

Priority Queues

A priority queue has to perform these two operations fast!

1. Insert an element into the set

2. Find the smallest element in the set

Insert

Find Min

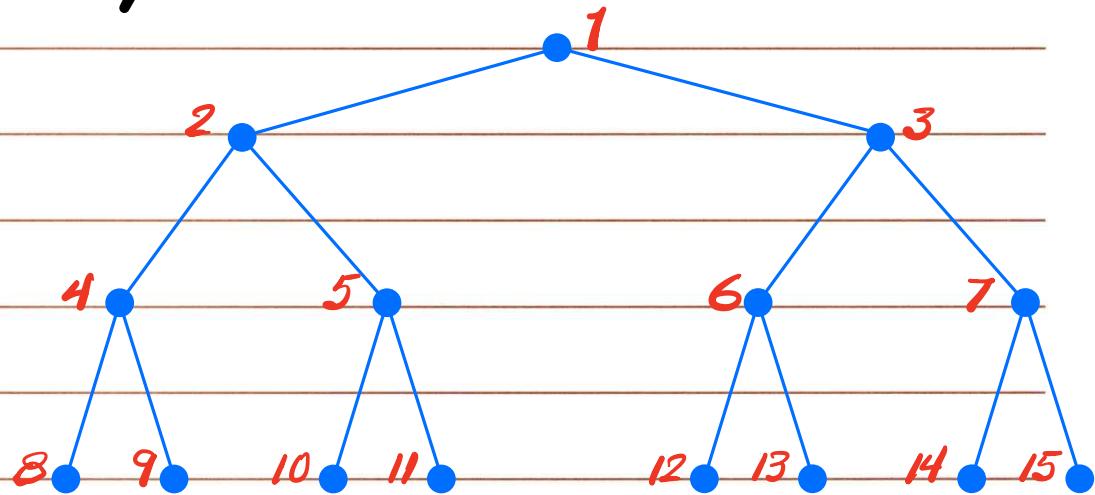
Array Implementation $O(1)$ $O(n)$

Sorted array $O(n)$ $O(1)$

Linked List $O(1)$ $O(n)$

Sorted linked list $O(n)$ $O(1)$

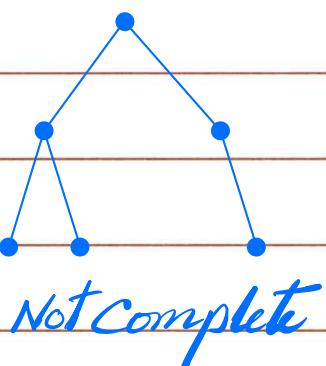
Def. A binary tree of depth k which has exactly $2^k - 1$ nodes is called a full binary tree.



Def. A binary tree with n nodes and of depth k is complete iff its nodes correspond to the nodes which are numbered 1 to n in the full binary tree of depth k .



complete



Not complete

Traversing a complete binary tree stored as an array.

$\text{Parent}(i)$ is at $\lfloor i/2 \rfloor$ if $i \neq 1$

if $i=1$, i is the root

$\text{Lchild}(i)$ is at $2i$ if $2i \leq n$

otherwise, it has no left child

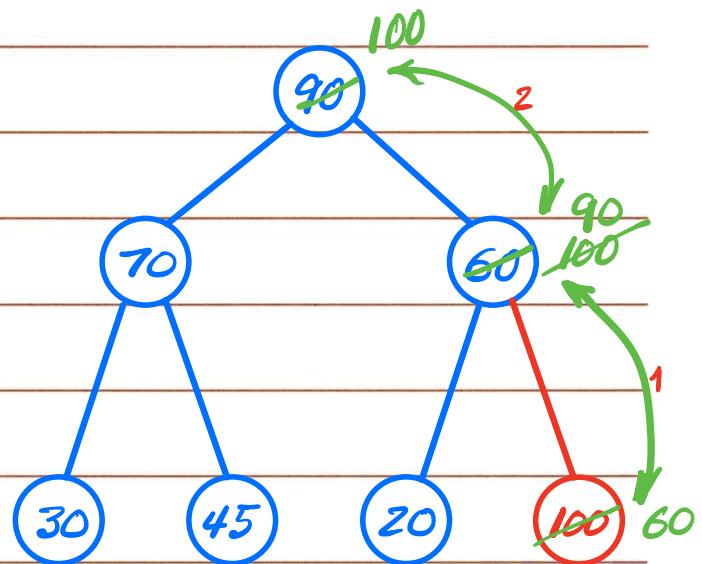
$\text{Rchild}(i)$ is at $2i+1$ if $2i+1 \leq n$

otherwise, it has no right child

Def. A binary heap is a complete binary tree with the property that the value of the key at each node is at least as large as the key values at its children (Maxheap)

