

# Inf 2B: Heaps and Priority Queues

Lecture 6 of ADS thread

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# Stacks, Queues, and Priority Queues

Stacks, queues, and *priority queues* are all ADTs for storing collections of elements. They differ in their access policy:

**Stacks:** Last-in-first-out (LIFO)

**Queues:** First-in-first-out (FIFO)

**Priority Queues:** Elements have a *priority* associated with them. An element with highest priority gets out first.

# The *PriorityQueue* ADT

- ▶ A *PriorityQueue* stores a collection of *elements*.
- ▶ Every element is associated with a *key*, which is taken from some linearly ordered set, such as the integers.
- ▶ Keys represent **priorities**:

*larger key* means *higher priority*.

Variant: *lower key* means *higher priority*.

- ▶ Not really different, just define a new order  $\leq^*$  on keys by

$$k_1 \leq^* k_2 \iff k_1 \geq k_2,$$

i.e., reverse existing order.

Different from *Dictionary*—here the meaning of a **key** is its **relative value** (in the collection).

# The *PriorityQueue* ADT

Methods of *PriorityQueue*:

- ▶ `insertItem( $k$ ,  $e$ )`: Insert element  $e$  with key  $k$ .
- ▶ `maxElement()`: Return an element with maximum key; an error occurs if the priority queue is empty.
- ▶ `removeMax()`: Return and remove an element with maximum key; an error occurs if the priority queue is empty.
- ▶ `isEmpty()`: Return `TRUE` if the priority queue is empty and `FALSE` otherwise.
- ▶ No `findElement( $k$ )` or `removeItem( $k$ )` methods (because  $k$  does not mean anything externally).

# The Search Tree Implementation

**Observation:** The **maximum** key in a binary search tree is always stored in the **rightmost interior vertex**.

Therefore, all *Priority Queue* methods can be implemented on an AVL tree with running time  $\Theta(\lg(n))$ .

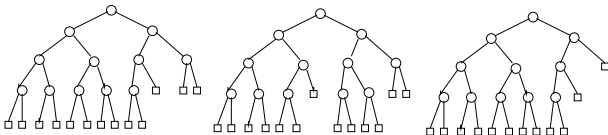
Could we do better?

`maxElement()` and `removeMax()` are *simpler* versions of the `findElement()` and `removeItem()` for *Dictionary*.

## Almost Complete Binary Trees

- ▶ All levels except maybe the last one have the maximum number of vertices.
- ▶ On the last level, all internal vertices are to the left of all leaves.

## Example Binary Trees



Which of these are “Almost Complete”?

Answer: First one only.

## Height of an Almost Complete Tree

**Theorem:** An almost complete binary tree with  $n$  internal vertices has height

$$\lfloor \lg(n) \rfloor + 1.$$

(We automatically have  $h = O(\lg n)$ ) **WHY?**

**Proof:** A *complete binary tree* of height  $h$  has  $2^h - 1$  internal vertices (proof by easy induction on  $h$ ).

For an *almost-complete tree*, of height  $h$  number of internal vertices  $n$  is:

- ▶ strictly more than number of internal vertices of a complete tree of height  $h - 1$ , so  $n \geq (2^{h-1} - 1) + 1 = 2^{h-1}$ ;
- ▶ at most the number of internal vertices of a complete tree of height  $h$ , so  $n \leq 2^h - 1 < 2^h$ .

Thus  $2^{h-1} \leq n < 2^h$ . Hence

$$\begin{aligned} h - 1 \leq \lg n < h &\Rightarrow h - 1 \leq \lfloor \lg n \rfloor < h \\ &\Rightarrow h = \lfloor \lg n \rfloor + 1. \end{aligned}$$



# Abstract Heaps

**Definition:** A *heap* is an almost complete binary tree whose internal vertices store items such that the following *heap condition* is satisfied:

(H) For every vertex  $v$  other than the root, **the key stored at  $v$  is smaller than or equal to the key stored at the parent of  $v$ .**

► So the maximum element is at the root.

