

# CS 561, Lecture 4

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# Outline

- Loop Invariants
- Heaps

# Correctness of Algorithms

- The most important aspect of algorithms is their correctness
- An algorithm by definition *always* gives the right answer to the problem
- A procedure which doesn't always give the right answer is a *heuristic*
- All things being equal, we prefer an algorithm to a heuristic
- How do we prove an algorithm is really correct?

# Loop Invariants

- A useful tool for proving correctness is loop invariants.
- Loop Invariants are essentially proof by induction

# Loop Invariants

Three things must be shown about a loop invariant

- **Initialization:** Invariant is true before first iteration of loop (Base Case)
- **Maintenance:** If invariant is true before iteration  $i$ , it is also true before iteration  $i + 1$  (Inductive Step)
- **Termination:** When the loop terminates, the invariant gives a property which can be used to show the algorithm is correct (Wrapup)

## Example Loop Invariant

- We'll prove the correctness of a simple algorithm which solves the following interview question:
- *Find the middle of a linked list, while only going through the list once*
- The basic idea is to keep two pointers into the list, one of the pointers moves twice as fast as the other
- (Call the head of the list the 0-th elem, and the tail of the list the  $(n - 1)$ -st element, assume that  $n - 1$  is an even number)

# Example Algorithm

```
GetMiddle (List l){
    pSlow = pFast = l;
    while ((pFast->next)&&(pFast->next->next)){
        pFast = pFast->next->next
        pSlow = pSlow->next
    }
    return pSlow
}
```

## Example Loop Invariant

- *Invariant:* At the start of the  $i$ -th iteration of the while loop,  $pSlow$  points to the  $i$ -th element in the list and  $pFast$  points to the  $2i$ -th element
- **Initialization:** True when  $i = 0$  since both pointers are at the head
- **Maintenance:** if  $pSlow$ ,  $pFast$  are at positions  $i$  and  $2i$  respectively before  $i$ -th iteration, they will be at positions  $i + 1$ ,  $2(i + 1)$  respectively before the  $i + 1$ -st iteration
- **Termination:** When the loop terminates,  $pFast$  is at element  $n - 1$ . Then by the loop invariant,  $pSlow$  is at element  $(n - 1)/2$ . Thus  $pSlow$  points to the middle of the list



## Challenge

- Figure out how to use a similar idea to determine if there is a loop in a linked list *without marking nodes!*

# What is a Heap

- “A heap data structure is an array that can be viewed as a nearly complete binary tree”
- Each element of the array corresponds to a value stored at some node of the tree
- The tree is completely filled at all levels except for possibly the last which is filled from left to right

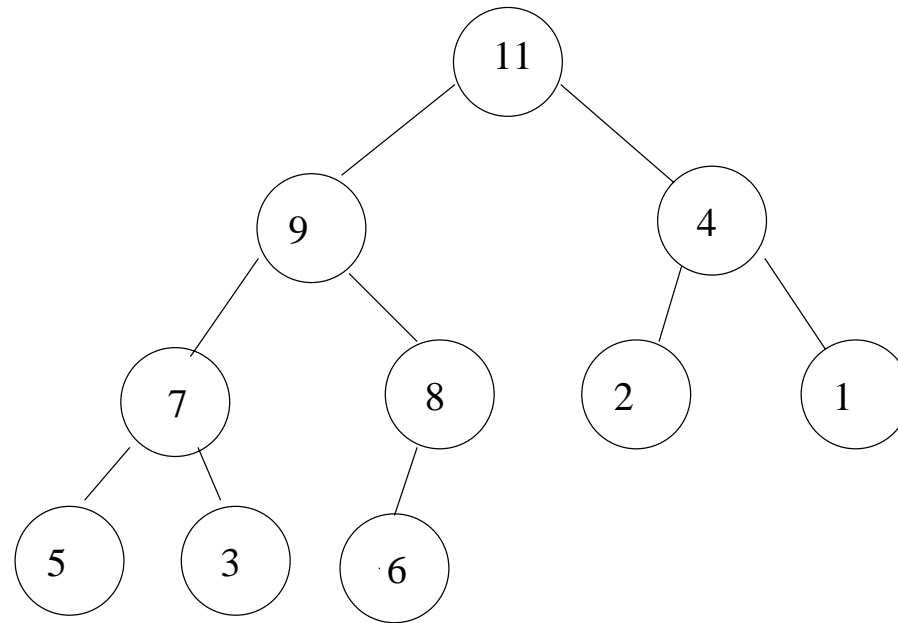
## — heap-size (A) —

- An array  $A$  that represents a heap has two attributes
  - length (A) which is the number of elements in the array
  - heap-size (A) which is the number of elems in the heap stored within the array
- I.e. only the elements in  $A[1..\text{heap-size}(A)]$  are elements of the heap

