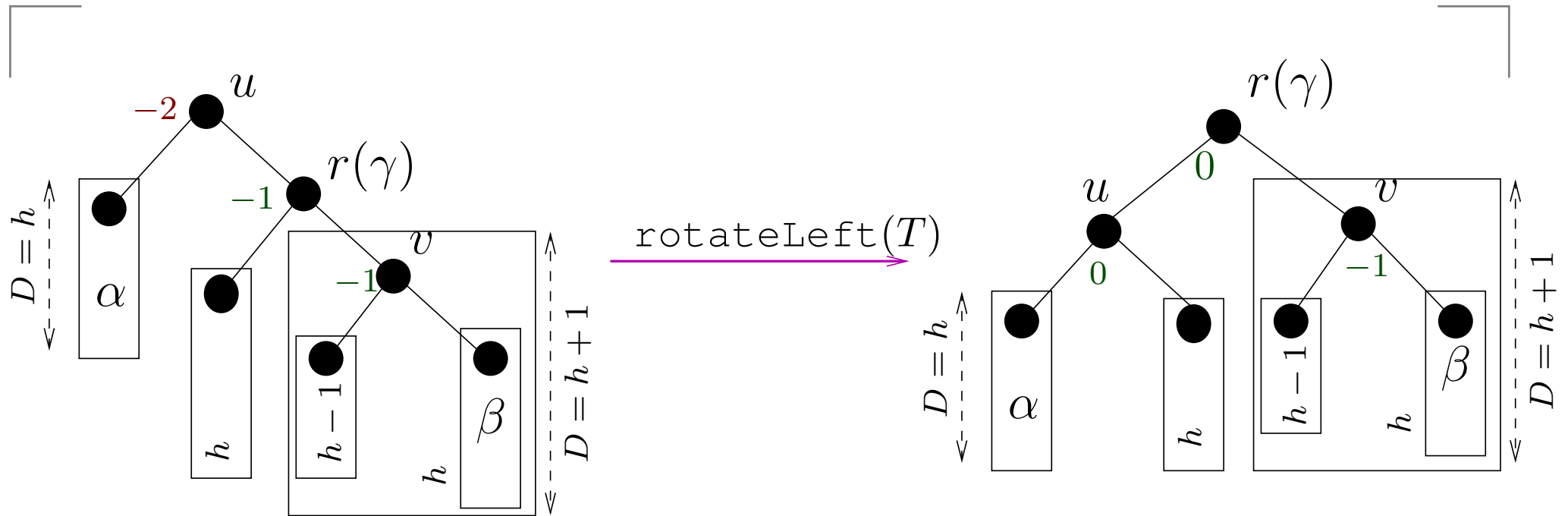
Now we can rotate $\text{tree}(v) = R(u)$

