

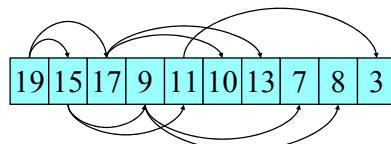
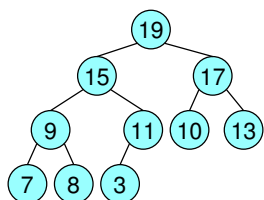
# CSE 502N

## Fundamentals of Computer Science

Fall 2004  
 Lecture 11:  
 Binary Heaps & Heapsort  
 Priority Queues  
 (CLRS 6)

## Binary heaps

- A binary heap is an array-based data structure that may be viewed as a (nearly) complete binary tree
  - » Each node corresponds to an array element
  - » Tree is filled on all levels except possibly the lowest level which is filled from left to right
- An array  $A$  that represents a heap has two attributes:
  - »  $length[A]$  = the number of elements in the array
  - »  $heap-size[A]$  = the number of elements in the heap



Parent( $i$ )      Left( $i$ )      Right( $i$ )  
 return  $\lfloor i/2 \rfloor$     return  $2i$       return  $2i+1$

## Heap properties

- A **max-heap** is a binary heap that maintains the **max-heap property**:  $A[\text{Parent}(i)] \geq A[i]$ 
  - » Largest element is stored at the root
- A **min-heap** is a binary heap that maintains the **min-heap property**:  $A[\text{Parent}(i)] \leq A[i]$ 
  - » Smallest element is stored at the root
- The **height** of a node is the number of edges on the longest simple downward path from the node to a leaf
- The height of the heap is the height of the root node
  - » Since the heap is a nearly complete binary tree, its height is  $\Theta(\lg n)$

## Maintaining the heap property

- Max-Heapify assumes the binary trees rooted at  $\text{Right}(i)$  and  $\text{Left}(i)$  are max-heaps, but  $A[i]$  may be smaller than its children
- What is the running time?
- $T(n) = (\text{time to adjust } i, \text{Left}(i), \text{Right}(i)) + (\text{time to run Max-Heapify on one of children})$ 
  - » Maximum size of a child subtree is  $2n/3$  (occurs when lowest level is half full)
- $T(n) \leq T(2n/3) + \Theta(1)$
- $T(n) = \Theta(\lg n)$

```

MAX-HEAPIFY(A, i)
1  l ← LEFT(i)
2  r ← RIGHT(i)
3  if l ≤ heap-size[A] and A[l] > A[i]
4    then largest ← l
5  else largest ← i
6  if r ≤ heap-size[A] and A[r] > A[largest]
7    then largest ← r
8  if largest ≠ i
9    then exchange A[i] ↔ A[largest]
10  MAX-HEAPIFY(A, largest)
    
```

## Building a heap

- Use Max-Heapify in a bottom-up manner to convert an array into a max-heap
  - » Elements  $A[\lfloor n/2 \rfloor + 1] \dots n$  are all leaves of the tree
- Loop invariant: at the start of each iteration of the for loop of lines 2-3, each node  $i+1, i+2, \dots, n$  is the root of a max-heap
- Initialization: prior to the first iteration of the loop,  $i = \lfloor n/2 \rfloor$ 
  - » Each node  $\lfloor n/2 \rfloor + 1, \lfloor n/2 \rfloor + 2, \dots, n$  is a leaf and is thus the root of a trivial max-heap
- Maintenance: by the loop invariant  $\text{Left}(i)$  and  $\text{Right}(i)$  are both roots of max-heaps
  - » This is the condition required for Max-Heapify to make node  $i$  a max-heap root
  - » Max-Heapify preserves property that nodes  $i+1, \dots, n$  are max-heap roots
  - » Following Max-Heapify, decrementing  $i$  reestablishes the loop invariant
- Termination: at termination,  $i=0$ ; by the loop invariant each node  $1, 2, \dots, n$  is a max-heap root
  - » Thus, node 1 is a max-heap root

```

BUILD-MAX-HEAP(A)
1  heap-size[A] ← length[A]
2  for i ← ⌊length[A]/2⌋ downto 1
3    do MAX-HEAPIFY(A, i)
    
```