The *last vertex* of a heap of height  $h$  is the rightmost internal vertex in the  $h$ th level.

## Finding the Maximum

**Algorithm** maxElement()

1. **return** *root.element*

Runtime is  $\Theta(1)$ .

# Insertion

**Algorithm** insertItem( $k, e$ )

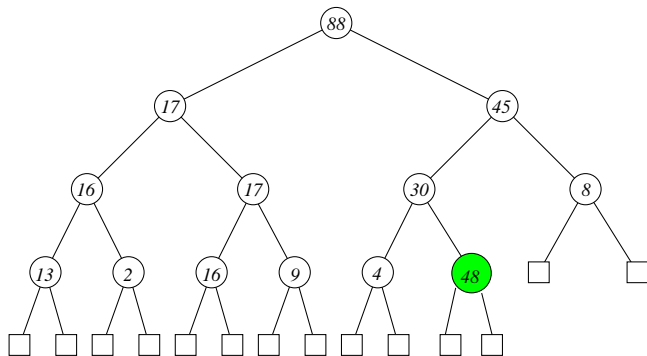
1. Create new last vertex  $v$ .
2. **while**  $v$  is not the root **and**  $k > v.parent.key$  **do**
3.       store the item stored at  $v.parent$  at  $v$
4.        $v \leftarrow v.parent$
5. store  $(k, e)$  at  $v$

“Bubble” the item up the tree.

Basically **swap**  $v$  with  $v.parent$  if  $v$ 's key is bigger.

Takes  $\Theta(1)$  for adding new last vertex (initially), and  $\Theta(1)$  for every **swap**. Hence  $\Theta(\lg n)$  worst-case in total.

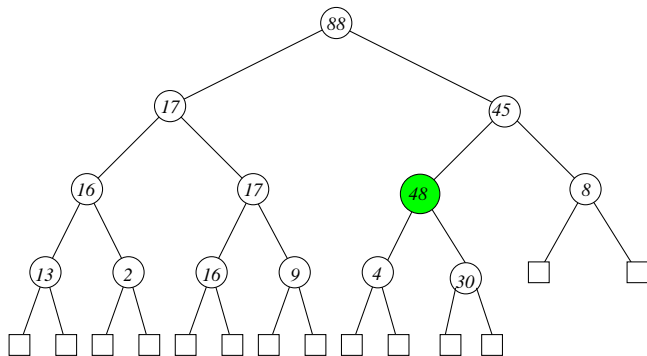
# insertItem



insertItem(48), first add at “last vertex”.

Need to swap 48 with parent 30, because  $48 > 30$ .

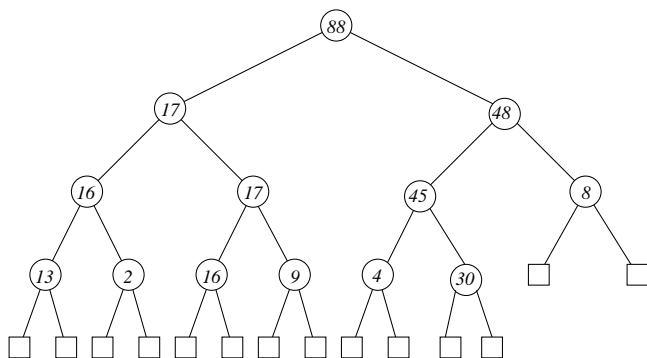
# insertItem



48 has now moved-up

Now need to swap 48 with parent 45, because  $48 > 45$ .

# insertItem



Done. 48 is less than root 88, no swap needed.

## Removing the Maximum

- ▶ **Idea:** Copy item in "last vertex" into root.
- ▶ Delete last vertex (*easy to delete at end of tree*).
- ▶ Now *parent greater than child* property might be false. Need to fix.
- ▶ New method  $\text{Heapify}(v)$ :
  - ▶ Let  $s$  be  $v.\text{left}$  or  $v.\text{right}$  (whichever has max key).
  - ▶ Swap  $s$  and  $v$ .
  - ▶ Call  $\text{Heapify}()$  recursively.
- ▶  $\Theta(h) = \Theta(\lg n)$  time in total. Formal proof in notes.