Rotate a subtree right



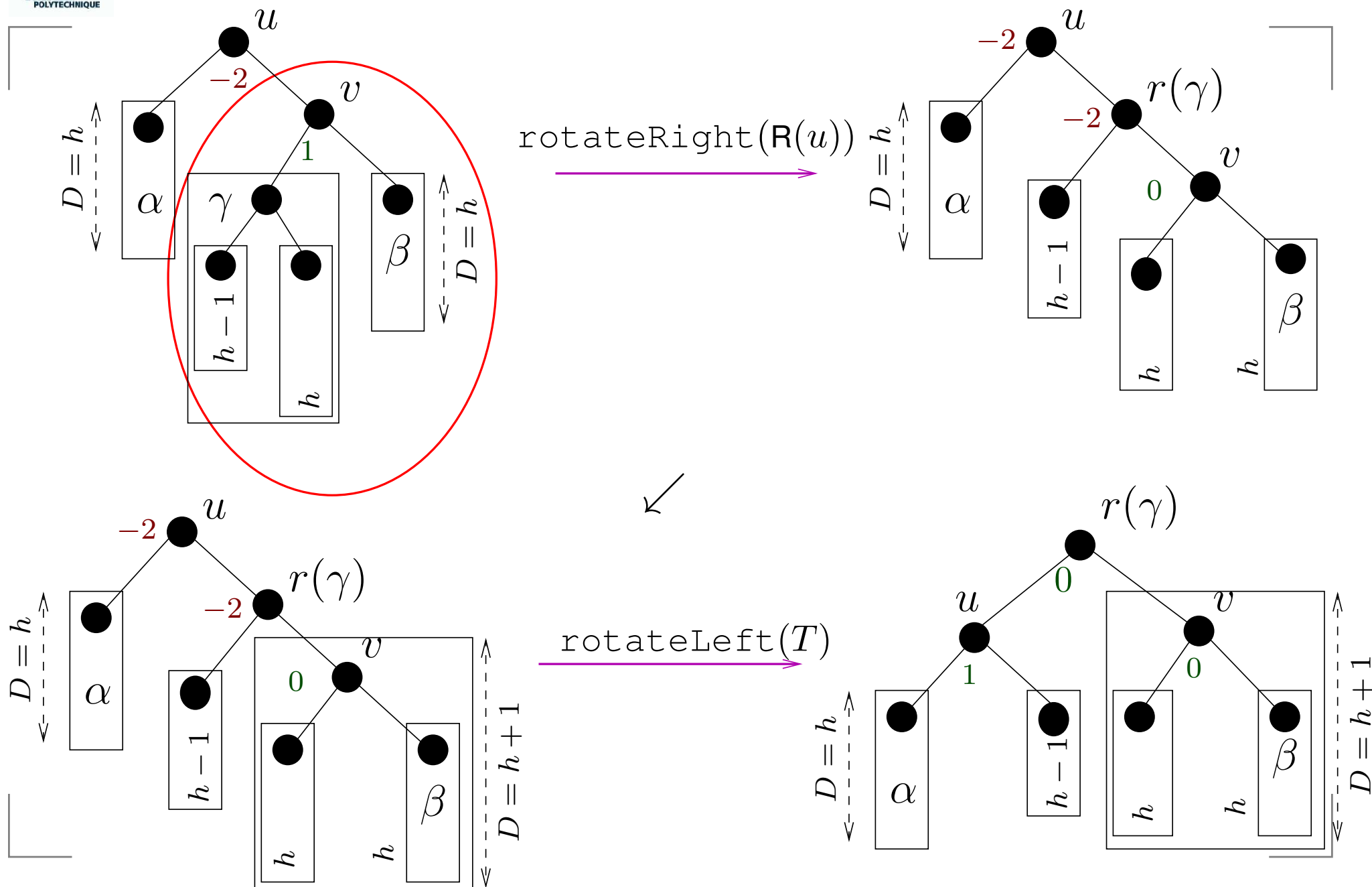
Rotate $R(u)$ right

Finally, rotate left

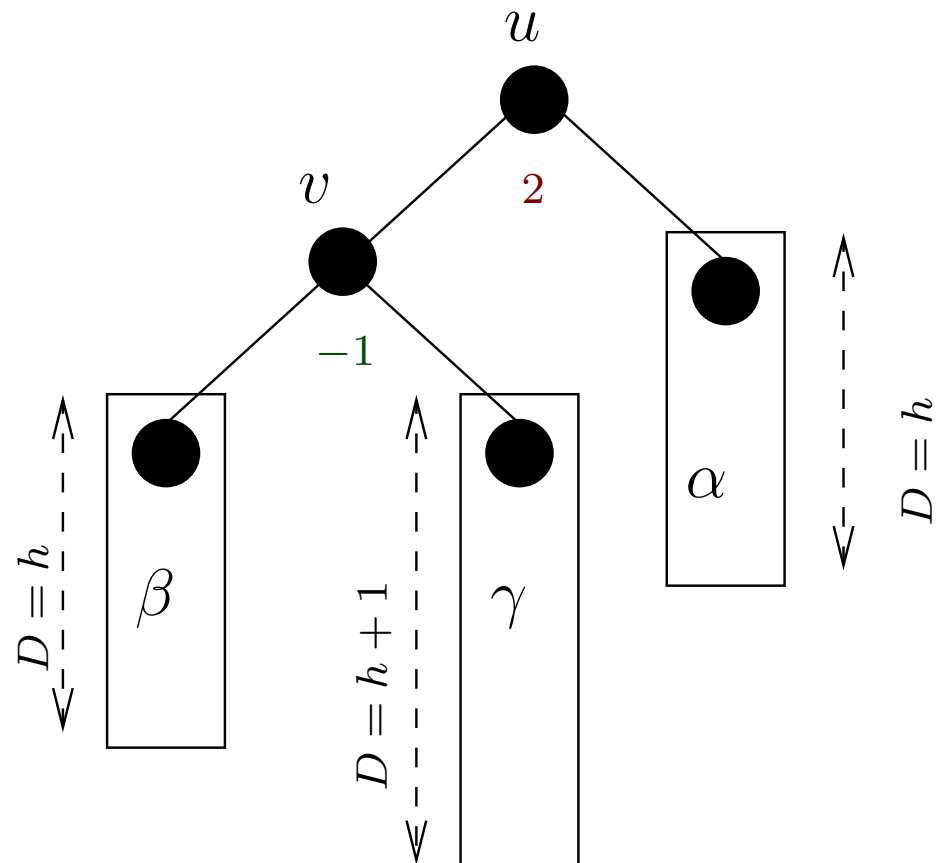


Rotate T left

Symmetric cases I



Symmetric cases II



Rebalance: `rotateLeft(L(u)), rotateRight(T)`



Rotations vs. optimism

- Get rid of rotations, and trust chance: probability that random BST is balanced?
- Given a sequence $\sigma \in \{1, \dots, n\}^n$, we insert it in a BST T
- Assume $|L(T)| = K$ and $|R(T)| = n - 1 - K$
- Assume uniform distribution on K i.e. $\forall k \leq n P(K = k) = \frac{1}{n}$

σ	(1,2,3)	(1,3,2)	(2,1,3)	(2,3,1)	(3,1,2)	(3,2,1)
T						
type	A	B	C	C	D	E

Type C (balanced) twice as likely as any other type!



The average BST balance

- Average depth for BSTs: $O(\log n)$ [Devroye, 1986]
- Average path length for BSTs: $O(n \log n)$ [Vitter & Flajolet, 1990]
- BSTs are well balanced even without rotations!



Heaps and priority queues



Queues reminder

- Queue operations:
 - `pushBack(v)`: insert v at the end
 - `popFront()`: return and remove element at the beginning
- Used in BFS (compute paths with fewest arcs, see Lecture 2)
