

Algorithm and Data Structures

Lecture notes: Heapsort, Cormen Chap. 6

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Outline

Introduction

Heaps

Operations on heaps

- Heapify

- Buildheap

Heapsort

Appendix : Priority queues

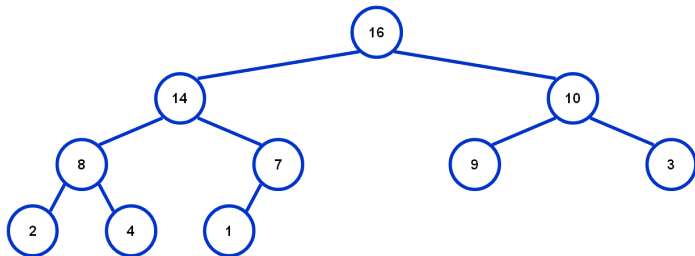
Exercises

Sorting

- ▶ So far we have seen different sorting algorithms such as selection sort, and insertion, and merge sort and quicksort
 - ▶ Merge sort runs in $O(n \log n)$ both in best, average and worst-case
 - ▶ Insertion/selection sort run in $O(n^2)$, but insertion sort is fast when array is nearly sorted, runs fast in practice
- ▶ Next on the agenda : Heapsort
- ▶ Prior to describe heapsort, we introduce the heap data structure and operations on that data structure

Heap : definition

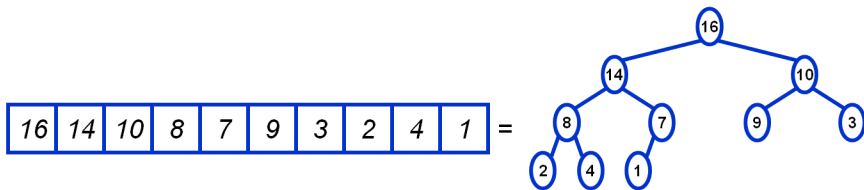
- ▶ A heap is a complete binary tree



- ▶ *Binary* because each node has at most two children
- ▶ *Complete* because each internal node, except possibly at the last level, has exactly two children

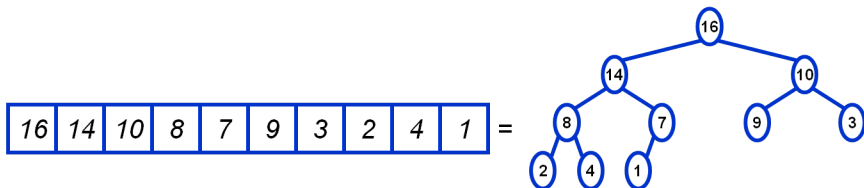
Heaps

- In practice, heaps are usually implemented as arrays



Heaps

- ▶ How to represent a complete binary tree as an array :
 - ▶ The root node is $A[1]$
 - ▶ Ordering nodes per levels starting at the root, and from left to right in a same level, then node i is $A[i]$
 - ▶ The parent of node i is $A[\lfloor i/2 \rfloor]$
 - ▶ The left child of node i is $A[2i]$
 - ▶ The right child of node i is $A[2i + 1]$



Referencing Heap Elements

```
function Parent(i)  
    return  $\lfloor i/2 \rfloor$  ;
```

```
function Left(i)  
    return  $2 \times i$  ;
```

```
function Right(i)  
    return  $2 \times i + 1$  ;
```

The Heap Property

- ▶ Heaps must satisfy the following relation :

$$A[\text{Parent}(i)] \geq A[i] \text{ for all nodes } i > 1$$

- ▶ In other words, the value of a node is at most the value of its parent

Heap Height

- ▶ The height of a node in the tree = the number of edges on the longest downward path to a leaf
- ▶ The height of a tree = the height of its root
- ▶ What is the height of an n -element heap? Why?
- ▶ Heap operations take at most time proportional to the height of the heap

Heap Operations

There are two main heap operations : heapify and buildheap.

