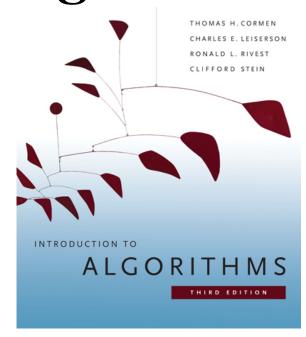
6.006- Introduction to Algorithms



Lecture 9

Prof. Constantinos Daskalakis

CLRS: 2.1, 2.2, 2.3, 6.1, 6.2, 6.3 and 6.4.

Lecture Overview

Priority Queues

Heaps

Heapsort

Priority Queue

This is an *abstract datatype* implementing a set *S* of elements, each associated with a key, supporting the following operations:

insert(S, x): insert element x into set S

 $\max(S)$: return element of S with largest key

extract $\max(S)$: return element of S with largest key and

remove it from S

increase_key(S, x, k): change the key-value of element x to the

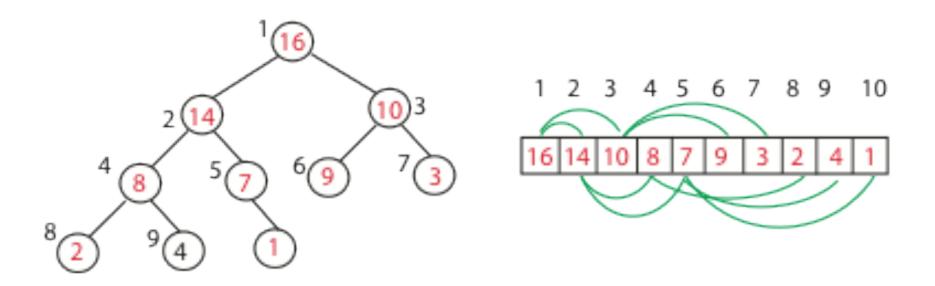
value k (assumed to be as large as current

value)

Heap

An implementation of a priority queue. It is an array object, visualized as a nearly complete binary tree.

Heap Property: The key of a node is \geq than the keys of its children.



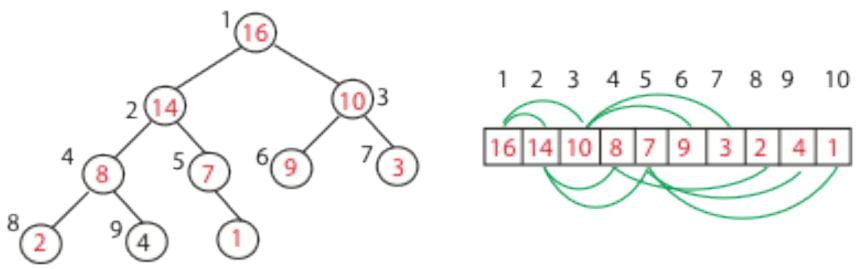
Visualizing an Array as a Tree

root of tree: first element in the array, corresponding to index = 1

If a node's index is i then:

parent(i) =
$$\left\lfloor \frac{i}{2} \right\rfloor$$
; returns index of node's parent, e.g. parent(5)=2

left(i) = 2i; returns index of node's left child, e.g. left(4)=8 right(i) = 2i + 1; returns index of node's right child, e.g. right(4)=9



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Note: no pointers required! Height of a binary heap $O(\log_2 n)$.

Heap-Size Variable

For flexibility we may only need to consider the first few elements of an array as part of the heap.

