

Chapter - 6

Heap sort

❊ Combine the advantages of both merge sort and insertion sort.

- $O(n \lg n)$ worst-case running time
- sorts in place
- when array nearly sorted, runs fast in practice

❊ Heap is combined of a binary tree and an array

1	2	3	4	5	6	7	8	9	10
16	14	10	8	7	9	3	2	4	1

PARENT (i)

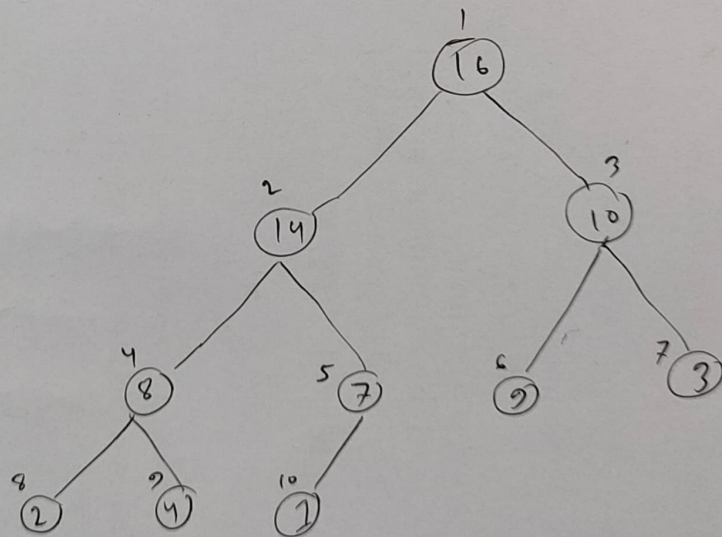
return $\lfloor i/2 \rfloor$

LEFT (i)

return $2i$

RIGHT (i)

return $2i+1$



- Always fill from the left.

⇒ In a max heap, parent will be larger or equal to the child
 $\text{parent} \geq \text{child}$

⇒ In min heap
 $\text{parent} \leq \text{child}$

⊗ Height of a node
 \Rightarrow longest simple downward path from node to a leaf

⊗ Height of the heap/root $\Rightarrow \theta(\lg n)$

⊗ List of function!

MAX-HEAPIFY $\Rightarrow O(\lg n)$

BUILD-MAX-HEAP $\Rightarrow O(n)$

HEAPSORT $\Rightarrow O(n \lg n)$

MAX-HEAP-INSERT $\Rightarrow O(\lg n)$

HEAP-EXTRACT-MAX $\Rightarrow O(\lg n)$

HEAP-INCREASE-KEY $\Rightarrow O(\lg n)$

HEAP-MAXIMUM $\Rightarrow O(\lg n)$

⊗ MAX-HEAPIFY (A, i)

$l = \text{LEFT}(i)$

$r = \text{RIGHT}(i)$ \rightarrow l exist on not

if $l \leq A.\text{heap-size}$ and $A[l] > A[i]$

largest = l

else

largest = i

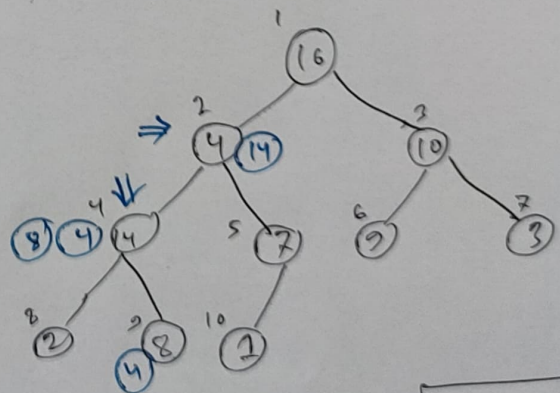
if $r \leq A.\text{heap-size}$ and $A[r] > A[\text{largest}]$

largest = r

if largest $\neq i$

exchange $A[i]$ with $A[\text{largest}]$

MAX-HEAPIFY (A, largest)



MAX-HEAPIFY (A, 2)
 MAX-HEAPIFY (A, 4)

can be called upto $\frac{2n}{3}$ times

Run time =

$T(n) = T(\frac{2n}{3}) + \theta(1)$

$= O(\lg n)$

\rightarrow maximum number of children

\rightarrow only one side will be call & divided into $\frac{2n}{3}$

Building a Heap

⇒ BUILD-MAX-HEAP(A, n)