Heapify() : restore the heap property :

- ▶ Consider node i in the heap with children l and r
- ▶ Nodes l and r are each the root of a subtree, each assumed to be a heap
- ▶ Problem : Node i may violate the heap property
- ▶ Solution : let the value of node i "float down" in one of its two subtrees until the heap property is restored at node i

Algorithm for heapify

An array $A[]$, where $\text{heap_size}(A)$ returns the dimension of A

Heapify(A, i)

$l = \text{Left}(i)$; $r = \text{Right}(i)$;

 if ($l \leq \text{heap_size}(A)$ & $A[l] > A[i]$)

$\text{largest} = l$;

 else

$\text{largest} = i$;

 if ($r \leq \text{heap_size}(A)$ & $A[r] > A[\text{largest}]$)

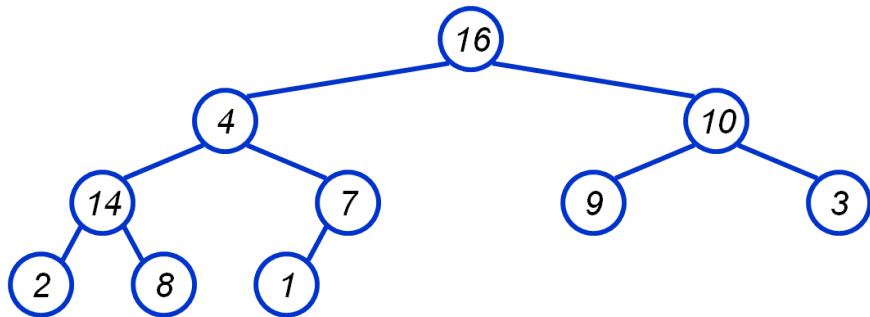
$\text{largest} = r$;

 if ($\text{largest} \neq i$)

 Swap($A, i, \text{largest}$) ;

 Heapify($A, \text{largest}$) ;

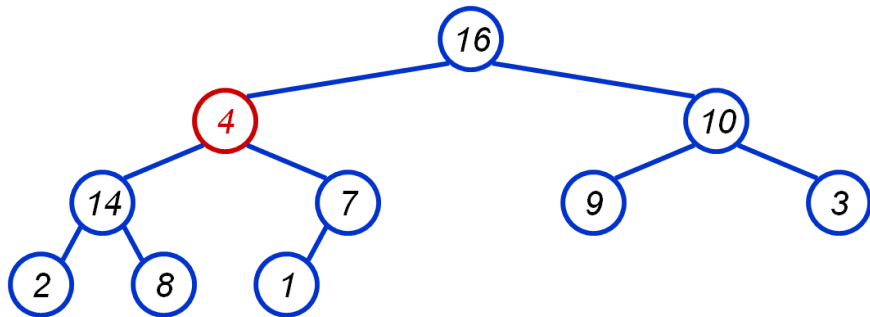
Heapify() Example



$A =$

16	4	10	14	7	9	3	2	8	1
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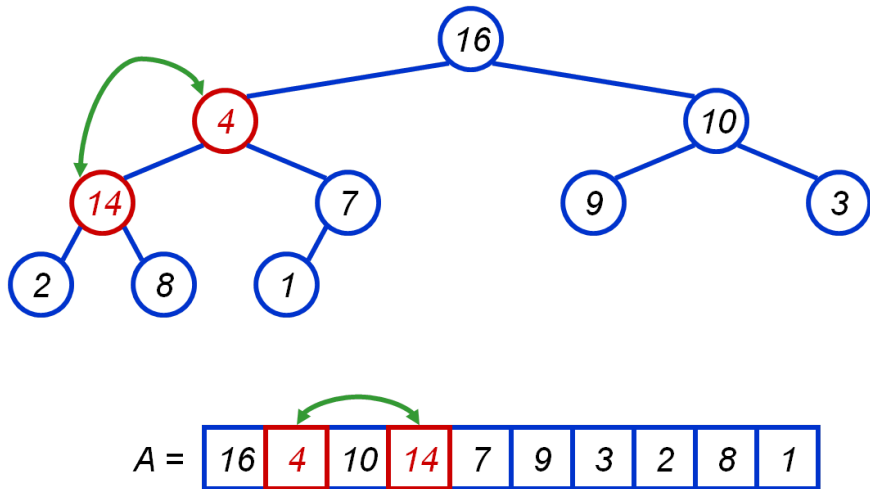
Heapify() Example