# Tree Structure

- $A[1]$  is the root of the tree
- For all  $i$ ,  $1 < i < \text{heap-size}(A)$ 
  - $\text{Parent}(i) = \lfloor i/2 \rfloor$
  - $\text{Left}(i) = 2i$
  - $\text{Right}(i) = 2i + 1$
- If  $\text{Left}(i) > \text{heap-size}(A)$ , there is no left child of  $i$
- If  $\text{Right}(i) > \text{heap-size}(A)$ , there is no right child of  $i$
- If  $\text{Parent}(i) < 0$ , there is no parent of  $i$

# Example



A:

1 2 3 4 5 6 7 8 9 10

11 9 4 7 8 2 1 5 3 6

## Max-Heap Property

- For every node  $i$  other than the root,  $A[\text{Parent}(i)] \geq A[i]$

# Max-Heap Property

- For every node  $i$  other than the root,  $A[\text{Parent}(i)] \geq A[i]$
- Parent is always at least as large as its children
- Largest element is at the root

(A Min-heap is organized the opposite way)

## Height of Heap

- Height of a node in a heap is the number of edges in the longest simple downward path from the node to a leaf
- Height of a heap of  $n$  elements is  $\Theta(\log n)$ . Why?



# Maintaining Heaps

- Q: How to maintain the heap property?
- A: *Max-Heapify* is given an array and an index  $i$ . Assumes that the binary trees rooted at  $Left(i)$  and  $Right(i)$  are max-heaps, but  $A[i]$  may be smaller than its children.
- *Max-Heapify* ensures that after its call, the subtree rooted at  $i$  is a Max-Heap

## Max-Heapify

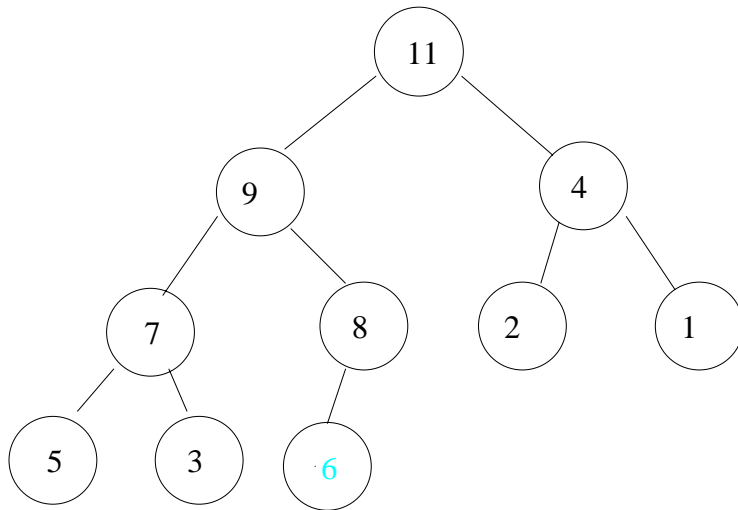
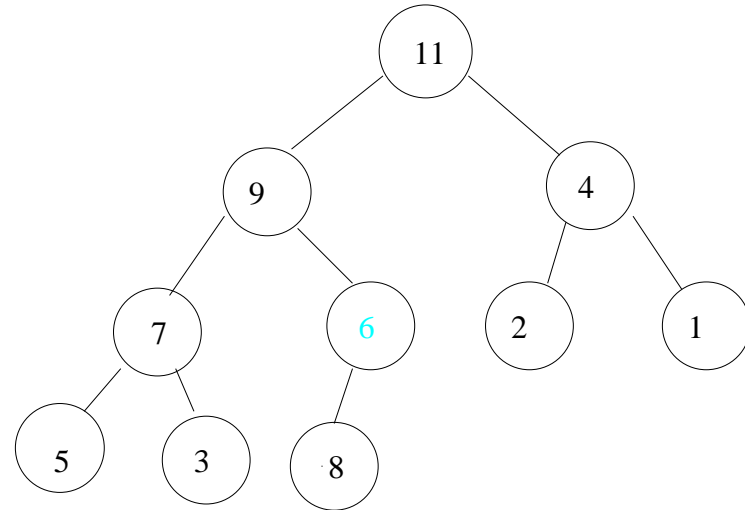
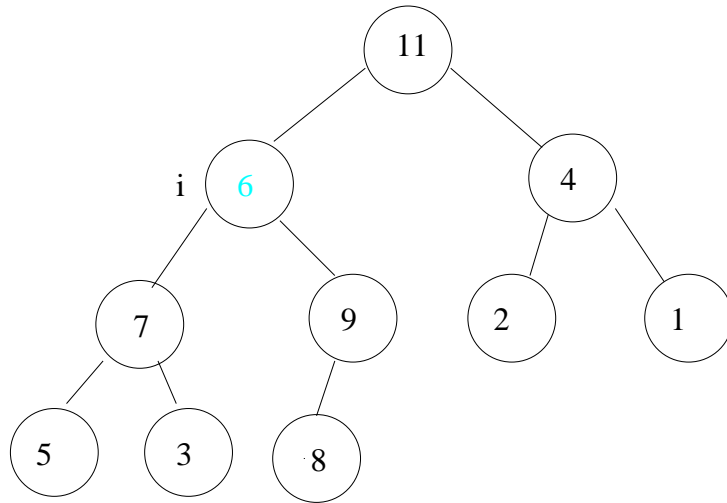
- Main idea of the Max-Heapify algorithm is that it percolates down the element that start at  $A[i]$  to the point where the subtree rooted at  $i$  is a max-heap
- To do this, it repeatedly swaps  $A[i]$  with its largest child until  $A[i]$  is bigger than both its children
- For simplicity, the algorithm is described recursively.