- If arcs are prioritized (e.g. travelling times for route segments): want queue to return *element with highest priority*

This may not be at the beginning of the queue



Priority queues

- V : a set; $(S, <)$: a totally ordered set
- **Priority queue** on V, S : set Q of pairs (v, p_v) s.t. $v \in V$ and $p_v \in S$
- Usually, p_v is a number
- E.g., if p_v is the rank of entrance of v in Q , then Q is a standard queue
- Supports three main operations:
 - `insert(v, p_v)`: inserts v in Q with priority p_v
 - `max()`: returns the element of Q with maximum priority
 - `popMax()`: returns and removes `max()`
- Implemented as **heaps**



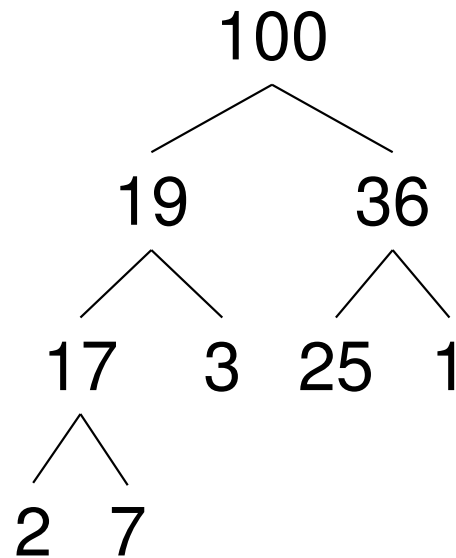
Heap

- A (binary) **heap** is an abstract, tree-like data structure which offers:
 - $O(\log |Q|)$ insert
 - $O(1)$ max
 - $O(\log |Q|)$ popMax
- max in $O(1)$: store max. priority element at BST root
- Invariants:
 - *shape property*: all levels except perhaps the last are fully filled; the last level is filled left-to-right
 - *heap property*: every node stores an element of higher priority than its subnodes