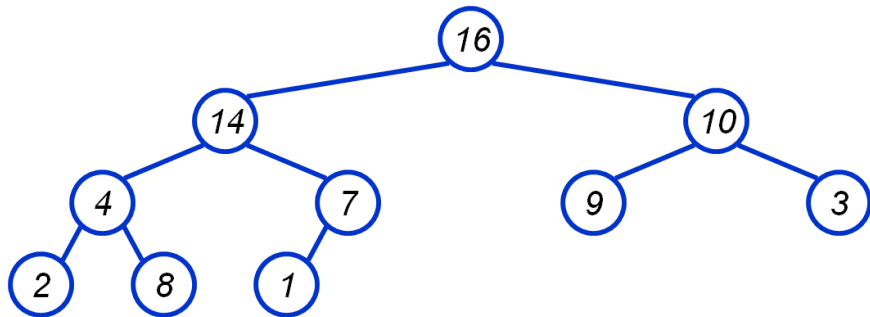
$A =$

16	4	10	14	7	9	3	2	8	1
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Heapify() Example



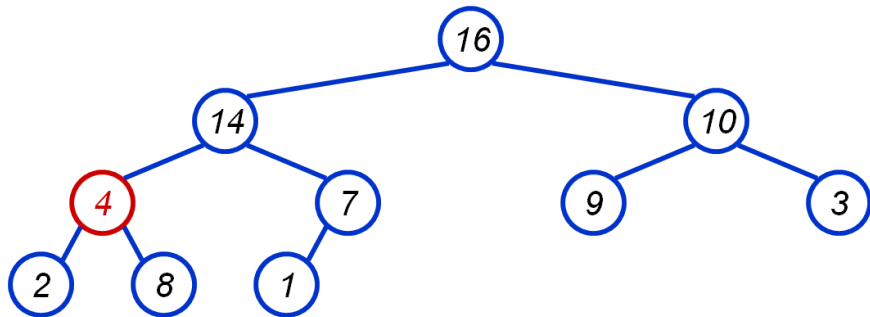
Heapify() Example



$A =$

16	14	10	4	7	9	3	2	8	1
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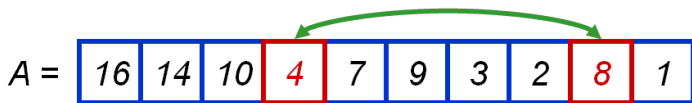
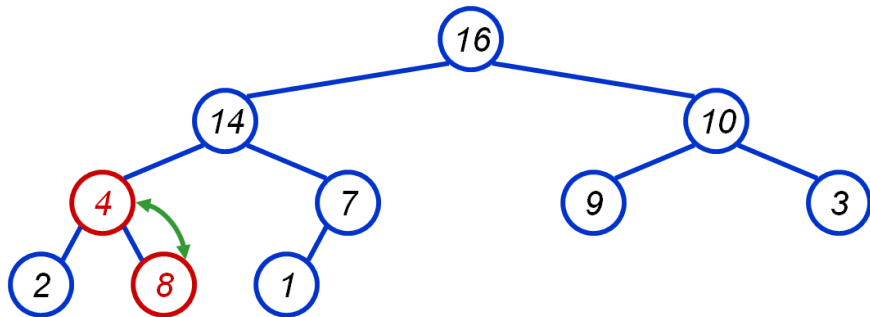
Heapify() Example



