

Heaps

Note Title

11/1/2007

A heap is a completely filled binary tree (ok, maybe not full leaves). leaves filled left to right.

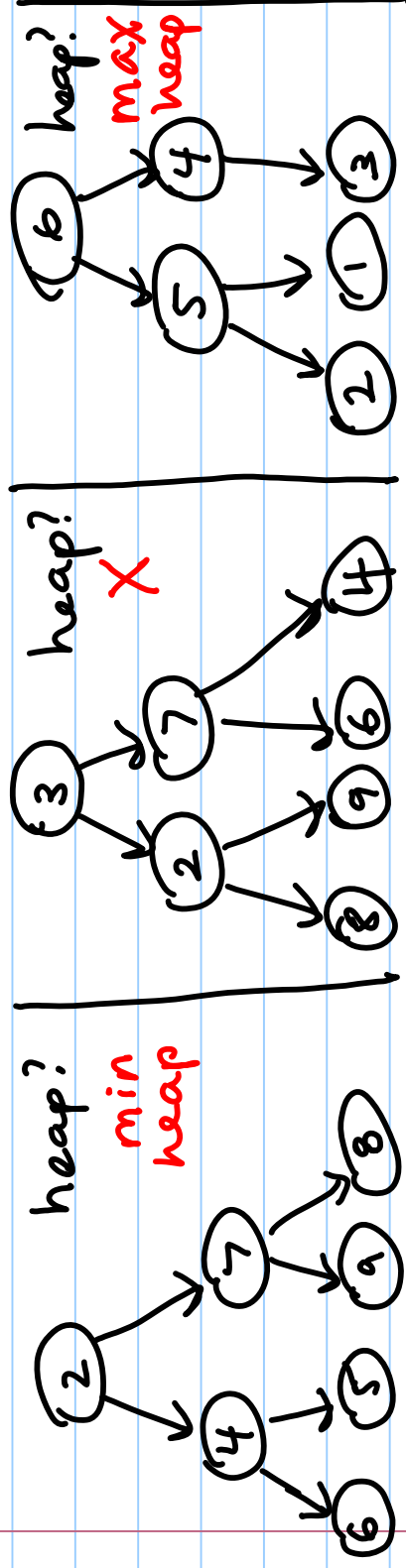
— Max-Heap

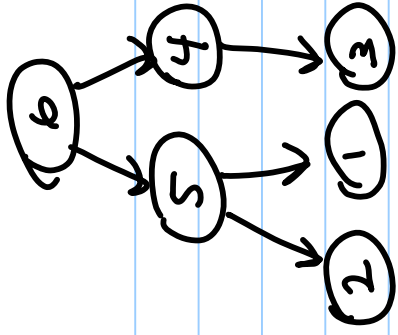
For all nodes, value of parent \geq value of self

OR

— Min Heap

For all nodes, value of parent \leq value of self





How long to find $\textcircled{1}$ in the heap?

worst-case search: n

worst-case height: $\log n$

What is easy to find in a max-heap?

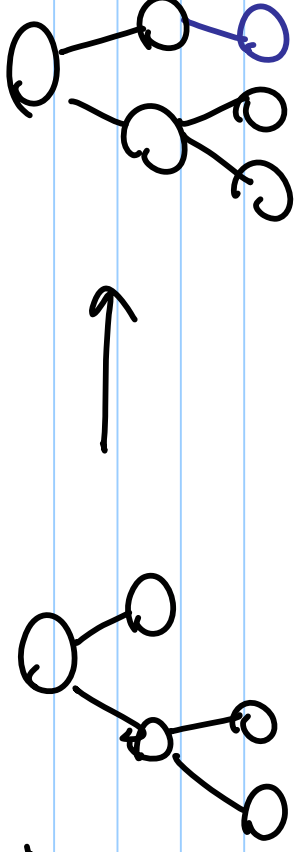
max is $O(1)$

Insert into a heap

- due to the structure of a heap, you always know the shape a heap will have after it goes from

n to $n+1$ elements.

example:



Insert into a heap

-so, to insert a node into a heap,

① add new node to leftmost open position in leaves

② adjust as needed to ensure heap property is met.