Example

Let $V = \mathbb{N}$, and for all $v \in V$ we let $p_v = v$



A balanced tree

Thm.

If Q is a binary heap, $B(Q) \in \{0, 1\}$

Proof

Follows trivially from the shape property. Since all levels are filled completely apart perhaps from the last, $B(Q) \in \{-1, 0, 1\}$. Since the last is filled left-to-right, $B(Q) \neq -1$

Cor.

A binary heap is a balanced binary tree

Warning: *NOT a BST/AVL*: heap property not compatible with BST invariant $L(v) \leq VR(v)$

Keep the heap balanced: need $O(\log |Q|)$ work to insert/remove



Insert

- Add new element (v, p_v) at the bottom of the heap (last level, leftmost free “slot”)
- Compare with its (unique) parent (u, p_u) ; if $p_u < p_v$, swap u and v 's positions in the heap
- Repeat comparison/swap until heap property is attained

Example: insert (1, 4, 2, 3, 5)

∅



Insert

- Add new element (v, p_v) at the bottom of the heap (last level, leftmost free “slot”)
- Compare with its (unique) parent (u, p_u) ; if $p_u < p_v$, swap u and v 's positions in the heap
- Repeat comparison/swap until heap property is attained

Example: insert (1, 4, 2, 3, 5)

insert 1

1

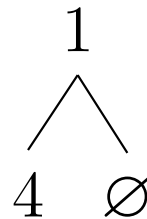


Insert

- Add new element (v, p_v) at the bottom of the heap (last level, leftmost free “slot”)
- Compare with its (unique) parent (u, p_u) ; if $p_u < p_v$, swap u and v 's positions in the heap
- Repeat comparison/swap until heap property is attained

Example: insert (1, 4, 2, 3, 5)

insert 4



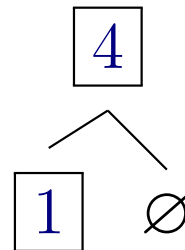


Insert

- Add new element (v, p_v) at the bottom of the heap (last level, leftmost free “slot”)
- Compare with its (unique) parent (u, p_u) ; if $p_u < p_v$, swap u and v 's positions in the heap
- Repeat comparison/swap until heap property is attained

Example: insert (1, 4, 2, 3, 5)

$1 < 4$, *swap*



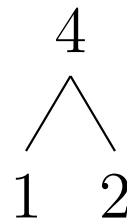


Insert

- Add new element (v, p_v) at the bottom of the heap (last level, leftmost free “slot”)
- Compare with its (unique) parent (u, p_u) ; if $p_u < p_v$, swap u and v 's positions in the heap
- Repeat comparison/swap until heap property is attained

Example: insert (1, 4, 2, 3, 5)

insert 2

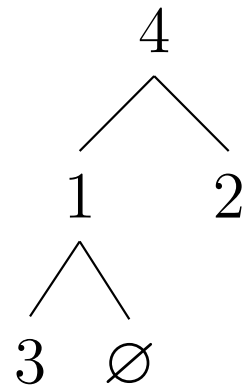


Insert

- Add new element (v, p_v) at the bottom of the heap (last level, leftmost free “slot”)
- Compare with its (unique) parent (u, p_u) ; if $p_u < p_v$, swap u and v 's positions in the heap
- Repeat comparison/swap until heap property is attained

Example: insert (1, 4, 2, 3, 5)