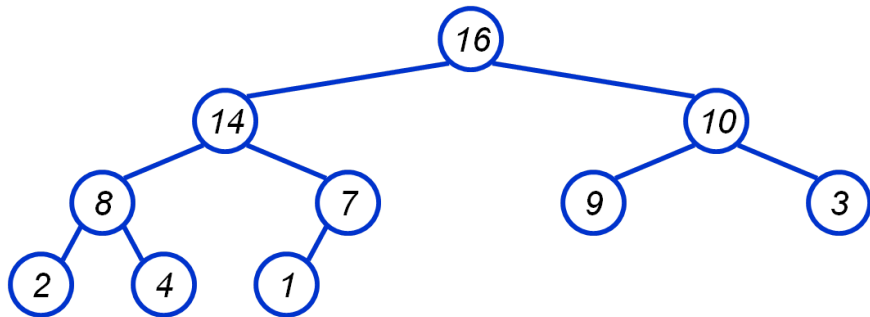
$A =$

16	14	10	4	7	9	3	2	8	1
----	----	----	---	---	---	---	---	---	---

Heapify() Example



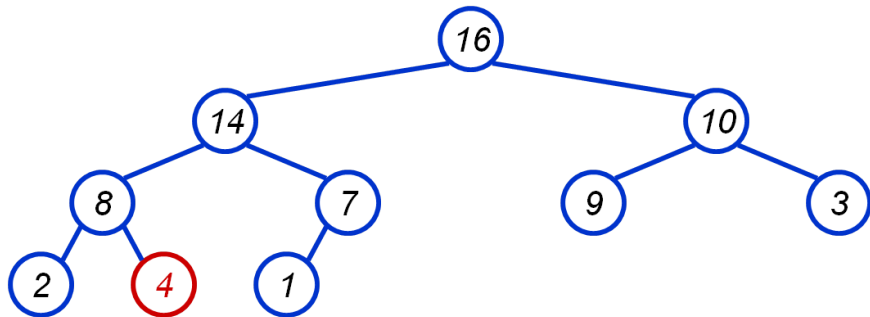
Heapify() Example



$A =$

16	14	10	8	7	9	3	2	4	1
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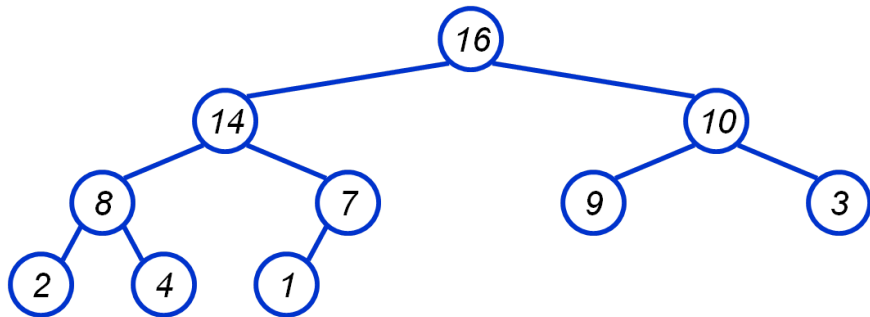
Heapify() Example



$A =$

16	14	10	8	7	9	3	2	4	1
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Heapify() Example



$A =$

16	14	10	8	7	9	3	2	4	1
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Analyzing Heapify()

- ▶ Number of basic operations performed before calling itself?
- ▶ How many times can Heapify() recursively call itself?
- ▶ What is the worst-case running time of Heapify() on a heap of size n ?

Analyzing Heapify()

- ▶ The work done in Heapify() is in $O(1)$
- ▶ If the heap at i has n elements, how many elements can the subtrees at l or r have? Answer : at most $2n/3$ (worst case : bottom row $1/2$ full)
- ▶ So time taken by Heapify() is given by the recurrence

$$T(n) \leq T(2n/3) + \Theta(1)$$

- ▶ Master Theorem applies to solve this recurrence, which corresponds to the case 2 of the restricted Master Theorem.

$$T(n) \in \Theta(\log n)$$

Heap Operations : BuildHeap()

- ▶ We can build a heap in a bottom-up manner by running Heapify() on successive subarrays
 - ▶ Note : for array of length n , all elements in range $A[n/2 + 1..n]$ are heaps (Why?)
 - ▶ Walk backwards through the array from $n/2$ to 1, calling Heapify() on each node.
- ▶ given an unsorted array A, make A a heap

BuildHeap(A)

