

# Data Structures

## A Supplement

(Based on [Manber 1989])

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# Heaps

- 🌐 A (max binary) heap is a **complete binary tree** whose keys satisfy the heap property:  
*the key of every node is greater than or equal to the key of any of its children.*
- 🌐 It supports the two basic operations of a **priority queue**:

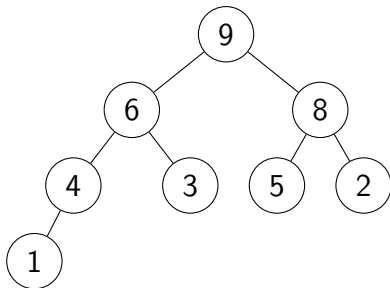
# Heaps

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*the key of every node is greater than or equal to the key of any of its children.*
- 🌐 It supports the two basic operations of a **priority queue**:
  - ☀ *Insert*( $x$ ): insert the key  $x$  into the heap.
  - ☀ *Remove*() : remove and return the largest key from the heap.

## Heaps (cont.)

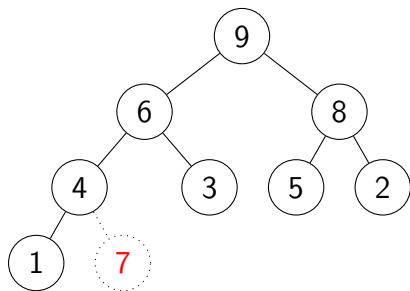
A complete binary tree can be represented implicitly by an array  $A$  as follows:

1. The root is stored in  $A[1]$ .
2. The left child of  $A[i]$  is stored in  $A[2i]$  and the right child is stored in  $A[2i + 1]$ .

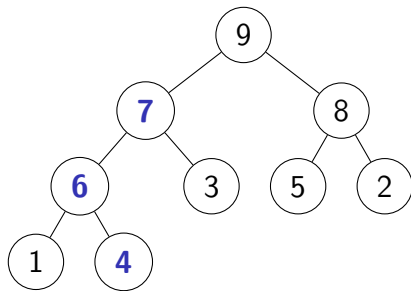


1	2	3	4	5	6	7	8	9	10	11	12
9	6	8	4	3	5	2	1	-	-	-	-

# Heaps (cont.)



Before *Insert*(7)



After *Insert*(7)

## Heaps (cont.)

**Algorithm Insert\_to\_Heap** ( $A, n, x$ );  
**begin**

$n := n + 1$ ;

$A[n] := x$ ;

$child := n$ ;

$parent := n \text{ div } 2$ ;

**while**  $parent \geq 1$  **do**

**if**  $A[parent] < A[child]$  **then**

$swap(A[parent], A[child])$ ;

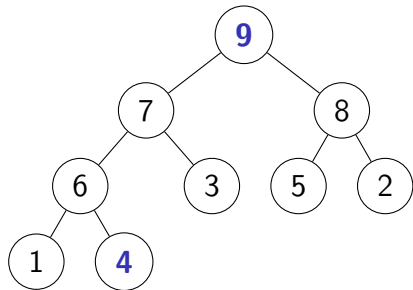
$child := parent$ ;

$parent := parent \text{ div } 2$

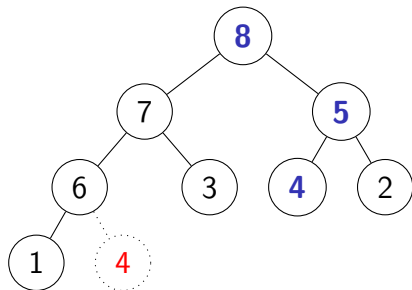
**else**  $parent := 0$

**end**

# Heaps (cont.)



Before *Remove()*



After *Remove()*

# Heaps (cont.)

```
Algorithm Remove_Max_from_Heap ( $A, n$ );  
begin  
  if  $n = 0$  then print “the heap is empty”  
  else  $Top\_of\_the\_Heap := A[1]$ ;  
     $A[1] := A[n]$ ;  $n := n - 1$ ;  
     $parent := 1$ ;  $child := 2$ ;  
    while  $child \leq n - 1$  do  
      if  $A[child] < A[child + 1]$  then  
         $child := child + 1$ ;  
      if  $A[child] > A[parent]$  then  
         $swap(A[parent], A[child])$ ;  
         $parent := child$ ;  
         $child := 2 * child$   
      else  $child := n$   
end
```