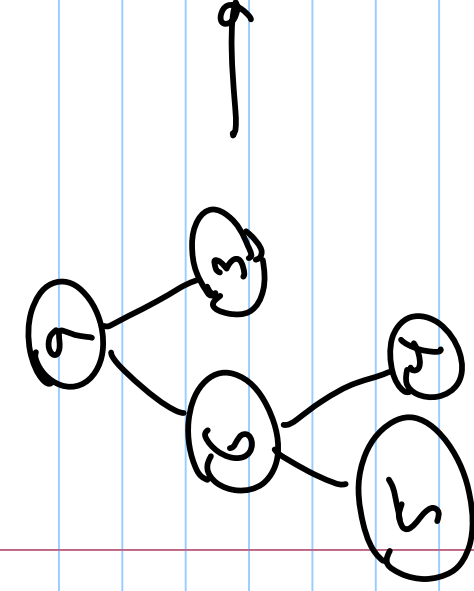
→ (a) compare new node to parent.
(b) swap if new node bigger.

(c) Repeat until swap not needed
or new node is root.

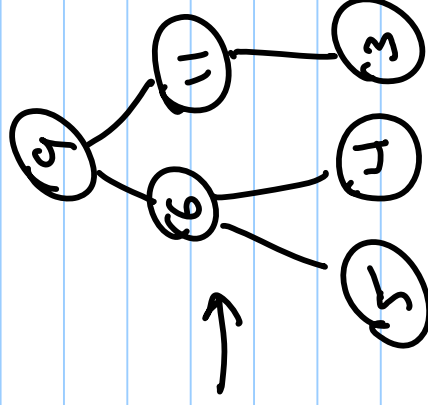
example:

Insert 11.

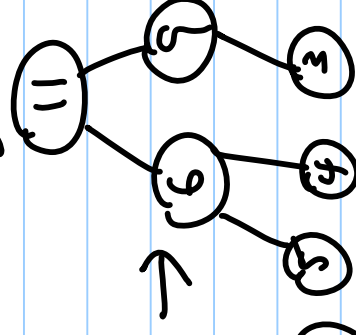
Step 1.



11 > 3
Step 2b.



11 > 9
Step 2b
again.



Step 2 called "bubbling up" the new node.

Insert runtime:

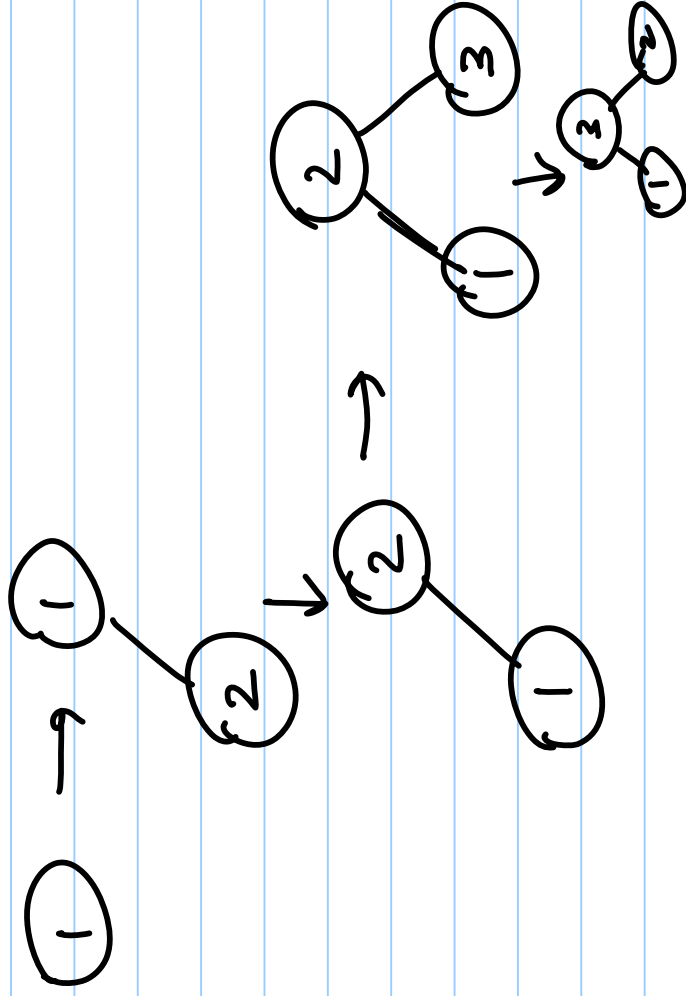
worst case \rightarrow bubble new node up all the way to root

