

Heaps: An array $A[1..n]$ is called a heap

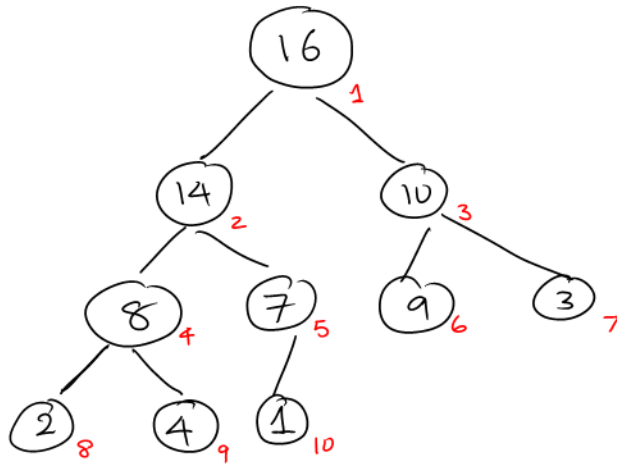
if $\forall i \geq 1$: $A[i] \geq A[2i]$ (if $2i \leq n$)
and $A[i] \geq A[2i+1]$ (if $2i+1 \leq n$).

Example $A =$

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

is a heap and it can be visualized
as a Binary Tree



root of tree: $A[1]$

Node with index i :

— parent of i has index $\lfloor i/2 \rfloor$.

$$\text{parent}(i) := \lfloor i/2 \rfloor$$

— left child of i is at index $2i$ (if $2i \leq n$)

$$\text{left}(i) = 2i$$

— right child of i is at index $2i+1$
(if $2i+1 \leq n$).

$$\text{Right}(i) = 2i+1.$$

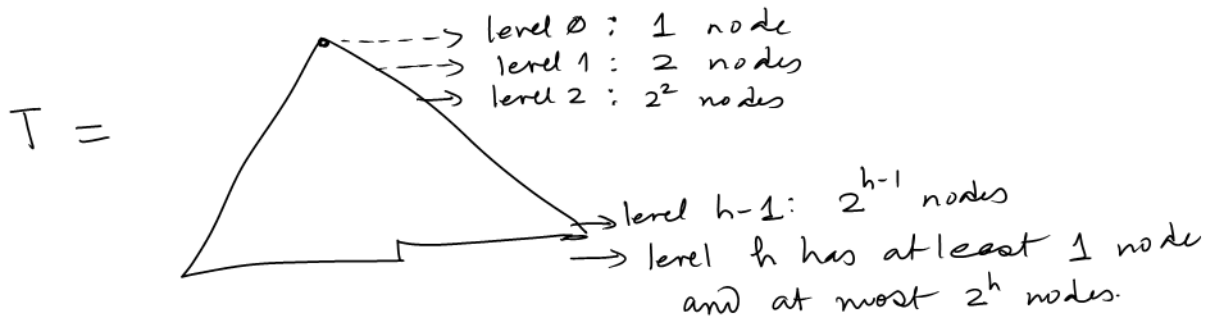
Note that $A[1..n]$ is a heap if for all $1 < i \leq n$
 $A[\text{parent}(i)] \geq A[i]$.

Q: What is the height of the heap?

$\equiv$ What is the height of the tree?

$\equiv$ # of edges on the longest root-to-leaf
path

Define $h :=$ height of the Binary Tree T .



$$\Rightarrow 1 + 2 + 2^2 + \dots + 2^{h-1} + 1 \leq n \leq 1 + 2 + 2^2 + \dots + 2^{h-1} + 2^h$$

$$\text{or } (2^h - 1) + 1 \leq n \leq 2^{h+1} - 1$$

$$\text{or } 2^h \leq n < 2^{h+1}$$

$$\Rightarrow h = \lfloor \log_2 n \rfloor.$$

Hence the height of heap h is $O(\log n)$.

Operations on a Heap

Assume we have a heap $A[1..N]$

an integer n , $1 \leq n \leq N$

Sequence of n numbers stored in

$A[1..n]$ satisfying the

heap property: $\forall i: 1 \leq i \leq n$

$$A[\text{parent}(i)] \geq A[i].$$

#1: MAXIMUM(A): Return ($A[1]$)

return the largest element.

$$\text{Time} = O(1).$$

#2: Increase-key (A, i, x): assumes that $x \geq A[i]$, $1 \leq i \leq n$ and $A[1..n]$ is a heap.

increases $A[i]$ to x and restores heap property

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A[i] := x;
while i > 1 and A[parent(i)] < A[i] do
    SWAP (A[parent(i)], A[i]);
    i := parent(i);

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$$\text{TIME} = O(\text{height}) = O(\log n).$$