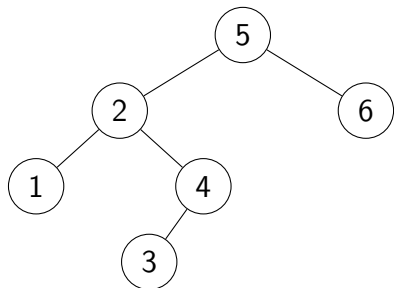


## Definition

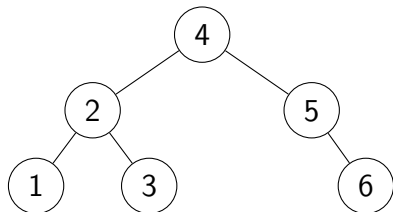
An AVL tree is a binary search tree such that, for every node, the **difference between the heights** of its left and right subtrees is **at most 1** (the height of an empty tree is defined as 0).

This definition guarantees a maximal height of  $O(\log n)$  for any AVL tree of  $n$  nodes.

# AVL Trees (cont.)



A binary search tree  
but NOT an AVL tree



A binary search tree  
and also an AVL tree

## AVL Trees (cont.)

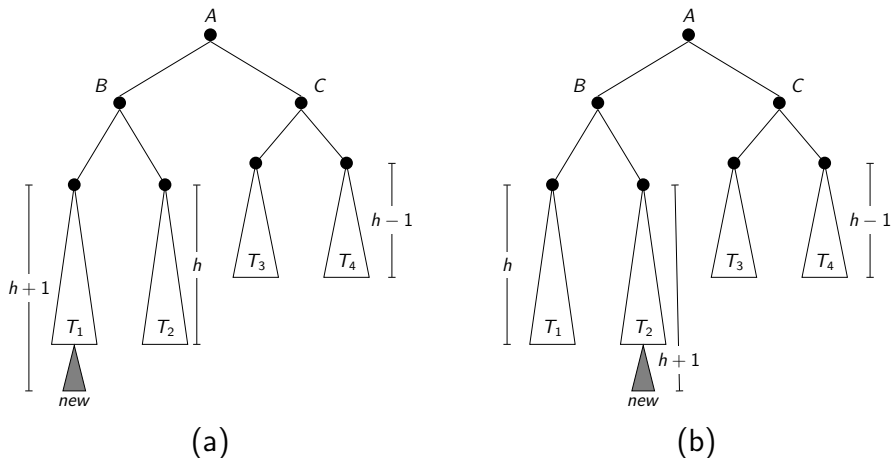
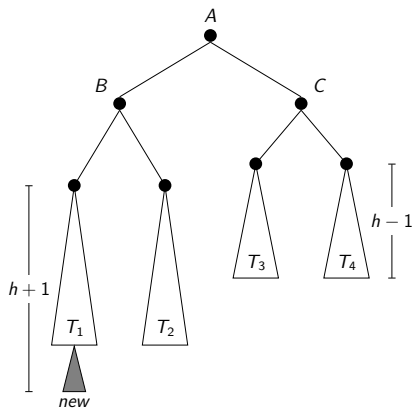


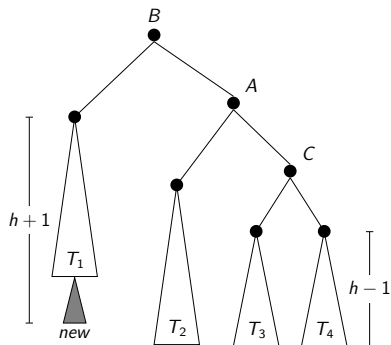
Figure: Insertions that invalidate the AVL property. Note that this tree rooted at  $A$  shown here may be part of a larger AVL tree.

Source: redrawn from [Manber 1989, Figure 4.13].