$O(\log n)$

$\underbrace{\hspace{1.5cm}}$ length of path from new node to root

Create a heap from an array of elements (Method I):

- one way is to insert n elements into the heap one after the other

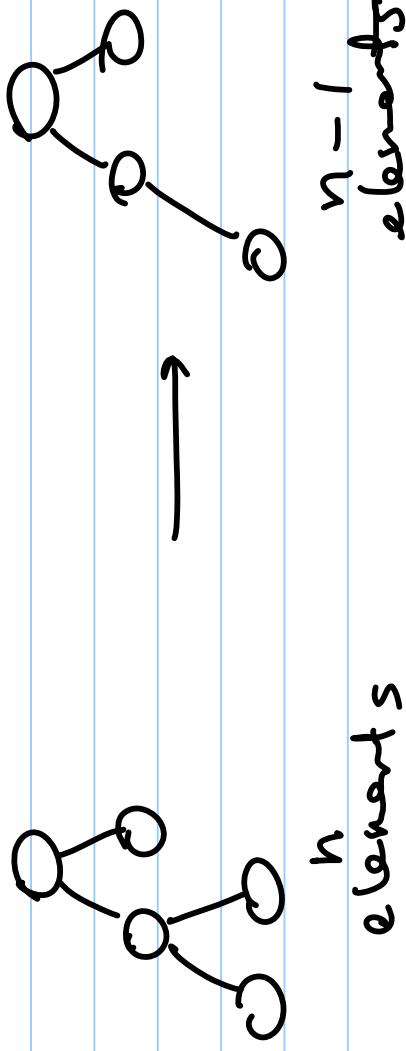


n calls to insert.

upper bound $O(n \log n)$

Delete Root.

We know the shape of a heap with $n-1$ elements.



