

Heaps, Heapsort, Priority Queues

So Far

Insertion Sort: $O(n^2)$ worst case

Linked List: $O(n)$ search, some operations $O(n^2)$

Heap: Data structure and associated algorithms,
Not garbage collection context

Binary Tree

- A binary tree is a structure that can be visualized as an upside-down tree where the root is at the top and the leaves are at the bottom
 - Each node in the tree has at most two children
 - If each level is completely filled-in then it is called a **complete** binary tree
- We can create binary trees in several ways!

Sample Binary Tree

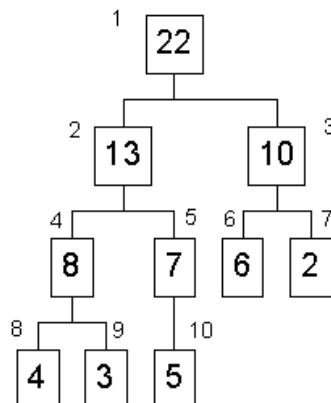
Parent/Child Relationship, Terminology

Heap Structure

- An array of objects than can be viewed as a complete binary tree such that:
 - Each tree node corresponds to elements of the array
 - The tree is complete except possibly the lowest level, filled from left to right
 - The heap property for all nodes i in the tree must be maintained except for the root:
 - Parent node(i) \geq i (could flip this if desired)

Heap Example

- Given array [22 13 10 8 7 6 2 4 3 5]



Note that the elements are not sorted, only max element at root of tree
Arrays are 1-based not 0-based

Height of the Heap

- The **height** of a node in the tree is the number of edges on the longest simple downward path from the node to a leaf; e.g. height of node 6 is 0, height of node 4 is 1, height of node 1 is 3.
- The height of the tree is the height from the root. As in any complete binary tree of size n , this is $\lg n$.
- Caveats: 2^h nodes at level h . $2^{h+1}-1$ total nodes in a complete binary tree.

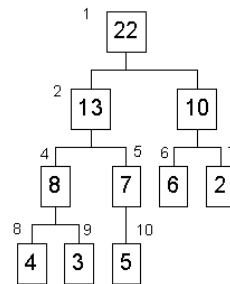
Heap Attributes

- A heap represented as an array A has two attributes:
 - $\text{Length}(A)$ – Size of the array
 - $\text{HeapSize}(A)$ - Size of the heap
- The property $\text{Length}(A) \geq \text{HeapSize}(A)$ must be maintained. (why ?)
- The heap property is stated as $A[\text{parent}(I)] \geq A[I]$

Computing Parents, Children

- The root of the tree is $A[1]$.
- Formula to compute parents, children in an array:

- $\text{Parent}(I) = A[\text{floor}(I/2)]$
- $\text{Left Child}(I) = A[2I]$
- $\text{Right Child}(I) = A[2I+1]$



Priority Queues

- Where might we want to use heaps? Consider the Priority Queue problem
 - Given a sequence of objects with varying degrees of priority, and we want to deal with the highest-priority item first.
- Managing air traffic control
 - Want to do most important tasks first.
 - Jobs placed in queue with priority, controllers take off queue from top
- Scheduling jobs on a processor
 - Critical applications need high priority
- Event-driven simulator with time of occurrence as key.
 - Use min-heap, which keeps smallest element on top, get next occurring event.

Extracting Max

- To support these operations we need to extract the maximum element from the heap:

```

HEAP-EXTRACT-MAX(A)
  remove A[1]
  A[1] ← A[n]      ; n is HeapSize(A), the length of the heap, not array
  n ← n-1          ; decrease size of heap
  Heapify(A,1,n)   ; Remake heap to conform to heap properties

```

Runtime: $\Theta(1)$ + Heapify time

Note: Arrays in this example are 1-based, not 0-based
 Successive removals will result in items in reverse sorted order!

Heapify Routine

- Heapify maintains heap property by “floating” a value down the heap that starts at I until it is in the right position.

