# Algorithms 2022: Data Structures

A Supplement (Based on [Manber 1989])

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### September 30, 2022

## 1 Heaps

### 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:
  - 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].

### Heaps (cont.)

 $\mathbf{end}$ 

Heaps (cont.)

```
Algorithm Insert_to_Heap (A, n, x);
begin
n := n + 1;
A[n] := x;
child := n;
parent := n \ div \ 2;
while parent \ge 1 \ do
if A[parent] < A[child] then
swap(A[parent], A[child]);
child := parent;
parent := parent \ div \ 2
else parent := 0
```

end

# 2 AVL Trees

### **AVL** Trees

**Definition 1.** 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.

/\* Let G(h) denote the least possible number of nodes contained in an AVL tree of height h; the empty tree has height -1 and a single-node tree has height 0. A recurrence relation for G(h) can be defined as follows:

$$\begin{cases} G(-1) &= 0\\ G(0) &= 1\\ G(h) &= G(h-1) + G(h-2) + 1, \ h \ge 1 \end{cases}$$

A precise solution to G(h) may be derived by establishing the relation G(h) = F(h+3) - 1, where F(i) is the *i*-th Fibonacci number (as defined in Chapter 3.5 of Manber's book) for which we already know the closed form; the proof is quite simple by induction. So, for any AVL tree with *n* nodes and of height *h*,  $n \ge G(h) \ge F(h+3) - 1 \ge ca^h$  (for some positive constants *c* and *a* and sufficiently large *n*). It follows that  $h = O(\log n)$ . \*/

AVL Trees (cont.)

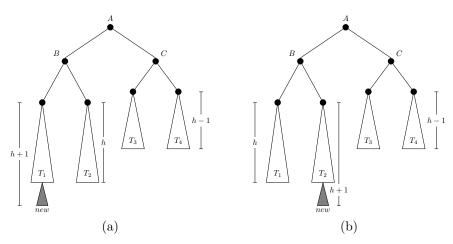
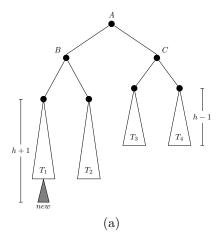
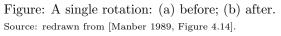
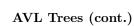


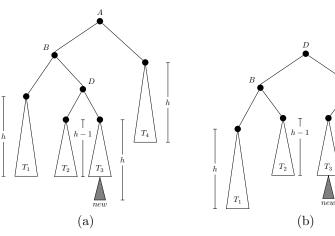
Figure: Insertions that invalidate the AVL property. Source: redrawn from [Manber 1989, Figure 4.13].

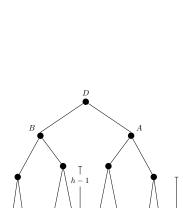
### AVL Trees (cont.)











 $T_{2}$ 

 $T_4$ 

 $T_2$ 

(b)

B

 $\dot{h} + 1$ 

 $T_1$ 

 $\sum_{new}$ 

A

 $T_3$ 

 $T_4$ 

Figure: A double rotation: (a) before; (b) after. Source: redrawn from [Manber 1989, Figure 4.15].

# 3 Union-Find

### Union-Find

- There are *n* elements  $x_1, x_2, \cdots, x_n$  divided into groups. Initially, each element is in a group by itself.
- Two operations on the elements and groups:
  - find(A): returns the name of A's group.
  - 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.)

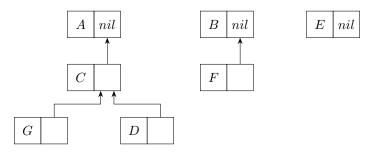


Figure: The representation for the union-find problem. Source: redrawn from [Manber 1989, Figure 4.16].

### 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 2** (Theorem 4.2). If balancing is used, then any tree of height  $h (\geq 0)$  must contain at least  $2^{h}$  elements.

/\* This can be proven by induction on the number  $n~(\geq 1)$  of elements/nodes. \*/

• Any sequence of m find or union operations (where  $m \ge n$ ) takes  $O(m \log n)$  steps.

### Union-Find (cont.)

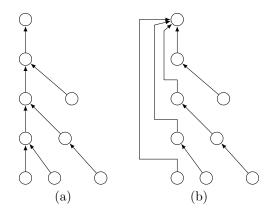


Figure: Path compression: (a) before; (b) after. Source: redrawn from [Manber 1989, Figure 4.17].

### Effect of Path Compression

**Theorem 3** (Theorem 4.3). If both balancing and path compression are used, any sequence of m find or union operations (where  $m \ge n$ ) 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
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
Code for Union-Find (cont.)
Algorithm Union(a,b);
begin
  x := Find(a);
  y := Find(b);
  if x \leftrightarrow 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
```

 $\operatorname{end}$