Delete Root

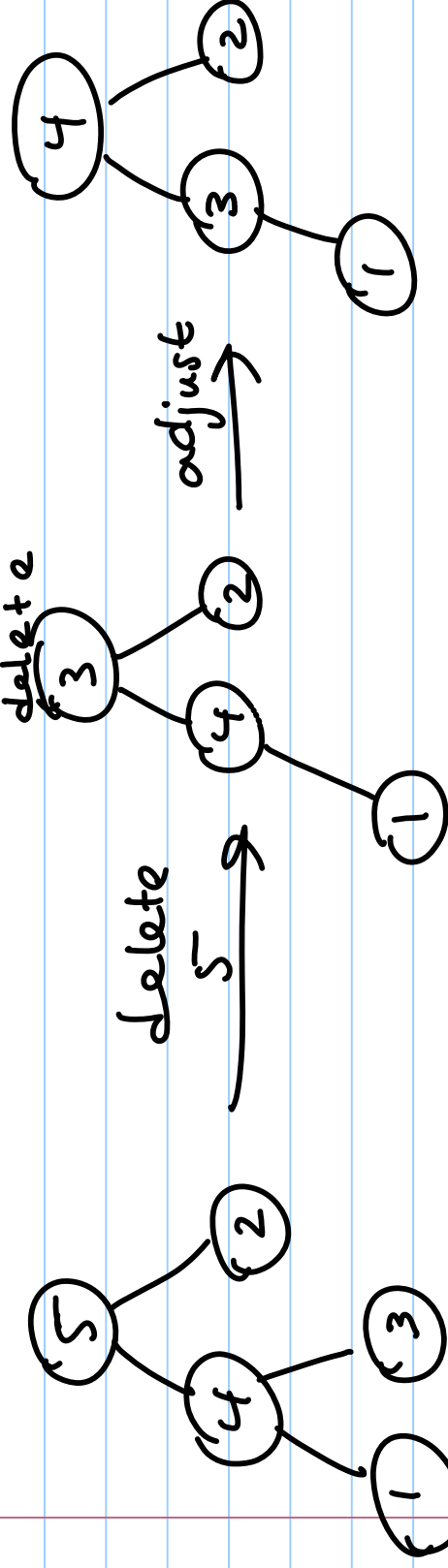
- ① Swap root with rightmost leaf
- ② Delete old root
- ③ Adjust to maintain heap property.

① Let $n = \text{new root}$, $i = \emptyset$
② Let $\text{largest} = \max(n, n.\text{left}, n.\text{right})$;

③ If $(\text{largest} \neq n) \{$
 swap (max, n) // n now on
 level $i+1$
 $i++$;
}

④ Repeat until no more swaps
or n is a leaf

Step ③ called "Bubbling Down"



Step 3b is two comparisons.

worst case - bubble down all the way
from root to leaf.

max path length $O(\log n)$

so runtime is $O(\log n)$

Another name for "Bubble Down"
is MAX-HEAPIFY.

1. Assume children of root are both
heaps
2. Bubble down root if needed so
whole data structure maintains heap
property.

Heap Sort (Array $A[1 \dots n]$) {

① Turn A into a max-heap.

→ ② Extract the Max using
Delete Root.

③ Repeat step 2 until only
one element
left in heap.
That element is min.

} . steps 2 and 3 called
"harvesting the heap"

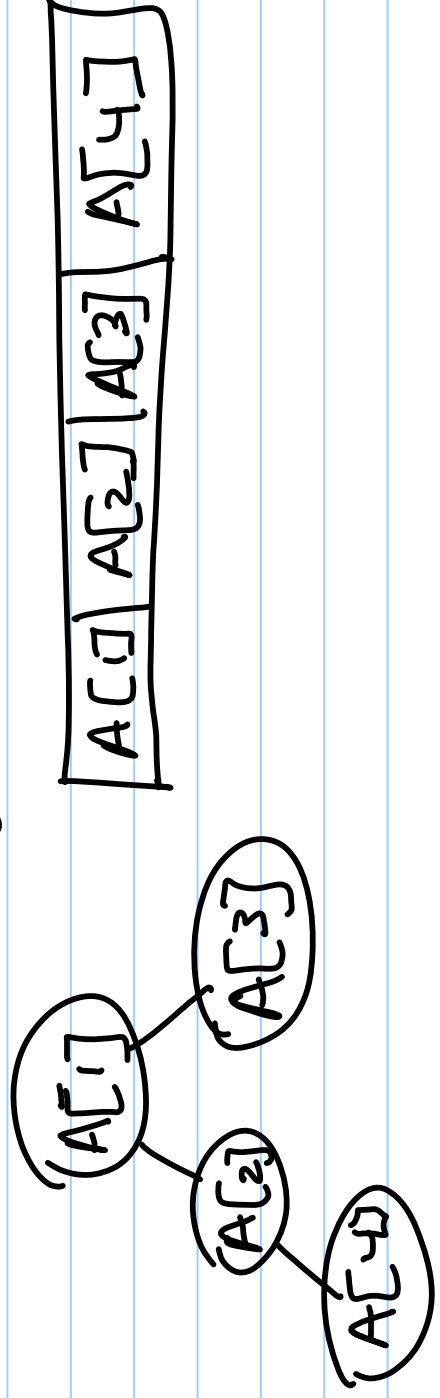
Runtime of Heapsort .