③ Insert (A, x) : Assumes that $A[1..n]$ is a heap and $n < N$.
 Inserts element x and restore the heap property.

$$\begin{cases} n := n+1; \\ A[n] := x; \\ \text{(note that } A[1..n] \text{ is a heap)} \\ \text{Increase-key}(A, n, x) \end{cases}$$

$$\begin{aligned} \text{TIME} &\equiv O(1) + \text{time for increase key} \\ &= O(1) + O(\log n) \\ &= O(\log n) \end{aligned}$$

4. First method for constructing a heap on n -elements.

Input: An array $A[1..N]$ on elements

Output: Permute elements of A so that $A[1..N]$ is a heap.

$$\begin{cases} n := 1; \text{ (note that } A[1..n] \text{ is a heap)} \\ \text{while } n < N \text{ do} \\ \quad [\text{insert}(A, A[n+1])]; \end{cases}$$

Insert increments n .
 Also $A[1..n]$ is a heap.

Running Time: Cost for insert is sum total of cost for inserting the 1st element + cost of inserting 2nd element + cost of inserting 3rd element + ... + cost of inserting i th element + ... + cost of inserting N th element.

$$\sim \log 1 + \log 2 + \log 3 + \dots + \log N$$

$$\equiv \Theta(N \log N). \quad [\text{See Assignment 1}].$$

5. Heapify (A, i)

Assumptions: Binary tree rooted at Left(i) and Right(i) are max-heaps, but $A[i]$ may be smaller than its children.

Max-heap property may be violated at the node i .

Objective: Restore the heap property.

The heap property can be restored as follows:

1. Compute $\text{MAX}(A[i], A[\text{left}(i)], A[\text{right}(i)])$;
(If $A[i]$ is largest, then heap property is satisfied)
2. If $A[i]$ is not the MAX, then
 $\text{SWAP}(A[i] \leftrightarrow \text{MAX}(A[\text{left}(i)], A[\text{right}(i)]))$
3. Note that heap property is satisfied at $A[i]$, but one of its children may not satisfy the property.
4. Recurse on the children, which do not satisfy the heap property.

TIME $\equiv O(\text{Height of the node } i)$

#6: An $O(N)$ algorithm for BUILDING Heap.

Let $A[1..N]$ be the array which needs to be permuted, so that A satisfies the heap property.

Note that $A[\frac{N}{2}+1, \frac{N}{2}+2, \dots, N]$ satisfy heap property (since they correspond to leaves).

[BUILD-MAX-HEAP(A)
For $i := \lfloor \frac{N}{2} \rfloor$ downto 1 do MAX-Heapify(A, i).

Cost of BUILDING THE Heap

$$= \sum_{h=0}^{\lfloor \log N \rfloor} \left(\frac{N}{2^{h+1}} \right) O(h) \rightarrow \text{Time for heapify for a node at height } h$$

$\xrightarrow{\quad} \# \text{ of nodes at height } h.$

$$= O(N)$$

Note $\sum_{h=0}^{\log N} \frac{h}{2^h} \leq \sum_{h=0}^{\infty} \frac{h}{2^h} = \sum_{h=0}^{\infty} h \left(\frac{1}{2} \right)^h = \frac{1/2}{(1-1/2)^2} = 2$

Note for $|x| < 1$
 $\sum_{k=0}^{\infty} x^k = \frac{1}{1-x}$

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

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

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

Set
 $x = 1/2$
 $k = h.$

#7 HeapSort:

Inplace sorting algorithm, doesn't require extra memory.

Input: An unsorted array A.

Output: Array A in sorted order.

HeapSort(A)

1. BUILD-MAX-Heap(A). $\text{Heapsize} := n;$

2. For $i := n$ down to 2 do

 Exchange $A[1] \leftrightarrow A[i];$

$\text{Heapsize} := \text{Heapsize} - 1;$

 Heapify(A, 1); $\text{-----} \rightarrow O(\log N) / \text{per call.}$

Correctness is obvious since root always contains the maximum value.

Complexity: Step 1 takes $O(N)$ time.

Step 2 requires at most $\log N$ time per call,
or in all $O(N \log N)$ time.

#8 Extract-Max(A): Delete the largest element and restore the heap property.

Step 1: Return $(A[1])$ [$\text{max} := A[1]$]

Step 2: $A[1] := A[\text{Heap-size}(A)]$

Step 3: $\text{Heap-size}(A) := \text{Heap-size}(A) - 1;$

Step 4: Heapify(A, 1)

Heap-size denotes the number of elements in the heap.

TIME = $O(\log n)$

Note that $A[1.. \text{Heap-size}(A)]$ is a valid Heap.

New Heap has one fewer element than the old heap.