Find-Max

Takes $O(1)$



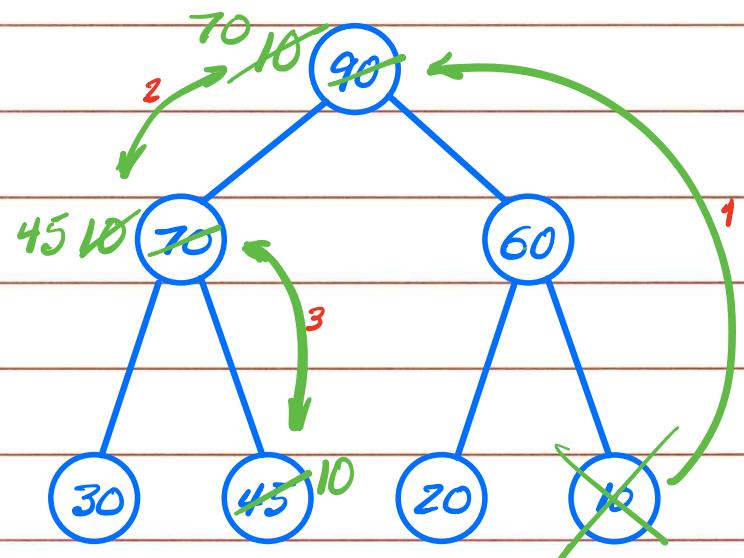
Insert

ex. insert 100

Takes $O(\lg n)$

Extract-Max

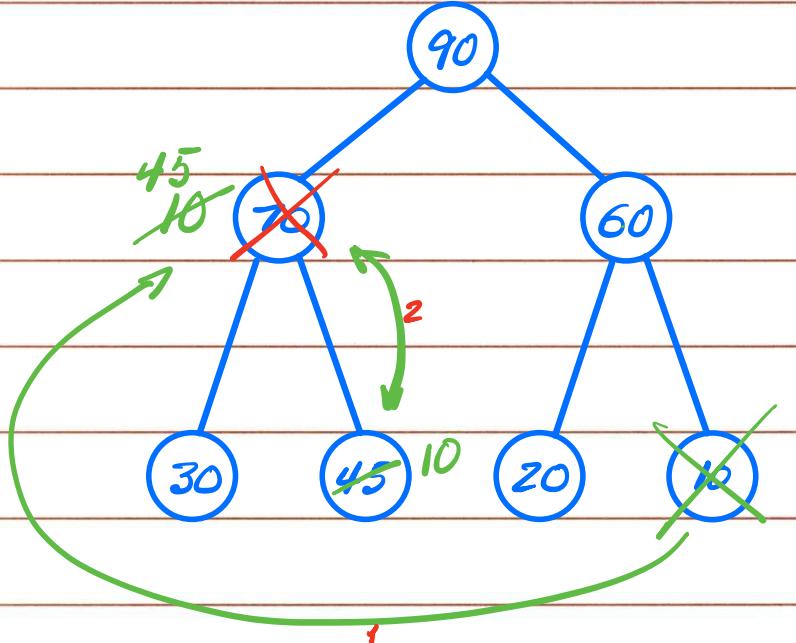
Takes $O(\lg n)$



Delete

ex. Delete 70

Takes O(lgn)

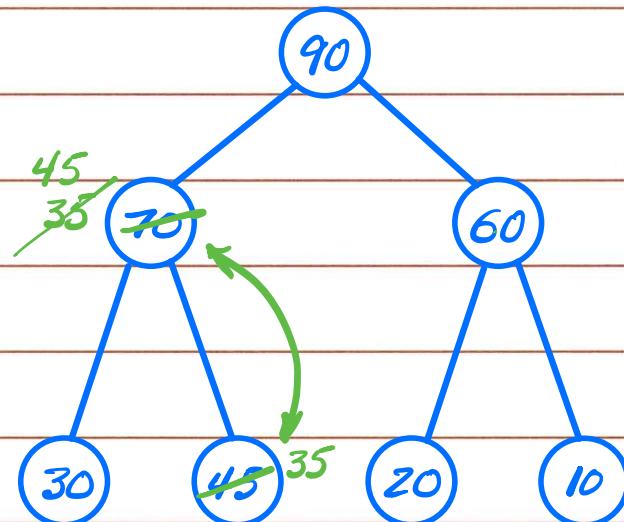


Decrease-key

ex. Decrease-key

70 to 35

Takes O(lgn)



Construction

Can be done in $O(n \lg n)$ time using n insert operations.