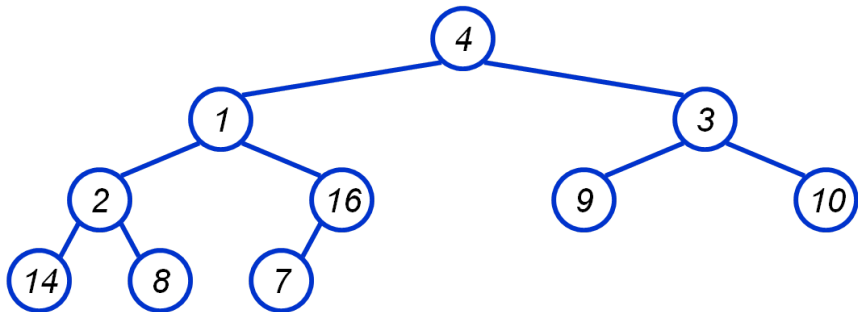
 heap_size(A) = length(A);

 for ($i = \lfloor \text{length}(A)/2 \rfloor$ downto 1)

 Heapify(A, i);

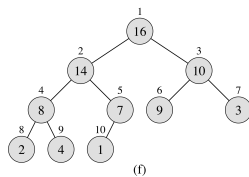
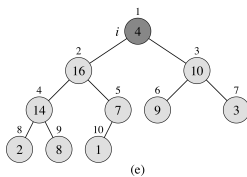
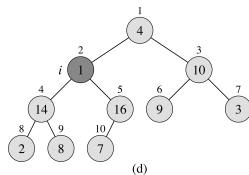
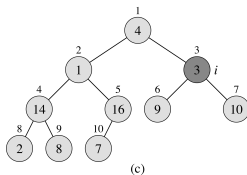
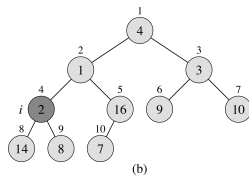
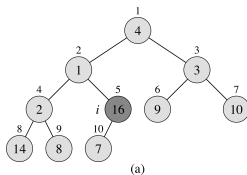
BuildHeap() Example

Work through example $A = [4, 1, 3, 2, 16, 9, 10, 14, 8, 7]$



A

4	1	3	2	16	9	10	14	8	7
---	---	---	---	----	---	----	----	---	---



Analyzing BuildHeap()

- ▶ Each call to Heapify() takes $O(\log n)$ time
- ▶ There are $O(n)$ such calls (specifically, $\lfloor n/2 \rfloor$)
- ▶ Thus the running time is $O(n \log n)$
 - ▶ Is this a correct asymptotic upper bound?
 - ▶ Is this an asymptotically tight bound?
- ▶ A tighter bound is $O(n)$

Analyzing BuildHeap() : Tight

Heap-properties of an n -element heap

- ▶ Height = $\lfloor \log n \rfloor$
- ▶ At most $\lceil \frac{n}{2^{h+1}} \rceil$ nodes of any height h
- ▶ The time for Heapify on a node of height h is $O(h)$

$$\begin{aligned} \sum_{h=0}^{\lfloor \log n \rfloor} \lceil \frac{n}{2^{h+1}} \rceil O(h) &= O(n \sum_{h=0}^{\lfloor \log n \rfloor} \frac{h}{2^h}) \\ &= O(n \sum_{h=0}^{\infty} \frac{h}{2^h}) \\ &= O(n) \end{aligned}$$

Analyzing BuildHeap() : Tight

$$\begin{aligned}\sum_{h=0}^{\infty} \frac{h}{2^h} &= \sum_{h=0}^{\infty} h \left(\frac{1}{2}\right)^h \\ &= \sum_{h=0}^{\infty} h x^h \text{ where } x = \frac{1}{2} \\ &= \frac{1/2}{(1 - \frac{1}{2})^2} \text{ the closed form of } \sum_{h=0}^{\infty} h \left(\frac{1}{2}\right)^h \\ &= 2\end{aligned}$$

Heapsort

- ▶ Given BuildHeap(), a sorting algorithm is easily constructed :
 - ▶ Maximum element is at $A[1]$
 - ▶ Swap $A[1]$ with element at $A[n]$, $A[n]$ now contains correct value
 - ▶ Decrement heap_size[A]
 - ▶ Restore heap property at $A[1]$ by calling Heapify()
 - ▶ Repeat, always swapping $A[1]$ for $A[\text{heap_size}(A)]$

Heapsort(A)