## AVL Trees (cont.)



(a)



(b)

Figure: A single rotation: (a) before; (b) after.

Source: redrawn from [Manber 1989, Figure 4.14].



# Union-Find

- There are  $n$  elements  $x_1, x_2, \dots, x_n$  divided into groups. Initially, each element is in a group by itself.
- Two operations on the elements and groups:
  - $\text{find}(A)$ : returns the name of  $A$ 's group.
  - $\text{union}(A, B)$ : combines  $A$ 's and  $B$ 's groups to form a new group with a unique name.
- To tell if two elements are in the same group, one may issue a find operation for each element and see if the returned names are the same.

# Union-Find (cont.)

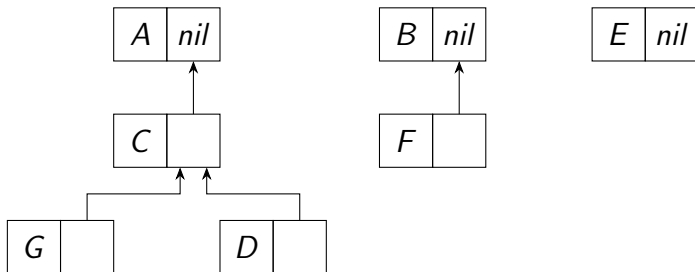


Figure: The representation for the union-find problem.

Source: redrawn from [Manber 1989, Figure 4.16].

# Balancing

- 🌐 The root also stores the number of elements in (i.e., the size of) its group.
- 🌐 To *balance* the tree resulted from a union operation, *let the smaller group join the larger group* and update the size of the larger group accordingly.

## Theorem (Theorem 4.2)

*If balancing is used, then any tree of height  $h$  ( $\geq 0$ ) must contain at least  $2^h$  elements.*

- 🌐 Any sequence of  $m$  find or union operations (where  $m \geq n$ ) takes  $O(m \log n)$  steps.



# Union-Find (cont.)

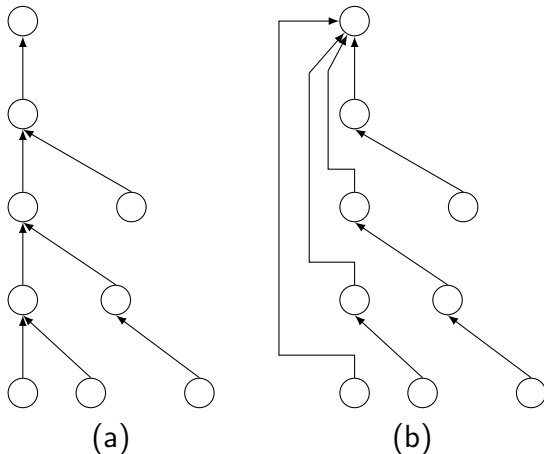


Figure: Path compression: (a) before; (b) after.

Source: redrawn from [Manber 1989, Figure 4.17].

# Effect of Path Compression

## Theorem (Theorem 4.3)

*If both balancing and path compression are used, any sequence of  $m$  find or union operations (where  $m \geq n$ ) takes  $O(m \log^* n)$  steps.*

The value of  $\log^* n$  intuitively equals the number of times that one has to apply  $\log$  to  $n$  to bring its value down to 1.

# Code for Union-Find

```
Algorithm Union_Find_Init(A,n);  
begin  
  for i := 1 to n do  
    A[i].parent := nil;  
    A[i].size := 1  
  end  
end
```

```
Algorithm Find(a);  
begin  
  if A[a].parent <> nil then  
    A[a].parent := Find(A[a].parent);  
    Find := A[a].parent;  
  else  
    Find := a  
  end  
end
```

# Code for Union-Find (cont.)

```
Algorithm Union(a,b);
begin
  x := Find(a);
  y := Find(b);
  if x <> y then
    if A[x].size > A[y].size then
      A[y].parent := x;
      A[x].size := A[x].size + A[y].size;
    else
      A[x].parent := y;
      A[y].size := A[y].size + A[x].size
    end
  end
end
```