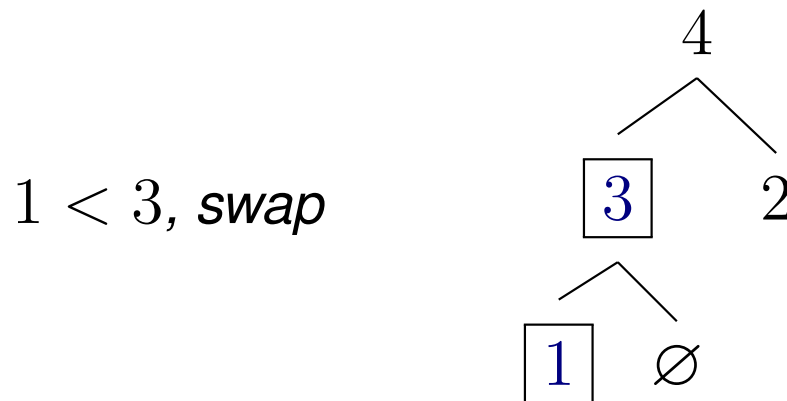
insert 3



Insert

- Add new element (v, p_v) at the bottom of the heap (last level, leftmost free “slot”)
- Compare with its (unique) parent (u, p_u) ; if $p_u < p_v$, swap u and v 's positions in the heap
- Repeat comparison/swap until heap property is attained

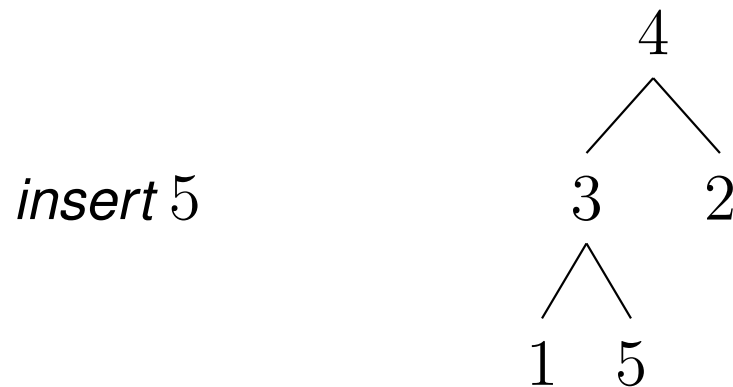
Example: insert (1, 4, 2, 3, 5)



Insert

- Add new element (v, p_v) at the bottom of the heap (last level, leftmost free “slot”)
- Compare with its (unique) parent (u, p_u) ; if $p_u < p_v$, swap u and v 's positions in the heap
- Repeat comparison/swap until heap property is attained

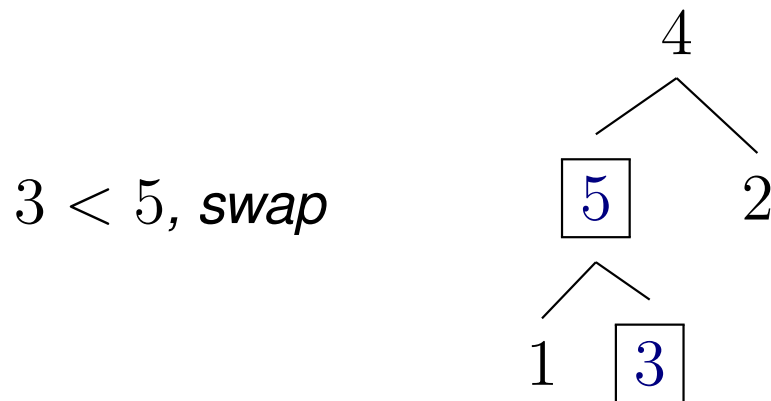
Example: insert (1, 4, 2, 3, 5)



Insert

- Add new element (v, p_v) at the bottom of the heap (last level, leftmost free “slot”)
- Compare with its (unique) parent (u, p_u) ; if $p_u < p_v$, swap u and v 's positions in the heap
- Repeat comparison/swap until heap property is attained