# Max-Heapify

Max-Heapify (A,i)

1.  $l = \text{Left}(i)$
2.  $r = \text{Right}(i)$
3.  $\text{largest} = i$
4. if ( $l \leq \text{heap-size}(A)$  and  $A[l] > A[i]$ ) then  $\text{largest} = l$
5. if ( $r \leq \text{heap-size}(A)$  and  $A[r] > A[\text{largest}]$ ) then  $\text{largest} = r$
6. if  $\text{largest} \neq i$  then
  - (a) exchange  $A[i]$  and  $A[\text{largest}]$
  - (b) Max-Heapify (A,largest)

# Example



## Analysis

- Let  $T(h)$  be the runtime of max-heapify on a subtree of height  $h$
- Then  $T(1) = \Theta(1)$ ,  $T(h) = T(h - 1) + 1$
- Solution to this recurrence is  $T(h) = \Theta(h)$
- Thus if we let  $T(n)$  be the runtime of max-heapify on a subtree of size  $n$ ,  $T(n) = O(\log n)$ , since  $\log n$  is the maximum height of heap of size  $n$

## Build-Max-Heap

- Q: How can we convert an arbitrary array into a max-heap?
- A: Use Max-Heapify in a bottom-up manner
- Note: The elements  $A[\lfloor n/2 \rfloor + 1], \dots, A[n]$  are all leaf nodes of the tree, so each is a 1 element heap to begin with

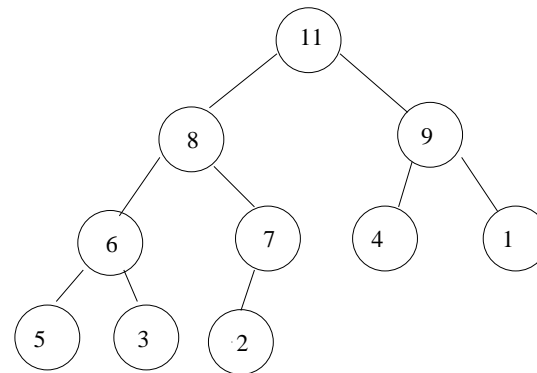
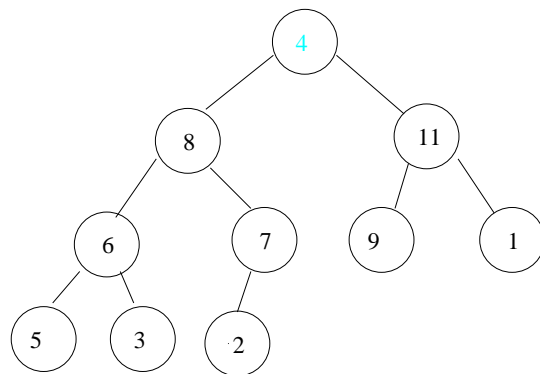
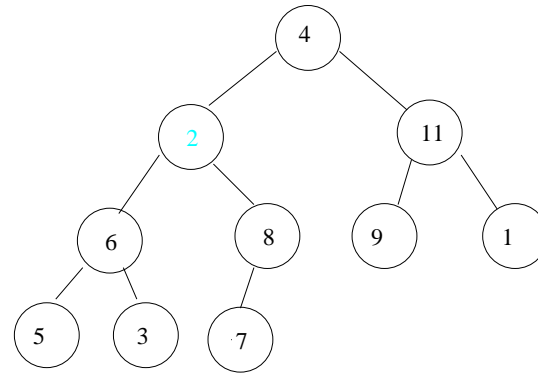
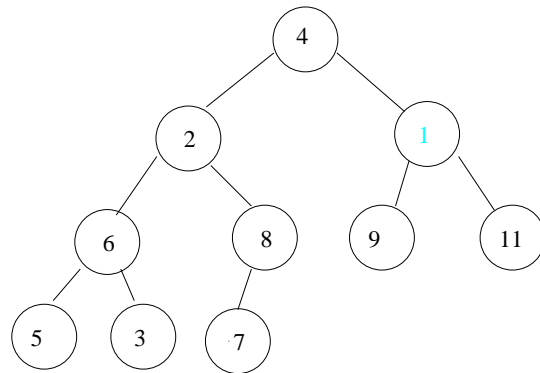
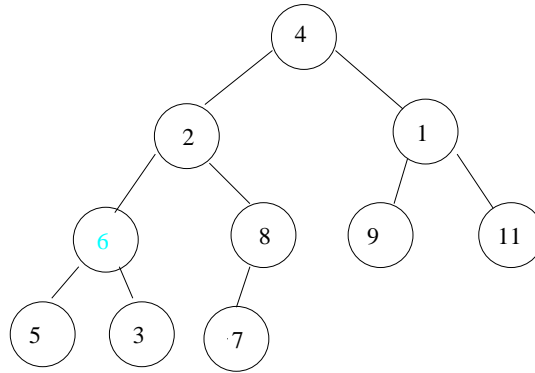
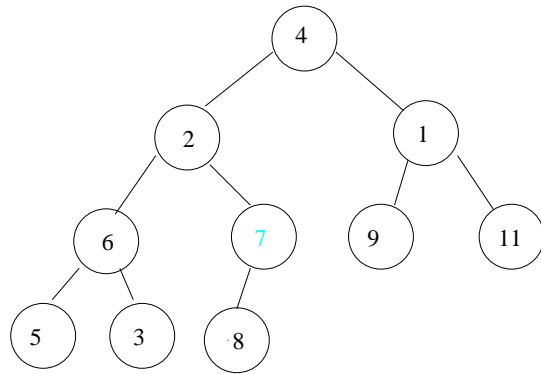
# Build-Max-Heap