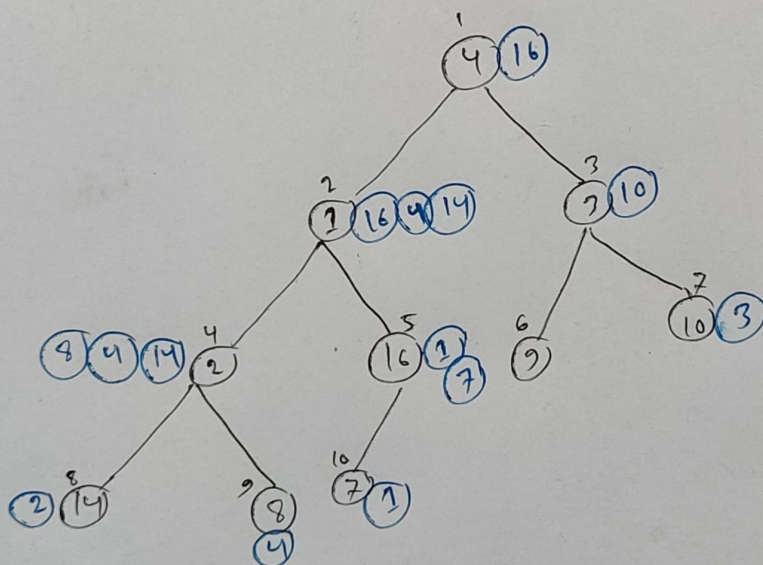
A.heap-size = n

for $i = \lfloor n/2 \rfloor$ downto 1

MAX-HEAPIFY(A, i)

$\lfloor \frac{n}{2} \rfloor = \frac{10}{2} = 5 \Rightarrow 5+1 = 6$ to 10 all leaf
= already a heap, no need to modify

1	2	3	4	5	6	7	8	9	10
4	1	3	2	16	9	10	14	8	7
16	14	10	8	7	9	3	2	4	1



MAX-HEAPIFY(A, 5)

4
3
2
1

Analysis 1:

MAX-HEAPIFY cost $O(\lg n)$

need to call $O(n)$

∴ run time = $O(n \lg n)$

Analysis 2:

for n ~~ele~~ element, height is $\lceil \lg n \rceil$

nodes, in max $\lceil \frac{n}{2^{h+1}} \rceil$

Time required for MAX-HEAPIFY = $O(h)$

$$\text{Then run time} = \sum_{h=0}^{\lceil \lg n \rceil} \lceil \frac{n}{2^{h+1}} \rceil O(h)$$

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

$$= O \left(n \sum_{h=0}^{\infty} \frac{h}{2^{h+1}} \right)$$

$$= O \left(n \cdot \frac{\frac{1}{2}}{(1-\frac{1}{2})^2} \right)$$

$$= O(n \cdot 2)$$

$$= O(n)$$

⊕ HEAPSORT (A, n)