# Removing the Maximum

## Algorithm removeMax()

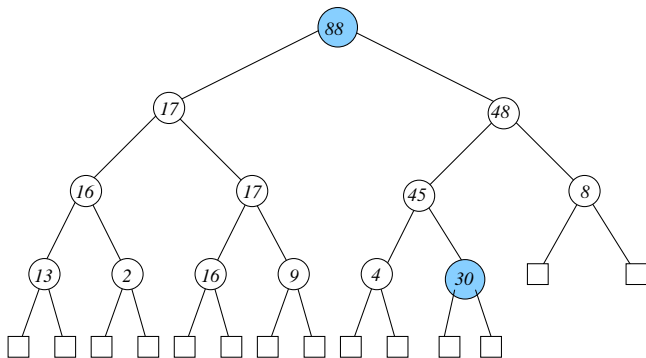
1.  $e \leftarrow \text{root.element}$
2.  $\text{root.item} \leftarrow \text{last.item}$
3. delete *last*
4. heapify(*root*)
5. **return**  $e$ ;

## Algorithm heapify( $v$ )

1. **if**  $v.\text{left}$  is an internal vertex  
**and**  $v.\text{left.key} > v.\text{key}$  **then**
2.      $s \leftarrow v.\text{left}$
3. **else**
4.      $s \leftarrow v$
5. **if**  $v.\text{right}$  is an internal vertex  
**and**  $v.\text{right.key} > s.\text{key}$  **then**
6.      $s \leftarrow v.\text{right}$
7. **if**  $s \neq v$  **then**
8.     swap the items of  $v$  and  $s$
9.     heapify( $s$ )

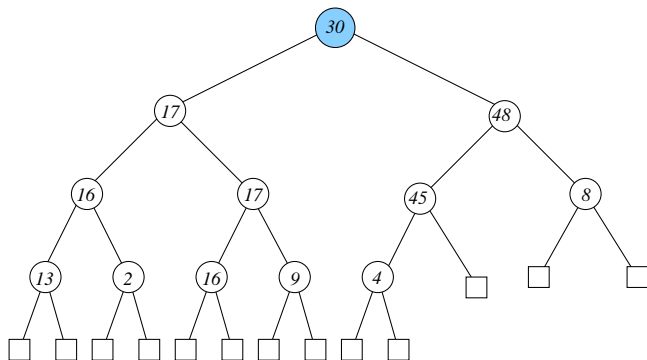


## removeMax



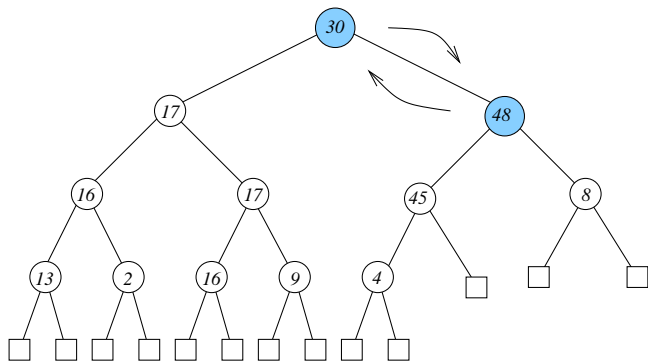
Need to copy over "last vertex" onto root.

## removeMax



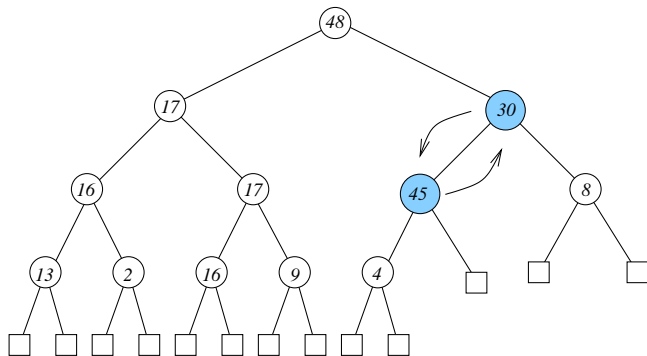
Now we call `heapify(root)`.