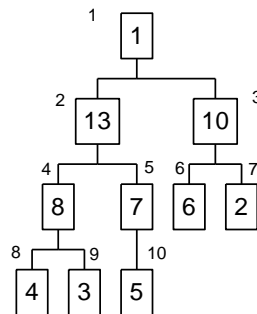
```

Heapify(A,I,n)      ; Array A, heapify node I, heapsize is n
  ; Note that the left and right subtrees of I are also heaps
  ; Make I's subtree be a heap.
  If  $2I \leq n$  and  $A[2I] > A[I]$ 
    ; see which is largest of current node and its children
    then largest ← 2I
    else largest ← I
  If  $2I+1 \leq n$  and  $A[2I+1] > A[largest]$ 
    then largest ← 2I+1
  If largest ≠ I
    then swap  $A[I] \leftrightarrow A[largest]$ 
    Heapify(A,largest,n)

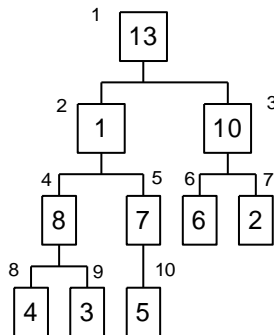
```

Heapify Example

- Heapify(A,1,10).
A=[1 13 10 8 7 6 2 4 3 5]

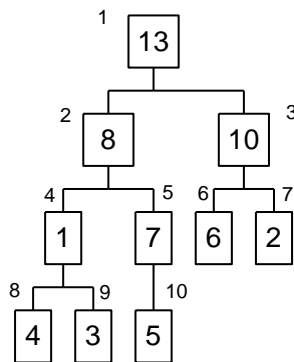


Heapify Example



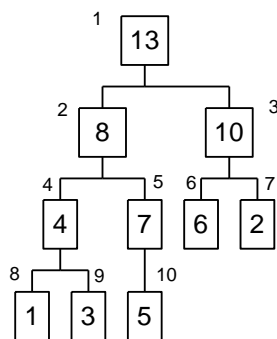
- Next is Heapify(A,2,10).
A=[13 1 10 8 7 6 2 4 3 5]

Heapify Example



- Next is Heapify(A,4,10).
A=[13 8 10 1 7 6 2 4 3 5]

Heapify Example



- Next is Heapify(A,8,10).
A=[13 8 10 4 7 6 2 1 3 5]
- On this iteration we have reached a leaf and are finished.

Heapify Runtime

- Later you will see recurrence relations, which recursively specify how long the algorithm takes to run:

$$T(n) \leq T\left(\frac{2n}{3}\right) + \Theta(1)$$

- We can always split the problem into at least $2/3$ the size.
- Solving this yields $\Theta(\lg n)$ runtime in the worst case, $O(1)$ in the best case, for $O(\lg n)$ overall.
- Basically we start at the top and move toward the bottom; since the tree is balanced we only make $\lg(n)$ swaps from the root to a leaf.

Building the Heap

- Given an array A , we want to build this array into a heap.
- Note: Leaves are already a heap! So start from the leaves and build up from there.

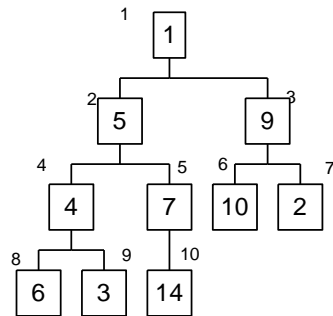
```
Build-Heap(A,n)
  for I = n downto 1          ; could we start at n/2?
    do Heapify(A,I,n)
```

- Start with the leaves (last $\frac{1}{2}$ of A) and consider each leaf as a 1 element heap. Call heapify on the parents of the leaves, and continue recursively to call Heapify, moving up the tree to the root.