BUILD-MAX-HEAP (A, n) $\Rightarrow O(n)$

for $i = n$ down to 2

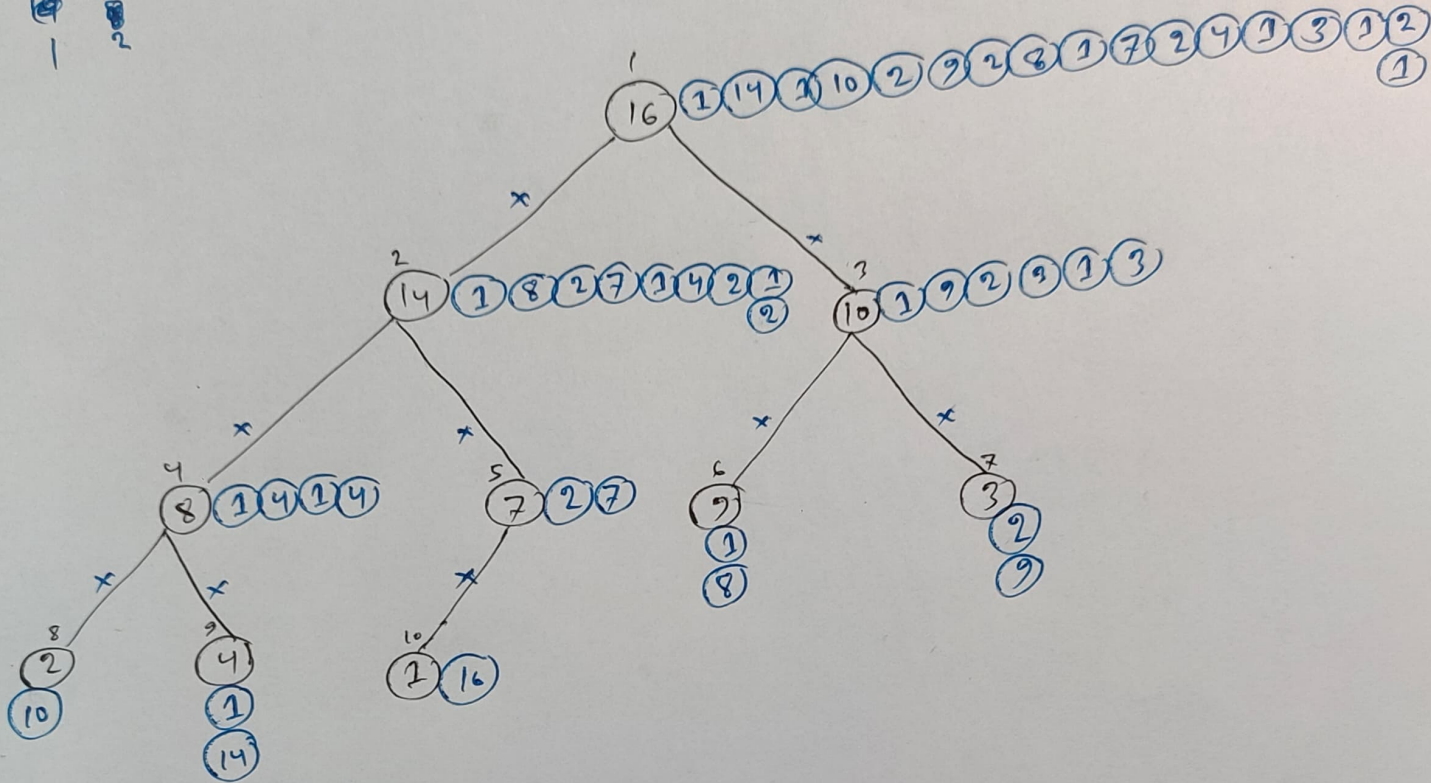
exchange $A[1]$ with $A[i]$

$A.\text{heap-size} = A.\text{heap-size} - 1$

MAX-HEAPIFY (A, 1) $\Rightarrow O(\lg n)$

$O(n \lg n)$

1	2	3	4	5	6	7	8	9	10
16	14	10	8	7	9	3	2	4	1
1	2	3	4	7	8	9	10	14	16



Priority Queue

- (i) max-priority \Rightarrow max-heap base
- (ii) min-priority \Rightarrow min-heap base

⊗ MAX-HEAP-MAXIMUM(A)

if $A.\text{heap-size} < 1$

error "heap underflow"

return $A[1]$

$\Theta(1)$

⊗ MAX-HEAP-EXTRACT-MAX(A)

$\theta(1)$ {

 $\text{max} = \text{MAX-HEAP-MAXIMUM}(A)$

 $A[1] = A[A.\text{heap-size}]$

 $A.\text{heap-size} = A.\text{heap-size} - 1$

 $\theta(\lg n) \leftarrow \text{MAX-HEAPIFY}(A, 1)$

 return max

$O(\lg n)$

⊗ MAX-HEAP-INCREASE-KEY(A, x, k)

Previous value $\Rightarrow x \leq k$
 New value

$\theta(1)$ {

 if $k < x.\text{key}$

 error "new key is smaller than current key"