## removeMax



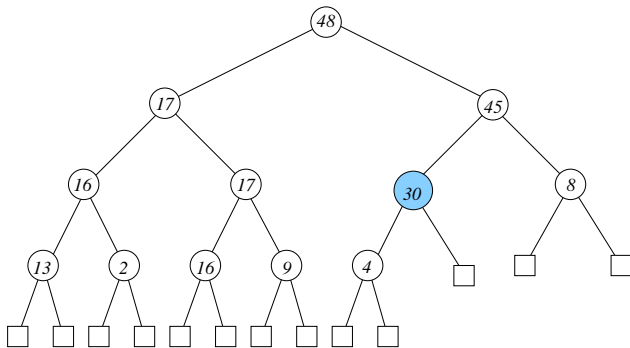
Max child of root is 48 on right, need to swap, and then call heapify on 30 as the child.

## removeMax



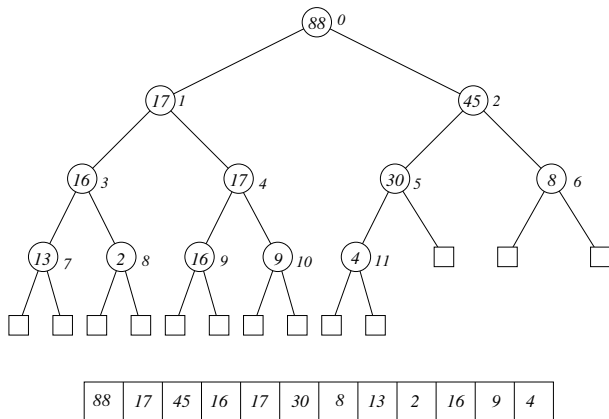
Max child of 30 is 45 on left, need to swap,  
and then call heapify on 30 as the child.

## removeMax



Max child of 30 is 4, less than 30. ok. Finish.

## Storing Heaps in Arrays



Direct mapping:  $j$ -th element of heap stored in index  $j - 1$ .  
Can use  $(2^i - 2) + j$  for index of  $j$ th element on level  $i$ .  
(depends on "Almost-complete" property).

## Working on Heaps as Arrays

- ▶ `maxElement()`: Just look at index 0 of array.
- ▶ `insertItem(k, e)`: Insert into index *size*.
  - ▶  $size \leftarrow size + 1$ .
  - ▶ Do "bubbling" using array structure:
    - ▶  $v$ 's left child is in index  $2v + 1$ ;
    - ▶ right child in index  $2v + 2$ .
- ▶ `removeMax()`:
  - ▶ Copy item at  $size - 1$  into index 0.
  - ▶  $size \leftarrow size - 1$ .
  - ▶ Do "swapping" using array structure.
- ▶ Using dynamic arrays get  $\Theta(\lg n)$  amortised time for `insertItem(k, e)` and `removeMax()`.

# Turning an Array into a Heap

**Algorithm** buildHeap( $H$ )

1.  $n \leftarrow H.length$
2. **for**  $v \leftarrow \lfloor \frac{n-2}{2} \rfloor$  **downto** 0 **do**
3.     heapify( $v$ )

**Theorem:** The running time of buildHeap is  $\Theta(n)$ , where  $n$  is the length of the array  $H$ .



## Resources

- ▶ The Java Collections Framework has an implementation of *PriorityQueue* (using heaps) in its `java.util` package:  
`http://java.sun.com/j2se/1.5.0/docs/api/java/util/PriorityQueue.html`
- ▶ If you have [GT]: read the “Priority Queues” chapter
- ▶ If you have [CLRS]: look at the “Heapsort” chapter (but ignore the sorting for now).