Step 1 : Build the heap.

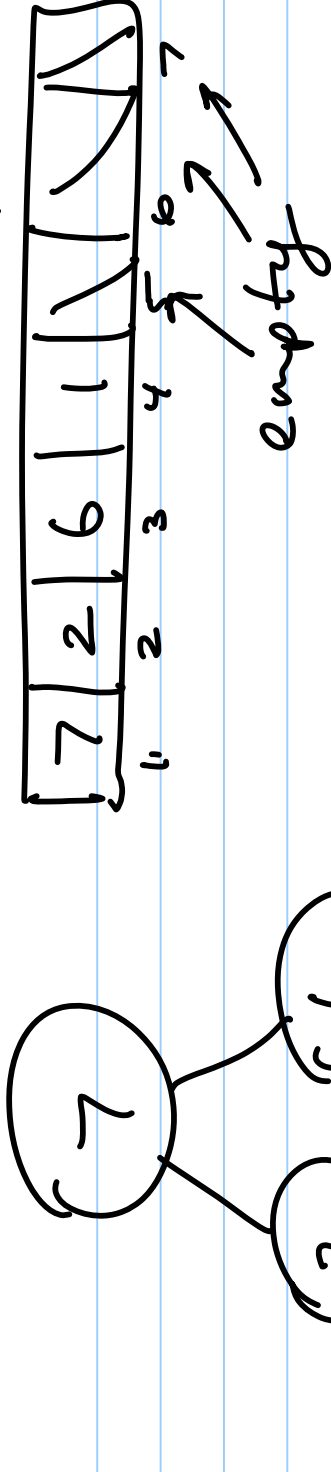
We know upper bound is $n \log n$.

Is there a way to build in place?

Treat array as a heap:



Storing a tree in an array...



Index of left child: $2 \cdot \text{index of parent}$

index of right child: $(2 \cdot \text{index of parent}) + 1$

(one-based indexing)

BUILD-MAX-HEAP {

① Treat array like a candidate heap.

② All leaves are heaps, so start at level before leaves.