Build-Heap Example

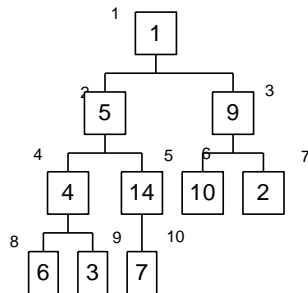
- Build-Heap(A,10)

A=[1 5 9 4 7 10 2 6 3 14]



Heapify(A,10,10) exits since this is a leaf.
 Heapify(A,9,10) exits since this is a leaf.
 Heapify(A,8,10) exits since this is a leaf.
 Heapify(A,7,10) exits since this is a leaf.
 Heapify(A,6,10) exits since this is a leaf.
 Heapify(A,5,10) puts the largest of A[5]
 and its children, A[10] into A[5]:

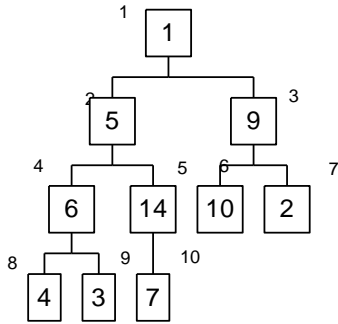
Build-Heap Example



A=[1 5 9 4 14 10 2 6 3 7]

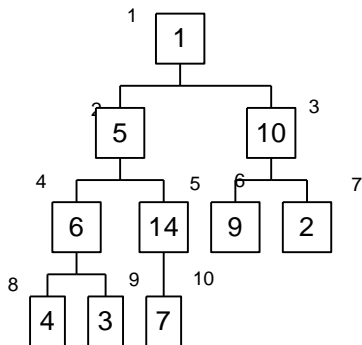
Heapify(A,4,10)

Build-Heap Example



- $A=[1\ 5\ 9\ 6\ 14\ 10\ 2\ 4\ 3\ 7]$
- $\text{Heapify}(A,3,10)$:

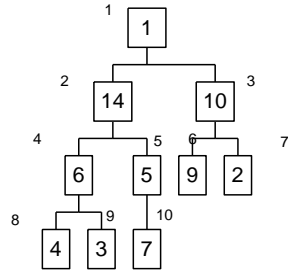
Build-Heap Example



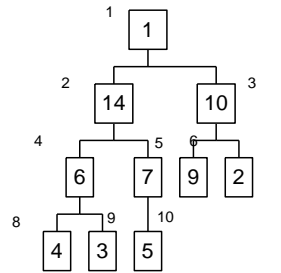
- $A=[1\ 5\ 10\ 6\ 14\ 9\ 2\ 4\ 3\ 7]$
- $\text{Heapify}(A,2,10)$:

Build-Heap Example

Heapify(A,2,10)



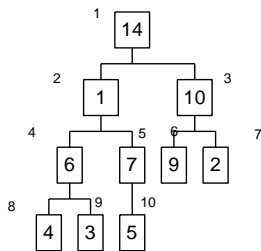
Heapify(A,5,10)



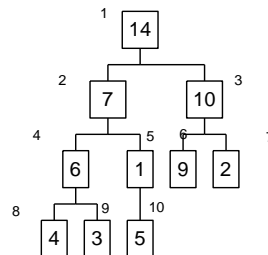
- A=[1 14 10 6 7 9 2 4 3 5]
- Heapify(A,1,10):

Build-Heap Example

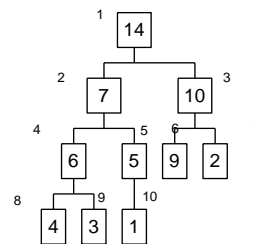
Heapify(A,1,10)



Heapify(A,2,10)



Heapify(A,5,10)



- Finished heap: A=[14 7 10 6 5 9 2 4 3 1]

Build-Heap Runtime

- Running Time
 - We have a loop of n times, and each time call heapify which runs in $(\lg n)$. This implies a bound of $O(n \lg n)$. This is correct, but is a loose bound! We can do better.
 - Key observation: Each time heapify is run within the loop, it is not run on the entire tree. We run it on subtrees, which have a lower height, so these subtrees do not take $\lg n$ time to run. Since the tree has more nodes on the leaf, most of the time the heaps are small compared to the size of n .

Build-Heap Runtime

- Property: In an n -element heap there are at most $n/(2^h)$ nodes at height h
 - The leaves are $h=1$ and root at $\lg n$, this is a slight change from the previous definition (leaves at height 0)
- The time required by Heapify when called in Build-Heap on a node at height h is $O(h)$; $h=\lg n$ for the entire tree.