The variable heap-size denotes the number of items of the array that are part of the heap:

Max-Heaps vs Min-Heaps

Max Heaps satisfy the Max-Heap Property

for all i, $A[i] \ge \max\{A[left(i)], A[right(i)]\}$

Min Heaps satisfy the Min-Heap Property

for all i, $A[i] \le \min\{A[left(i)], A[right(i)]\}$

Operations with Heaps

build max heap: produce a max-heap from an unordered

array in O(n);

max heapify: correct a single violation of the heap property

occurring at the root of a subtree in $O(\log n)$;

insert, extract_max : O(log n)

heapsort: sort an array of size n in $O(n \log n)$ using heaps

Max_heapify

Max_heapify

correct a single violation of the heap property occurring at the root of a subtree in $O(\log n)$;

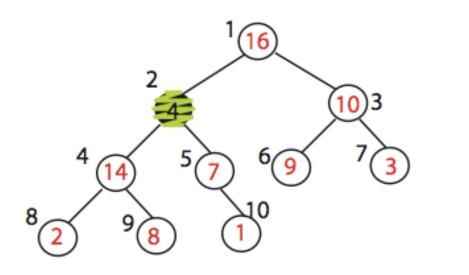
Assume that the trees rooted at left(i) and right(i) are max-heaps, but element A[i] violates the max-heap property;

i.e. A[i] is smaller than at least one of A[left(i)] or A[right(i)].

The goal is to correct the violation.

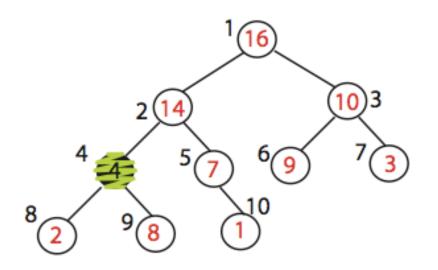
Do this by trickling element A[i] down the tree, making the subtree rooted at index i a max-heap.

Max_heapify (Example)



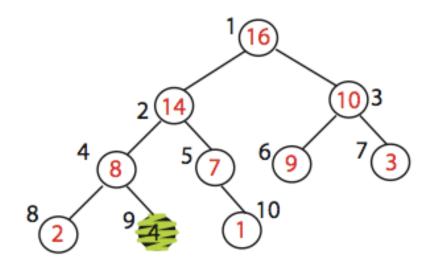
 $MAX_HEAPIFY (A,2)$ heap_size[A] = 10

Max_heapify (Example)



Exchange A[2] with A[4]
Call MAX_HEAPIFY(A,4)
because max_heap property
is violated

Max_heapify (Example)



Exchange A[4] with A[9] No more calls

Max_heapify (Pseudocode)

$Max_heapify(A, i)$

Find the index of the largest element among A[i], A[left(i)] and A[right(i)]

If this index is different than i, exchange A[i] with largest element; then recurse on subtree

```
r \leftarrow \operatorname{left}(i)

r \leftarrow \operatorname{right}(i)

if l \leq \operatorname{heap-size}(A) and A[l] > A[i]

then largest \leftarrow l

else largest \leftarrow i

if r \leq \operatorname{heap-size}(A) and A[r] > A[\operatorname{largest}]

then largest \leftarrow r

if largest \neq i

then exchange A[i] and A[\operatorname{largest}]

\operatorname{MAX-HEAPIFY}(A, \operatorname{largest})
```

IMPORTANT NOTE: If element A[i] is smaller than both A[left(i)] and A[right(i)], I insist on swapping it with the largest of the two and not with either one of them, arbitrarily.

Build_Max_heap

Build_Max_Heap(A)

Convert A[1...n] to a max heap.

Observation: Elements $A[\lfloor n/2 \rfloor + 1 \dots n]$ are leaves of the tree because 2i > n, for all $i \ge \lfloor n/2 \rfloor + 1$

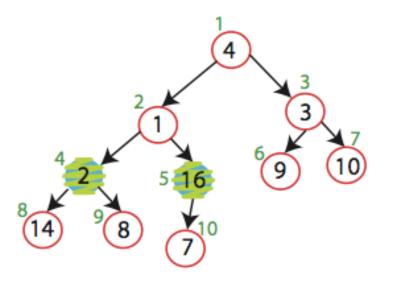
so heap property may only be violated at nodes $1...\lfloor n/2 \rfloor$ of the tree