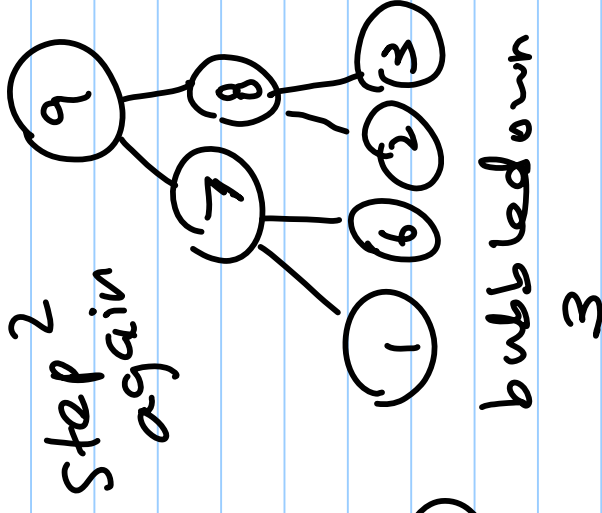
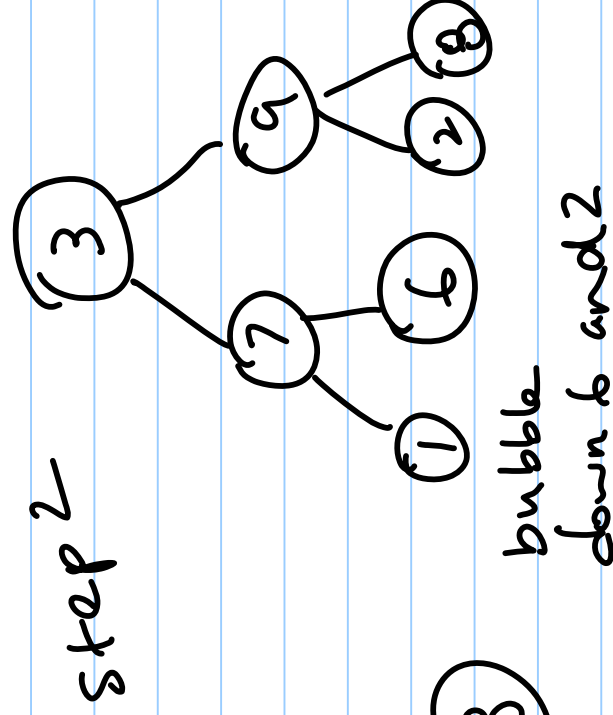
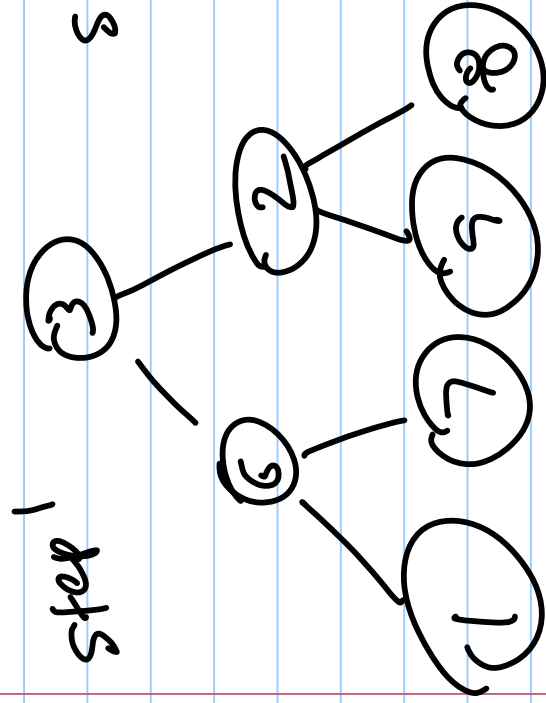
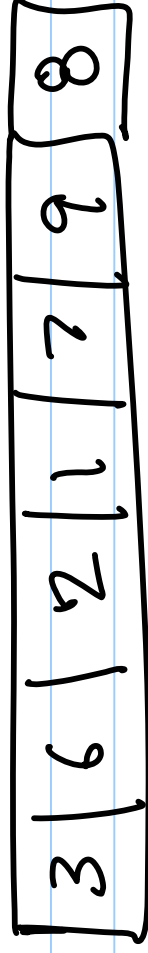
→ Go from left to right,
calling MAX-HEAPIFY on each node.

③ Repeat on all higher levels until you finish calling MAX-HEAPIFY on root.

}

So you call MAX-HEAPIFY around $\frac{n}{2}$ times.

Example of Build-Max-Heap:



Steps 2 and 3: Call DeleteRoot repeatedly
(harvest heap)

Is there a way to harvest heap in place?
Yes.

Let $last = n$

Put max in $A[last]$.

Treat heap as $A[1 \dots (last-1)]$

$last--;$

Repeat.

Runtime of Heapsort?

- ① Build heap in place.
- ② Harvest heap in place.

Worst case of Step 1: $O(n \log n)$.

But note, when at low levels,

performance of MAX-HEAPIFY is not $O(\log n)$, because height of subtree rooted at lower level nodes $< \log n$.