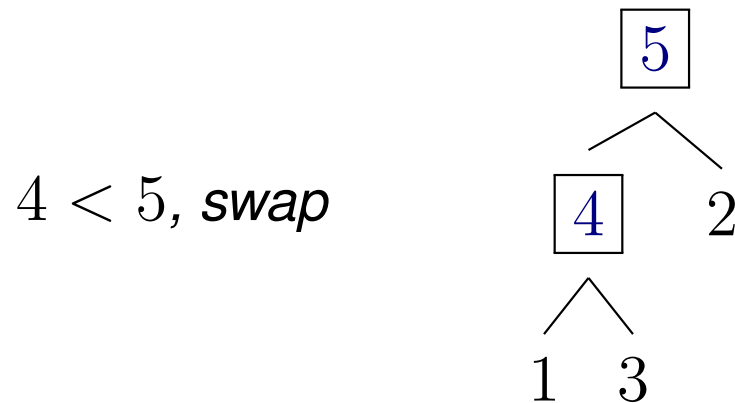
Example: insert (1, 4, 2, 3, 5)



Insert

- Add new element (v, p_v) at the bottom of the heap (last level, leftmost free “slot”)
- Compare with its (unique) parent (u, p_u) ; if $p_u < p_v$, swap u and v 's positions in the heap
- Repeat comparison/swap until heap property is attained

Example: insert (1, 4, 2, 3, 5)





Insertion maintains the heap

- Worst case: `insert` takes time proportional to tree depth: $O(\log n)$
- The shape property is maintained:
 - when adding a new element at last level in leftmost free slot
 - when swapping node values along a path to the root
- The heap property is not maintained after adding a new element
- However, it is restored after the sequence of swaps

Thm.

The insertion operation maintains the heap



Max

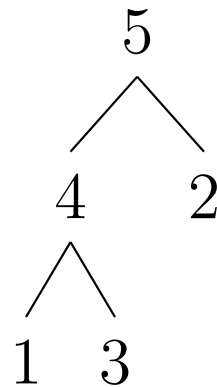
- *Easy*: return the root of the heap tree
- Evidently $O(1)$



Removal of max

- Let $\text{last}(Q)$ be the rightmost non-empty element of the last heap level
- Move node $\text{last}(Q)$ to the root $r(Q)$
- Compare v with its children u, w : if $p_v \geq p_u, p_v \geq p_w$, heap is in correct order
- Otherwise, swap v with $\max_p(u, v)$ (use \min_p if min-heap) and repeat comparison/swap until termination