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Build_Max_Heap(A):

heap\_size(A) = length(A)

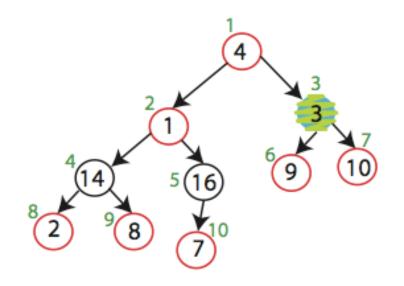
for i \leftarrow \lfloor length[A]/2 \rfloor downto 1

do Max_Heapify(A, i)
```

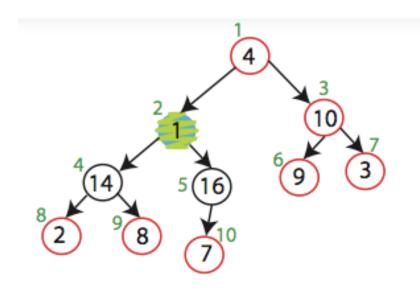


A 4 1 3 2 16 9 10 14 8 7

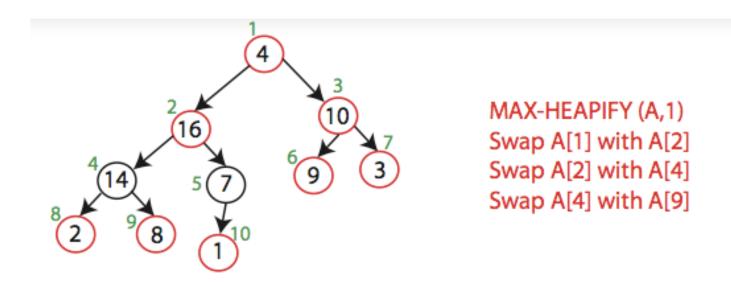
MAX-HEAPIFY (A,5) no change MAX-HEAPIFY (A,4) Swap A[4] and A[8]



MAX-HEAPIFY (A,3) Swap A[3] and A[7]



MAX-HEAPIFY (A,2) Swap A[2] and A[5] Swap A[5] and A[10]



Running Time: Trivially $O(n \log n)$, since I need to Heapify O(n) times.

Observe, however, that Heapify only pays O(1) time for the nodes that are one level above the leaves, and in general $O(\ell)$ for the nodes that are ℓ levels above the leaves. O(n) time overall!

Heapsort

Recall Naïve Algorihm..

Sorting Strategy:

Find largest element of array, place it in last position; then find the largest among the remaining elements, and place it next to the largest, etc...

In notation:

- 1. $last_element = n;$
- 2. Find maximum element A[i] of array A[1...last_element];
- 3. Swap A[i] and A[last_element];
- 4. last_element = last_element 1;
- 5. Go to step 2

We have a fast data structure for step 2! (which is also the most costly)

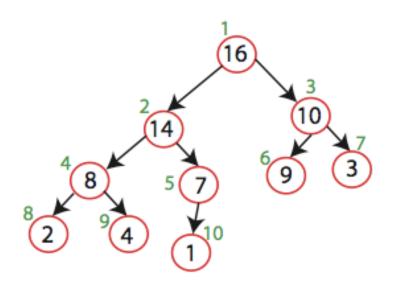


Sorting Strategy:

1. Build Max Heap from unordered array;

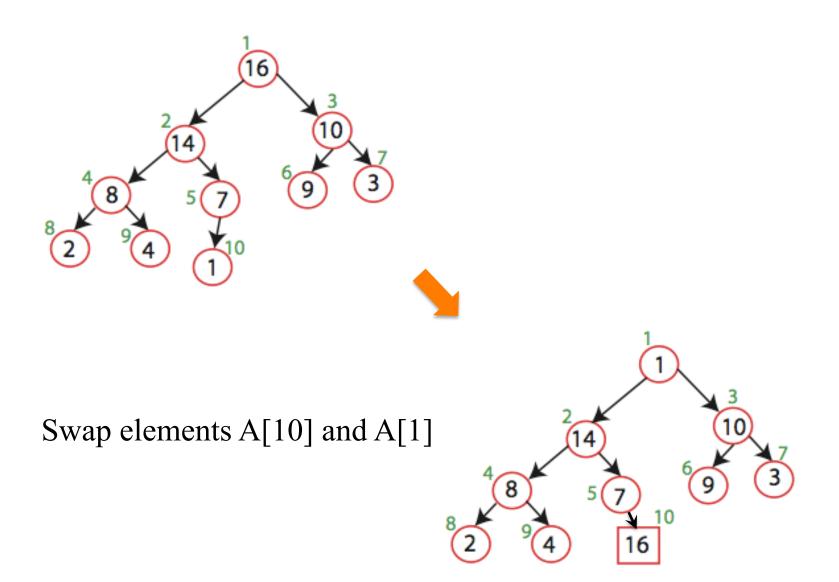
A 4 1 3 2 16 9 10 14 8 7





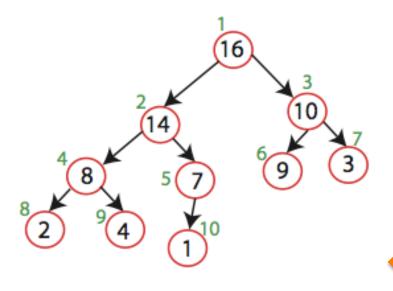
Sorting Strategy:

- 1. Build Max Heap from unordered array;
- 2. Find maximum element; this is A[1];
- 3. Swap elements A[n] and A[1]: now max element is at the end of the array!

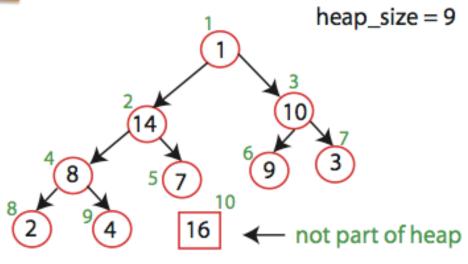


Sorting Strategy:

- 1. Build Max Heap from unordered array;
- 2. Find maximum element A[1];
- 3. Swap elements A[n] and A[1]: now max element is at the end of the array!
- 4. Discard node *n* from heap (by decrementing heap-size variable)

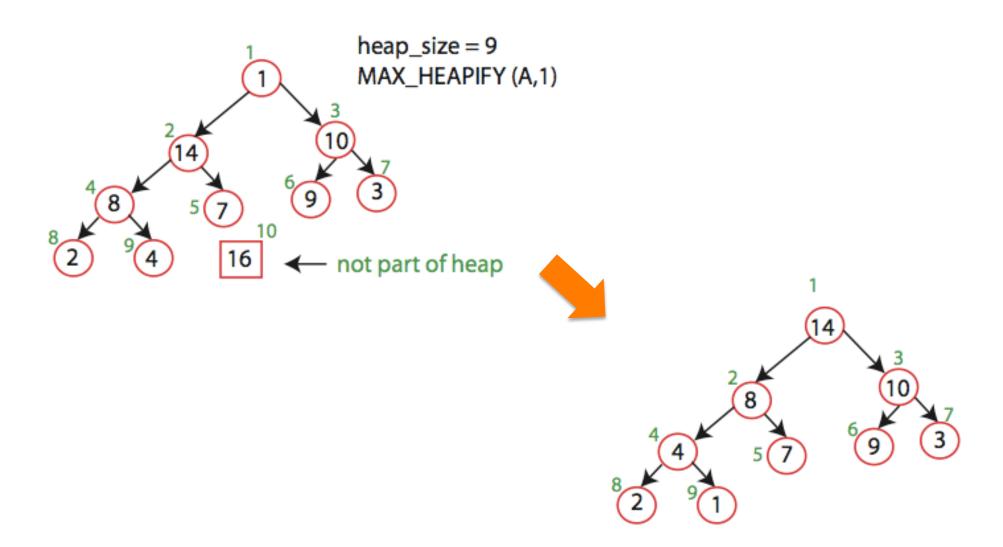


Swap elements A[10] and A[1] heap_size = heap_size-1