Build-Heap Runtime

- Cost of Build-Heap is then:

$$T(n) = \sum_{h=1}^{\text{heap_height}} (\#nodes_at_h)(\text{Heapify-Time})$$

$$T(n) = \sum_{h=1}^{\lg n} \frac{n}{2^h} O(h)$$

$$T(n) = O\left(\sum_{h=1}^{\lg n} \frac{n}{2^h} h\right)$$

Build-Heap Runtime

- We know that: $\sum_{n=0}^{\infty} nx^n = \frac{x}{(1-x)^2}$

- If $x=1/2$ then $(1/2)^n=1/(2^n)$ so:

$$\sum_{n=0}^{\infty} h\left(\frac{1}{2}\right)^h = \frac{1/2}{(1-1/2)^2} = 2$$

- Substitute this in our equation, which is safe because the sum from 0 to infinity is LARGER than the sum from 1 to lgn.

$$T(n) = O\left(\sum_{h=1}^{\lg n} \frac{n}{2^h} h\right) < O\left(n \sum_{h=1}^{\infty} \frac{h}{2^h}\right) < O(n \cdot 2) = O(n)$$

Heapsort

- Once we can build a heap and heapify, sorting is easy... just remove max N times

```

HeapSort(A,n)
  Build-Heap(A,n)
  for I ← n downto 2
    do   Swap(A[1] ↔ A[I])
         Heapify(A,1,I-1)

```

Runtime is $O(n \lg n)$ since we do Heapify on $n-1$ elements, and we do Heapify on the whole tree.

Note: In-place sort, required no extra storage variables unlike Merge Sort, which used extra space in the recursion.

Heap Variations

- Heap could have min on top instead of max
- Heap could be k -ary tree instead of binary
- Priority Queue
 - Desired Operations
 - Insert(S,x) puts element x into set S
 - Max(S,x) returns the largest element in set S
 - Extract-Max(S) removes the largest element in set S

Priority Queue implemented via Heap

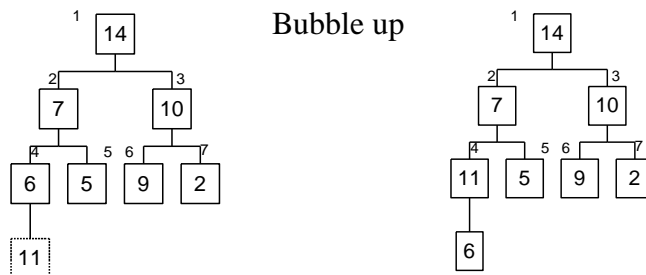
- $\text{Max}(S,x)$
 - Just return root element. Takes $O(1)$ time.
- $\text{Insert}(S,x)$
 - Similar idea to heapify, put new element at end, bubble up to proper place toward root

```

Heap-Insert(A,key)
  n ← n+1
  I ← n
  while I > 1 and A[⌊i/2⌋] < key
    do  A[I] ← A[⌊i/2⌋]
       I ← ⌊i/2⌋
  A[I] ← key
  
```

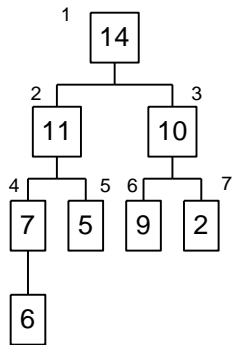
Insert Example

- Insert new element “11” starting at new node on bottom, $I=8$



Insert Example

- Bubble up once more



Stop at this point, since parent (index 1, value 14) has a larger value

Runtime = $O(\lg n)$ since we only move once up the tree

Extract Max

- To extract the max, copy the last element to the root and heapify

```

Heap-Extract-Max(A,n)
  max ← A[1]
  A[1] ← A[n]
  n ← n-1
  Heapify(A,1,n)
  return max
  
```

$O(\lg n)$ time

Can implement priority queue operations in $O(\lg n)$ time, $O(n)$ to build