## Running time of Build-Max-Heap

- Easy to derive an  $O(n \lg n)$  bound
  - » Upper bound for Max-Heapify is  $O(\lg n)$
  - » Build-Max-Heap calls Max-Heapify  $\lfloor n/2 \rfloor$  times
- Can we derive a tighter bound?
- Time to execute Max-Heapify varies with the height of the node in the tree
  - » Execution time for node of height  $h$  is  $O(h)$
- An  $n$ -element heap has height  $\lfloor \lg n \rfloor$
- An  $n$ -element heap has at most  $\lceil n/2^{h+1} \rceil$  nodes of any height  $h$ 

$$\sum_{h=0}^{\lfloor \lg n \rfloor} \left\lceil \frac{n}{2^{h+1}} \right\rceil O(h) = O\left( n \sum_{h=0}^{\lfloor \lg n \rfloor} \frac{h}{2^h} \right) = O\left( n \sum_{h=0}^{\infty} \frac{h}{2^h} \right) = O(n)$$
- We can show that Build-Max-Heap is  $O(n)$

## Heapsort

```

HEAPSORT(A)
1  BUILD-MAX-HEAP(A)
2  for i ← length[A] downto 2
3    do exchange A[1] ↔ A[i]
4      heap-size[A] ← heap-size[A] - 1
5      MAX-HEAPIFY(A, 1)
    
```

- Heapsort sorts array  $A$  in place, back to front (max to min)
- At the completion of each iteration of Max-Heapify,  $A[1]$  is the maximum element left in the heap
- Each iteration of the **for** loop places the maximum element in its sorted position and reduces the heap-size (removes it from the heap), then runs Max-Heapify to find the maximum element left in the heap
- $T(n) = O(n) + (n-1)O(\lg n) = O(n \lg n)$

## Priority queues

- A **priority queue** is a data structure for maintaining a set of  $S$  elements with associated *key* values
- Elements may be inserted into a **priority queue** at any time
- In **max-priority queues** only the maximum element may be removed from the queue
  - » The **increase-key** operation may be used to increase the value of an element's key and possibly change the relative ordering of elements
  - » **Min-priority queues** implement symmetric operations
- Priority queues are useful for a multitude of scheduling tasks where the highest priority element/job/packet should be scheduled first
- Max-priority queues leverage the constant time Heap-Maximum operation

```
HEAP-MAXIMUM(A)
1 return A[1]
```

## Extract-Max

- For priority queues implemented with max-heaps, Heap-Extract-Max can be used to remove the maximum element
- $T(n) = O(\lg n)$

```
HEAP-EXTRACT-MAX(A)
1 if heap-size[A] < 1
2   then error "heap underflow"
3 max ← A[1]
4 A[1] ← A[heap-size[A]]
5 heap-size[A] ← heap-size[A] - 1
6 MAX-HEAPIFY(A, 1)
7 return max
```

$O(1)$

$O(\lg n)$

## Increase-Key

- Given the index  $i$  of an element and a new *key* value (greater than the current *key* value), Heap-Increase-Key increases the key value of the element and ensures max-heap property holds
- After changing key value, Heap-Increase-Key traverses a path from  $A[i]$  toward the root
  - » Compares keys of an element and its parent and exchanges if the element's key is greater than the parent's key
- $T(n) = O(\lg n)$

```
HEAP-INCREASE-KEY(A, i, key)
1 if key < A[i]
2   then error "new key is smaller than current key"
3 A[i] ← key
4 while i > 1 and A[PARENT(i)] < A[i]
5   do exchange A[i] ↔ A[PARENT(i)]
6   i ← PARENT(i)
```

## Insert

- Given a new key value, Max-Heap-Insert creates a new heap element with the give key value and ensures max-heap property is maintained
- Increases heap-size by 1
- Assigns new element the minimum key value
- Calls Heap-Increase-Key using new element's index and new key value
- $T(n) = O(\lg n)$

```
MAX-HEAP-INSERT(A, key)
1 heap-size[A] ← heap-size[A] + 1
2 A[heap-size[A]] ← -∞
3 HEAP-INCREASE-KEY(A, heap-size[A], key)
```