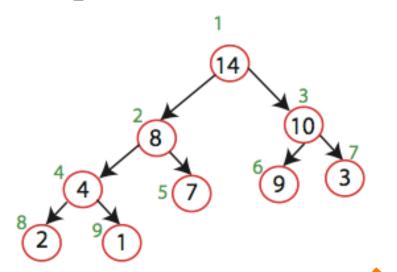
Sorting Strategy:

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- 5. New root may violate max heap property, but its children are max heaps. Run max_heapify to fix this.

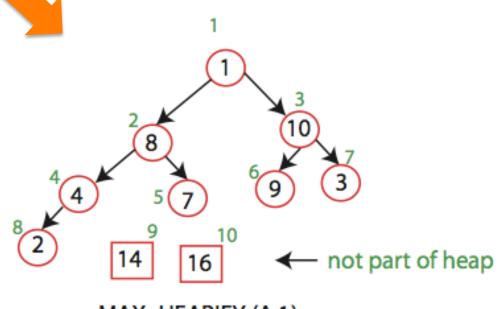


Sorting Strategy:

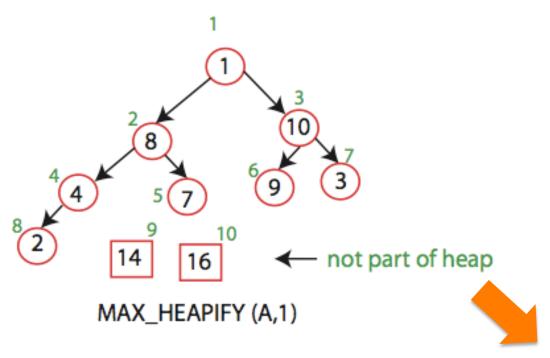
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- 6. Go to step 2.



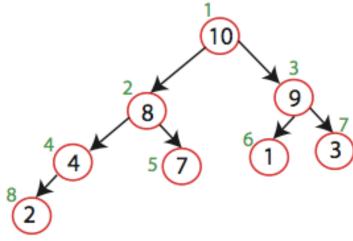
Swap elements A[9] and A[1]

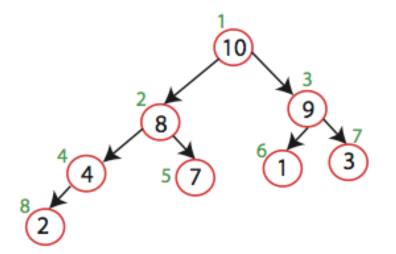


MAX_HEAPIFY (A,1)

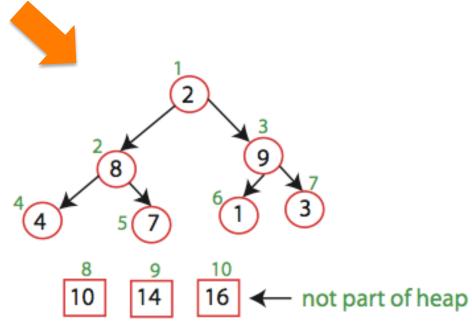


Max_Heapify(A,1)





Swap elements A[8] and A[1]



and so on...

Running time:

after n iterations the Heap is empty every iteration involves a swap and a heapify operation; hence it takes $O(\log n)$ time

Overall $O(n \log n)$

Discussion: Other operations?

Operations with Heaps

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array in O(n);

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