Build-Max-Heap (A)

1. heap-size (A) = length (A)
2. for ( $i = \lfloor \text{length}(A)/2 \rfloor; i > 0; i --$ )
  - (a) do Max-Heapify (A,i)

# Example

A = 4 2 1 6 7 9 11 5 3 8





## Loop Invariant

- Loop Invariant: “At the start of each iteration of the for loop, each node  $i + 1, i + 2, \dots, n$  is the root of a max-heap”

## Correctness

- **Initialization:**  $i = \lfloor n/2 \rfloor$  prior to first iteration. But each node  $\lfloor n/2 \rfloor + 1, \lfloor n/2 \rfloor + 2, \dots, n$  is a leaf so is the root of a trivial max-heap
- **Termination:** At termination,  $i = 0$ , so each node  $1, \dots, n$  is the root of a max-heap. In particular, node 1 is the root of a max heap.

## Maintenance

- **Maintenance:** First note that if the nodes  $i + 1, \dots, n$  are the roots of max-heaps before the call to Max-Heapify ( $A, i$ ), then they will be the roots of max-heaps after the call. Further note that the children of node  $i$  are numbered higher than  $i$  and thus by the loop invariant are both roots of max heaps. Thus after the call to Max-Heapify ( $A, i$ ), the node  $i$  is the root of a max-heap. Hence, when we decrement  $i$  in the for loop, the loop invariant is established.

# Time Analysis

(Naive) Analysis:

- Max-Heapify takes  $O(\log n)$  time per call
- There are  $O(n)$  calls to Max-Heapify
- Thus, the running time is  $O(n \log n)$

# Time Analysis

Better Analysis. Note that:

- An  $n$  element heap has height no more than  $\log n$
- There are at most  $n/2^h$  nodes of any height  $h$  (to see this, consider the min number of nodes in a heap of height  $h$ )
- Time required by Max-Heapify when called on a node of height  $h$  is  $O(h)$ .
- Thus total time is:  $\sum_{h=0}^{\log n} \frac{n}{2^h} O(h)$

# Analysis

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

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

$$= O(n) \quad (3)$$

# Analysis

The last step follows since for all  $|x| < 1$ ,

$$\sum_{i=0}^{\infty} ix^i = \frac{x}{(1-x)^2} \quad (4)$$

Can get this equality by recalling that for all  $|x| < 1$ ,

$$\sum_{i=0}^{\infty} x^i = \frac{1}{1-x},$$

and taking the derivative of both sides!

# Heap-Sort

Heap-Sort (A)

1. Build-Max-Heap (A)
2. for ( $i = \text{length}(A); i > 1; i --$ )
  - (a) do exchange  $A[1]$  and  $A[i]$
  - (b)  $\text{heap-size}(A) = \text{heap-size}(A) - 1$
  - (c) Max-Heapify (A,1)



# Analysis

- Build-Max-Heap takes  $O(n)$ , and each of the  $O(n)$  calls to Max-Heapify take  $O(\log n)$ , so Heap-Sort takes  $O(n \log n)$
- Correctness???