Runtime of BUILD-MAX-HEAP

$$\text{is } \sum_{h=1}^{\log_2 n} \left(\begin{array}{c} \# \text{ nodes} \\ \text{with height } h \end{array} \right) O(h).$$

$$= \sum_{h=1}^{\log_2 n} \left(\begin{array}{c} \# \text{ nodes} \\ \text{with height } h \end{array} \right) (a^h)$$

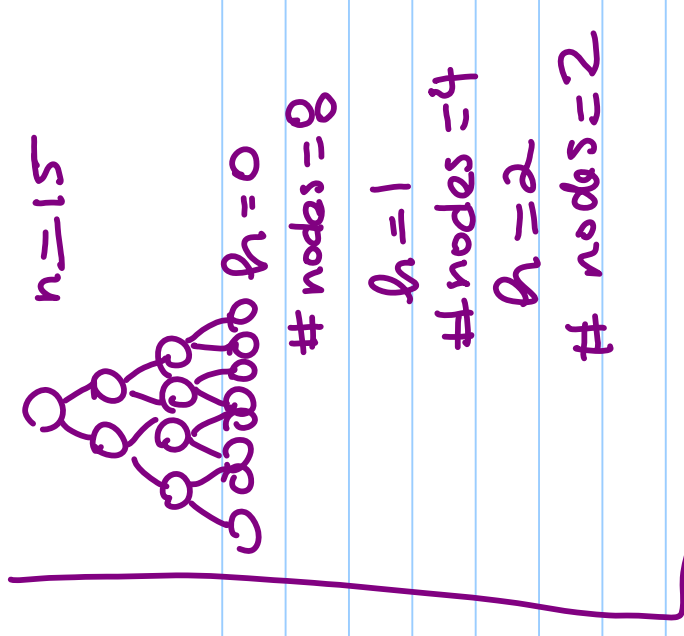
$$= \sum_{h=0}^{\log_2 n} \left(\begin{array}{c} \# \text{ nodes} \\ \text{with height } h \end{array} \right) (a^h)$$

$$= a \sum_{h=0}^{\log_2 n} \left\lceil \frac{n}{2^{h+1}} \right\rceil (a^h)$$

(ignore ceiling)

$$= \frac{1}{2} a n \sum_{h=0}^{\log_2 n} h \left(\frac{1}{2} \right)^h$$

$$\leq \frac{1}{2} a n \sum_{h=0}^{\infty} h \left(\frac{1}{2} \right)^h = \frac{1}{2} a n \left[\frac{\frac{1}{2}}{(1-\frac{1}{2})^2} \right]$$



runtime \leq an

runtime $\in O(n)$ Wow!

derivation of $\sum_{k=0}^{\infty} k(x)^k = \frac{x}{(1-x)^2}, \quad x < 1$

$$\cdot \sum_{k=0}^{\infty} (x)^k = \frac{1}{1-x}, \quad x < 1$$

derivative of both sides

$$\sum_{k=0}^{\infty} k x^{k-1} = (-1) \left(\frac{1}{(1-x)^2} \right) (-1)$$

multiply
both sides
by

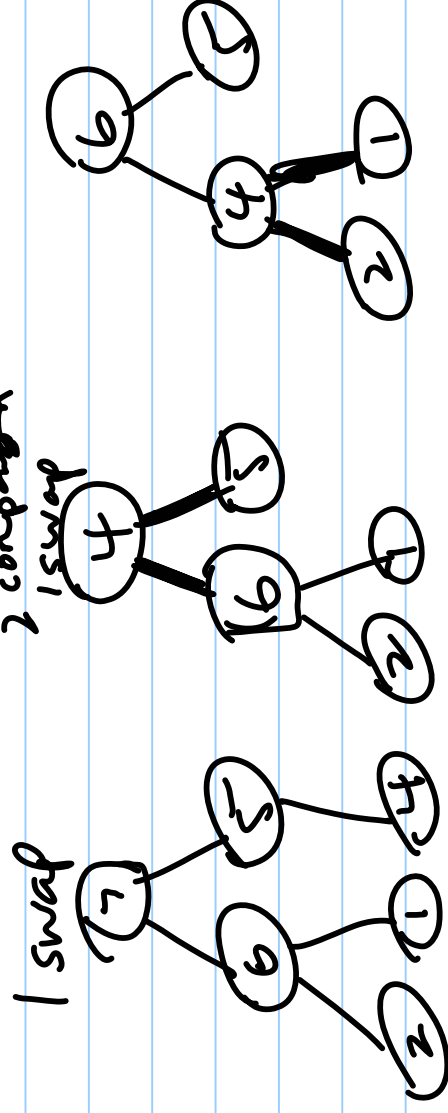
$$\sum_{k=0}^{\infty} kx^k = \frac{x}{(1-x)^2} \quad .$$