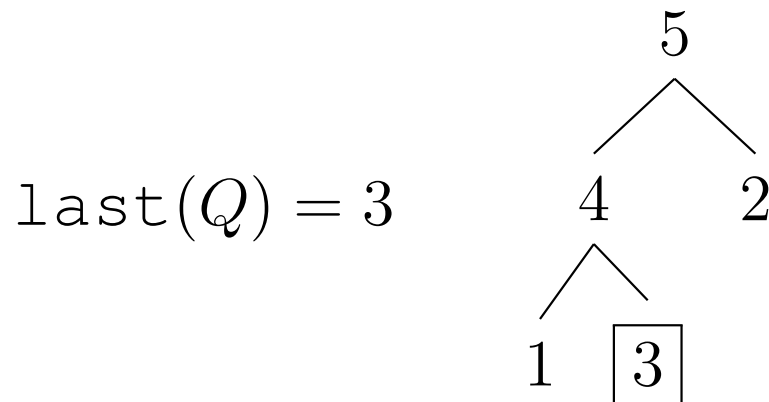
original tree





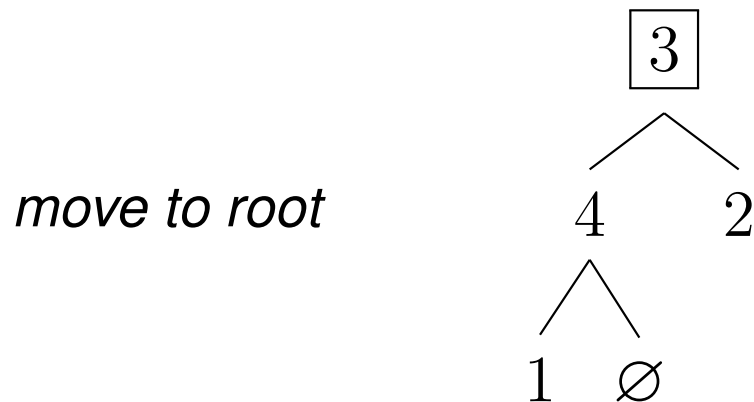
Removal of max

- Let $\text{last}(Q)$ be the rightmost non-empty element of the last heap level
- Move node $\text{last}(Q)$ to the root $r(Q)$
- Compare v with its children u, w : if $p_v \geq p_u, p_v \geq p_w$, heap is in correct order
- Otherwise, swap v with $\max_p(u, v)$ (use \min_p if min-heap) and repeat comparison/swap until termination



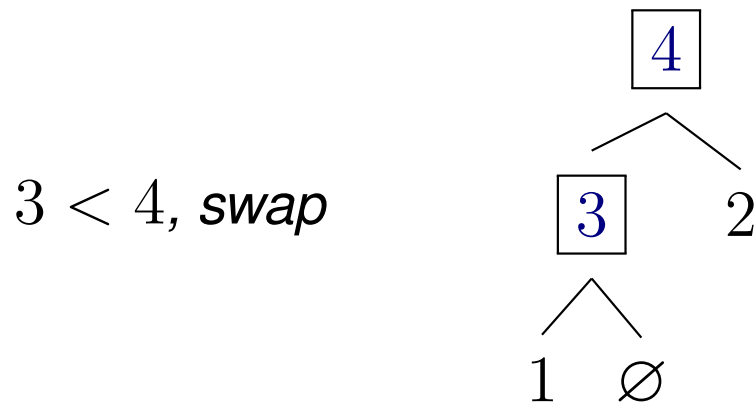
Removal of max

- Let $\text{last}(Q)$ be the rightmost non-empty element of the last heap level
- Move node $\text{last}(Q)$ to the root $r(Q)$
- Compare v with its children u, w : if $p_v \geq p_u, p_v \geq p_w$, heap is in correct order
- Otherwise, swap v with $\max_p(u, v)$ (use \min_p if min-heap) and repeat comparison/swap until termination



Removal of max

- Let $\text{last}(Q)$ be the rightmost non-empty element of the last heap level
- Move node $\text{last}(Q)$ to the root $r(Q)$
- Compare v with its children u, w : if $p_v \geq p_u, p_v \geq p_w$, heap is in correct order
- Otherwise, swap v with $\max_p(u, v)$ (use \min_p if min-heap) and repeat comparison/swap until termination





Efficient construction

- Insert n elements of V in an empty heap
- Trivially: each insert takes $O(\log n)$, get $O(n \log n)$ to construct the whole heap
- Instead:
 1. arbitrarily put the element in a binary tree with the shape property (can do this in $O(n)$)
 2. lower level first, move nodes down using the same swapping procedure as for `popMax`
- At level ℓ , moving a node down costs $O(\ell)$ (worst-case)
- There's $\leq \lceil \frac{n}{2^{\ell+1}} \rceil$ nodes at level ℓ and $O(\log n)$ possible levels

$$\sum_{\ell=0}^{\lceil \log n \rceil} \frac{n}{2^{\ell+1}} O(\ell) = O\left(n \sum_{\ell=0}^{\lceil \log n \rceil} \frac{1}{2^{\ell}}\right) \leq O\left(n \sum_{\ell=0}^{\infty} \frac{1}{2^{\ell}}\right) = O(2n) = O(n)$$

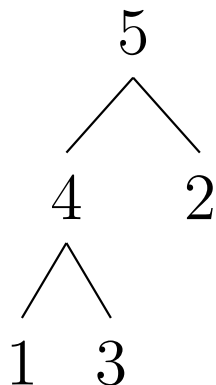
Implementation



- A priority queue is implemented as a heap
- A heap can be implemented as a tree
- But it needn't be!