 $x.\text{key} = k$

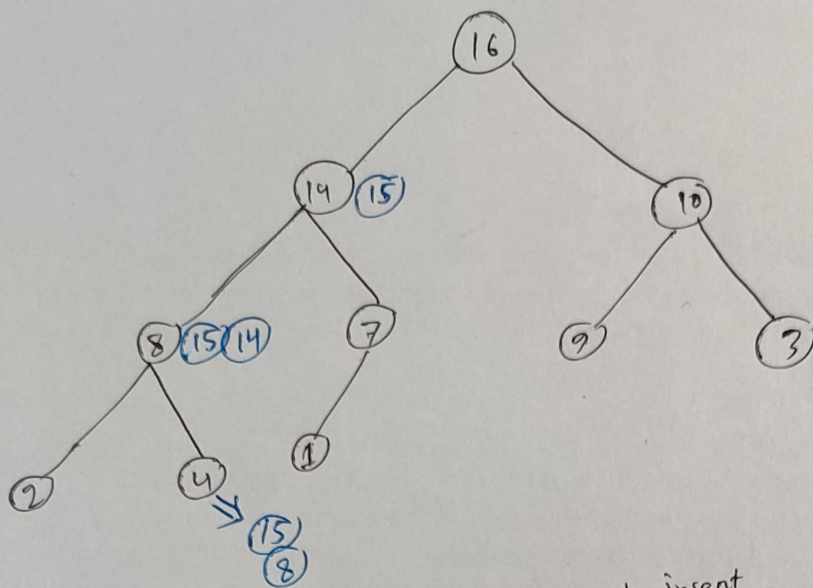
$\theta(n) \leftarrow$ find the index i in array A where object x occurs

$\theta(\lg n) \leftarrow$ while $i > 1$ and $A[\text{PARENT}(i)].\text{key} < A[i].\text{key}$

exchange $A[i]$ with $A[\text{PARENT}(i)]$, updating the information that maps priority queue object to array indices

$i = \text{PARENT}(i)$

\Rightarrow Total runtime = $O(\lg n)$



⊗ MAX-HEAP-INSERT (A, x, n)

key to insert

heap-size

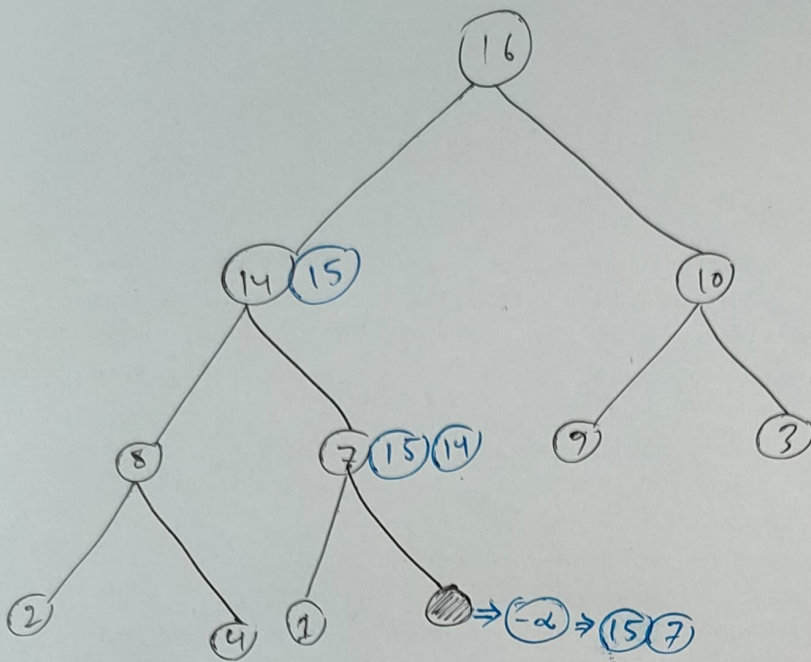
$\theta(1)$ {

- if $A.\text{heap-size} == n$
error "heap overflow"
- $A.\text{heap-size} = A.\text{heap-size} + 1$
- $k = x.\text{key}$
- $x.\text{key} = -\infty$
- $A[A.\text{heap-size}] = x$

$O(n)$ ← map x to index heap-size in the array

$O(\lg n)$ ← MAX-HEAP-INCREASE-KEY (A, n, k)

Total runtime = $O(\lg n)$



Quiz- Question

- pseudocode with some changes
- Draw the operation set
- Time analysis
- comparison between two algorithm