Runtime of Heapsort?

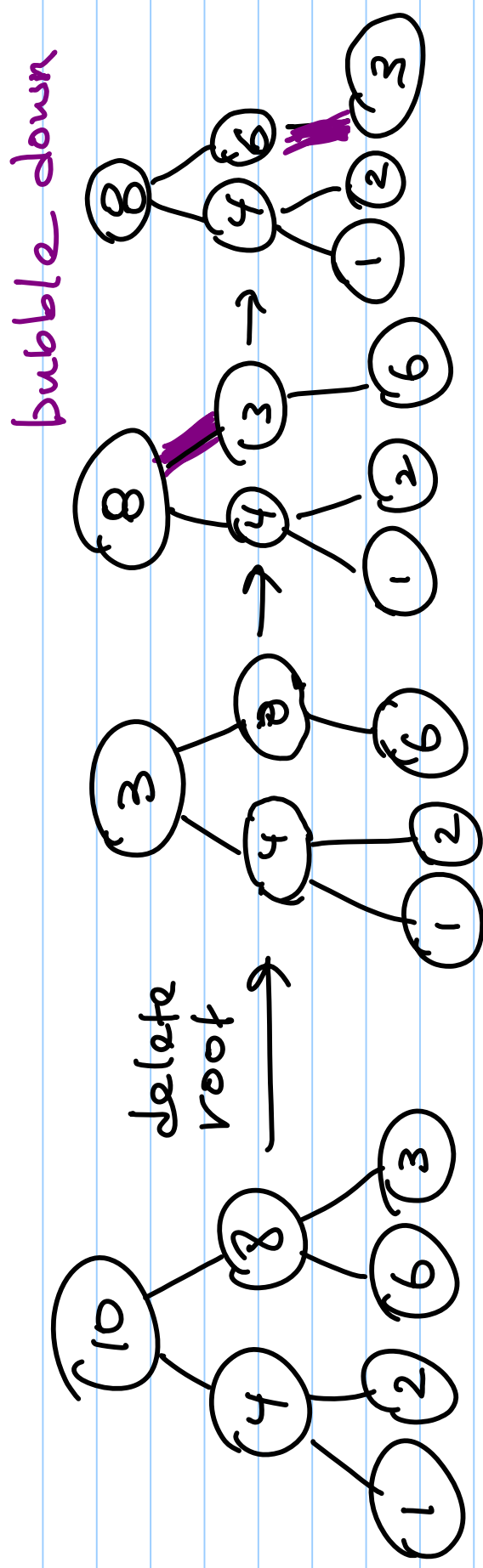
① Build heap in place: $O(n)$

② Harvest heap in place:

Call Delete Root $O(n)$ times.



Each call to DeleteRoot starts with a new node at the ROOT which could bubble all the way down to a leaf.



worst
case
So path length for all Delete Root
calls is $\log_2 n$.

So harvesting runtime is $O(n \log n)$.

Runtime of Heapsort?

- ① Build heap in place: $O(n)$
 - ② Harvest heap in place: $O(n \log n)$
- TOTAL RUNTIME: $O(n \log n)$

Another use of a heap
is a priority queue.

Convenient because if you have a multithreaded
machine
you can return max to one thread that
needs it
and heapify in another thread.