Can we do better?

Bottom up construction

of nodes

1



of swaps
 $\lg n$

$n/8$

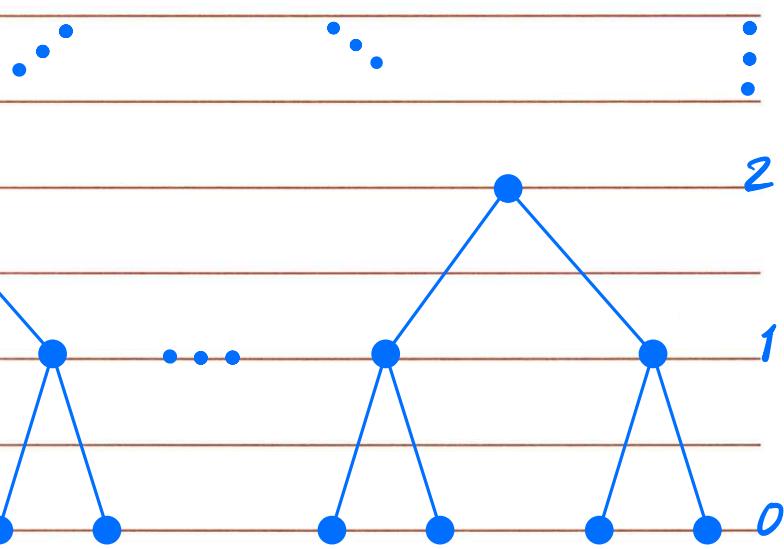
$n/4$

$n/2$

2

1

0



$T = \text{Maximum number of swaps needed}$

$$T = n/4 * 1 + n/8 * 2 + n/16 * 3 + \dots$$

$$T_{1/2} = n/8 * 1 + n/16 * 2 + n/32 * 3 + \dots$$

$$T - T_{1/2} = n/4 * 1 + \underbrace{n/8 * 1 + n/16 * 1}_{n/2} + \dots$$

$$T_{1/2} = n/2$$

$$T = n$$

Bottom up construction takes $O(n)$

Q: What is the best run time to merge two binary heaps of size n ?

A: $O(n)$ using linear time construction

Finding the top k elements in an array.

Input: An unsorted array of length n.

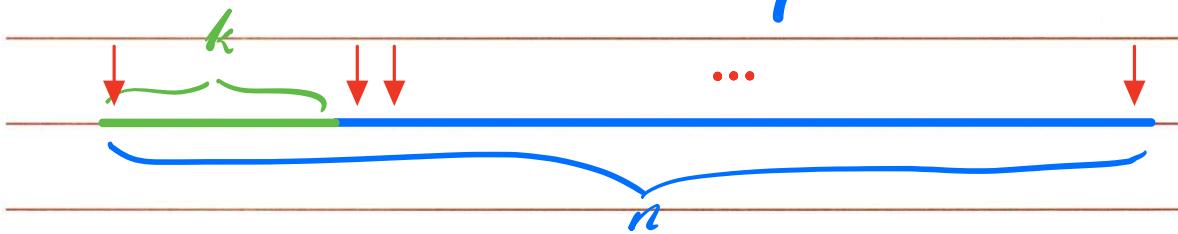
Output: Top k values in the array ($k < n$)

Constraints:

- You cannot use any additional memory
- Your algorithm should run in time $O(n \lg k)$

Solution:

- Construct a Minheap of size k .



- More through the rest of the $(n-k)$ elements.
If we encounter an element larger than the element at the top of our minheap, swap them.

This involves extract-min and insert ($O(\lg k)$)

$$\text{Overall complexity} = O(k) + O((n-k)\lg k) = O(n\lg k)$$