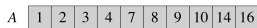
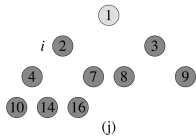
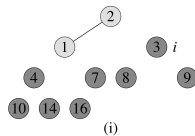
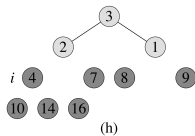
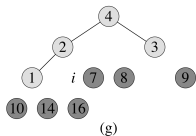
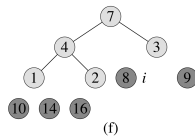
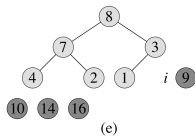
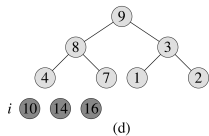
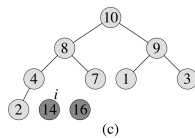
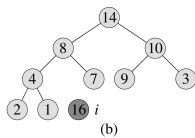
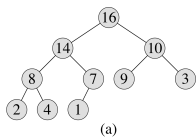
 BuildHeap(A);

 for ($i = \text{length}(A)$ downto 2)

 Swap($A[1]$, $A[i]$);

 heap_size(A) = heap_size(A) - 1;

 Heapify(A, 1);



(k)

Analyzing Heapsort

- ▶ The call to BuildHeap() takes $O(n)$ time
- ▶ Each of the $n - 1$ calls to Heapify() takes $O(\log n)$ time
- ▶ Thus the total time taken by HeapSort()

$$= O(n) + (n - 1)O(\log n)$$

$$= O(n) + O(n \log n)$$

$$= O(n \log n)$$

Note, like merge sort, the running time of heapsort is independent of the initial state of the array to be sorted. So best case and average case of heapsort are in $O(n \log n)$

Priority Queues

- ▶ Heapsort is a nice algorithm, but in practice Quicksort is faster
- ▶ But the heap data structure is useful for implementing **priority queues** :
 - ▶ A data structure like queue or stack, but where a value or key is associated to each element, representing the priority of the corresponding element. The element with the highest priority is served first
 - ▶ Supports the operations Insert(), Maximum(), and ExtractMax()

Priority Queue Operations

- ▶ **function** $\text{Insert}(S, x)$ inserts the element x into set S
- ▶ **function** $\text{Maximum}(S)$ returns the element of S with the maximum key
- ▶ **function** $\text{ExtractMax}(S)$ removes and returns the element of S with the maximum key
- ▶ Think how to implement these operations using a heap?

Priority queue : extracting the max element

```
ExtractMax(A)
    max = A[1]
    A[1] = A[A.heap-size]
    A.heap-size = A.heap-size - 1
    Heapify(A,1)
    return max
```

Since Heapify runs in $\log n$, extracting the largest element of a priority queue based on a heap takes $\log n$

Priority queue : inserting an element

```
Insert(A, key)
  A.heap-size = A.heap-size + 1
  A[A.heap-size] = key
  i = A.heap-size
  while i > 1 and A[Parent(i)] < A[i]
    swap(A[i], A[Parent(i)])
    i = Parent(i)
```

The number of iterations execute by the while loop is bound above by $\log n$, therefore inserting an element of a priority queue based on a heap takes $\log n$

Exercises

1. Insertion sort and merge sort are stable algorithms while heapsort and quicksort are not. Can you explain why this is so?
2. Does this array $A = [23, 11, 14, 9, 13, 10, 1, 5, 7, 12]$ is a heap (max-heap)? If not makes it a heap.
3. Run $Heapify(A, 3)$ on the array $A = [27, 17, 3, 16, 13, 10, 1, 5, 7, 12, 4, 8, 9, 0]$
4. $Heapify(A, i)$ in the class notes is a recursive algorithm. Write an equivalent iterative algorithm.
5. Run the algorithm $BuildHeap(A)$ on the array $A = [8, 4, 27, 10, 72, 19, 9, 22, 6]$
6. Run $Heapsort(A)$ on the the array $A = [3, 15, 2, 29, 6, 14, 25, 7, 5]$
7. What is the running time of $Heapsort$ on an array A of length n that is sorted in decreasing order?
8. Show what happens when $Heap - Insert(A, 10)$ is run on the heap $A = [15, 13, 9, 5, 12, 8, 7, 4, 0, 6, 2, 1]$