Binary trees in arrays



<i>Node</i>	5	4	2	1	3
<i>Index</i>	0	1	2	3	4
		i	$2i + 1$	$2i + 2$	

- Heap Q of n elements stored in an array q of length n
- $q_0 = r(Q)$
- **Subnodes**
If $q_i = v$, then $q_{2i+1} = L(v)$ and $q_{2i+2} = R(v)$ (whenever $2i + 1, 2i + 2 < n$)
- **Parent**
If $v \neq q_0$, $P(v) = q_j$, where $j = \lfloor \frac{i-1}{2} \rfloor$

We now have all the elements: start implementing!



k -ary Search Trees



Tries

search

look for a key

use a total order

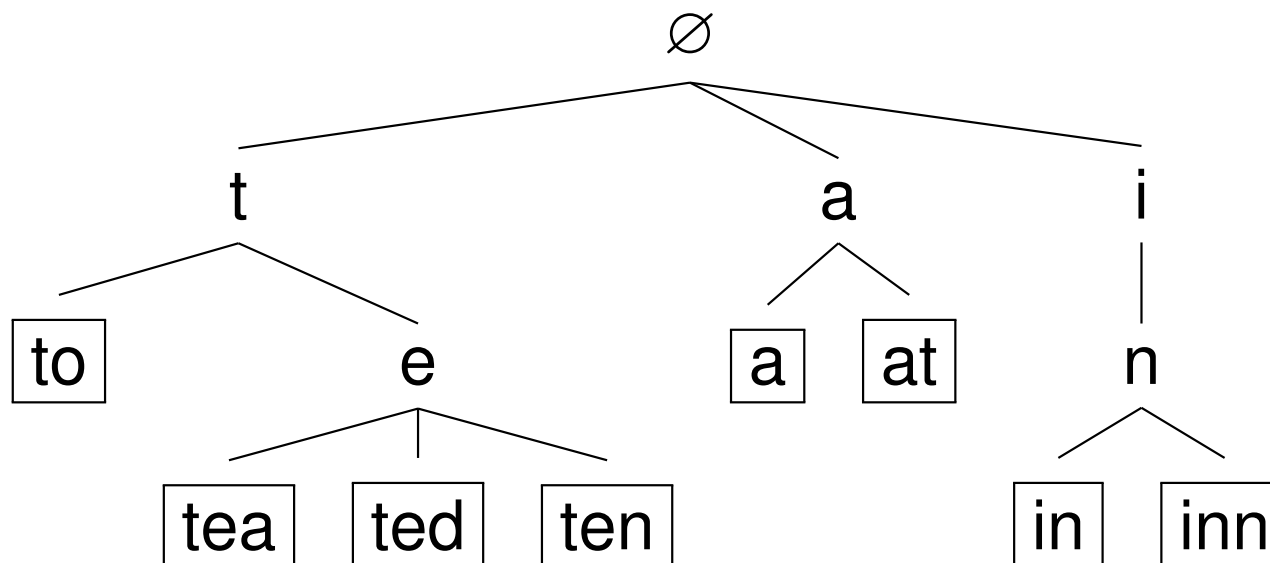
hashing

use key to find its position

each key defines a path to a leaf

Trie example

$V = \{a, at, to, tea, ted, ten, in, inn\}$



- Each key is stored at a leaf node ℓ
- Each non-leaf node v contains a prefix of all keys stored in the tree rooted at v
- The trie root node is \emptyset , the empty string



Trie properties

- Path on trie corresponding to key k : given by key itself

Compare with hash functions: hash value specified by key

- If max length key is m , path length $O(m)$
- `find`, `insert` and `delete` take worst-case $O(m)$
- If m constant w.r.t. $n = |V|$, then methods are $O(1)$
- *Comparison to hash functions:*
 - With respect to hashing, tries support “ordered iteration”
 - Hash tables need re-hashing (expensive) as they become full; tries adjust to size gracefully
 - No need to construct good hash functions

Warning: there are several trie variants



End of Lecture 7