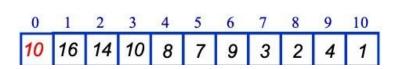
Heap Sort and Priority Queue

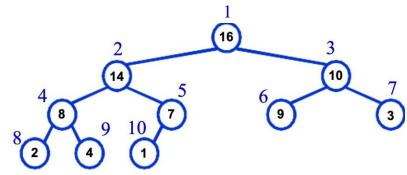
Charles Aunkan Gomes
Lecturer, Dept. of CSE
United International University
charles@cse.uiu.ac.bd



Binary Heaps

- The (binary) heap data structure is an array object that can be viewed as a complete binary tree
- ■Each node of the tree corresponds to an element of the array that stores the value in the node.
 - ■The tree is completely filled on all levels except possibly the lowest, where it is filled from the left up to a point.





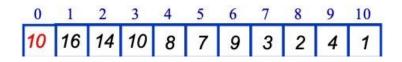
Binary Heaps

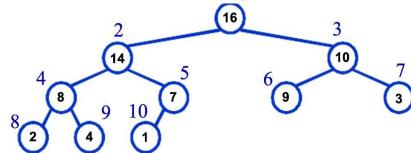
- To represent a complete binary tree as an array:
 - The root node is A[1]
 - Node i is A[i]
 - The parent of node i is A[i/2]
 - The left child of node i is A[2i]
 - The right child of node i is A[2i + 1]

```
Right(i)
return 2i+1
Left(i)
return 2i
```

return floor (i/2)

Parent(i)



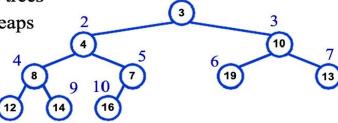


Types of Binary Heaps

• Min-Heaps:

■ The element in the root is less than or equal to all elements in both of its sub-trees

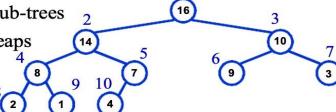
■ Both of its sub-trees are Min-Heaps



• Max-Heaps:

■ The element in the root is greater than or equal to all elements in both its sub-trees

■ Both of its sub-trees are Max-Heaps

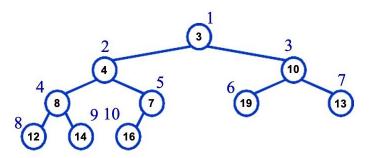


The Min-Heap Property

•Min-Heaps satisfy the heap property:

 $A[Parent(i)] \le A[i]$; for all nodes i > 1

- ■The value of a node is at least the value of its parent
- ■The smallest element in a min-heap is stored at the root
- ■Where is the largest element ???
 - \bullet Ans: At one of the leaves [leaf indices are n/2+1 to n]

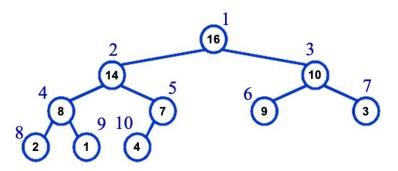


The Max-Heap Property

Max-Heaps satisfy the heap property:

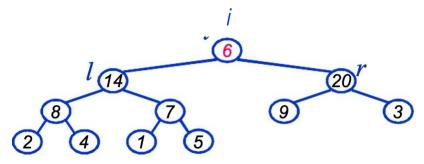
 $A[Parent(i)] \ge A[i]$; for all nodes i > 1

- ■The value of a node is at most the value of its parent
- ■The largest element in a max-heap is stored at the root
- ■Where is the smallest element ???
 - \bullet Ans: At one of the leaves [leaf indices are n/2+1 to n]



Max-Heap Operations: Max-Heapify()

- Max-Heapify(): maintain the max-heap property
 - Given: a node i in the heap with children l and r: two subtrees rooted at l and r, assumed to be heaps
 - ■Problem: The subtree rooted at *i* may violate the heap property (*How?*)
 - Action: let the value of the parent node "float down" so subtree at i satisfies the heap property
 - ◆What will be the basic operation between i, l, and r?



Max-Heap Operations: Max-Heapify()

```
MAX-HEAPIFY (A, i)

1 l \leftarrow \text{LEFT}(i)

2 r \leftarrow \text{RIGHT}(i)

3 if l \leq heap\text{-}size[A] and A[l] > A[i]

4 then largest \leftarrow l

5 else largest \leftarrow i

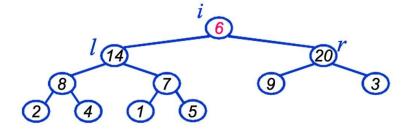
6 if r \leq heap\text{-}size[A] and A[r] > A[largest]

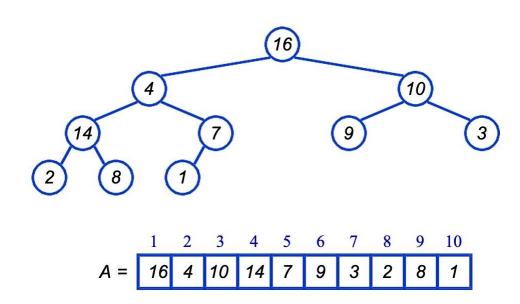
7 then largest \leftarrow r

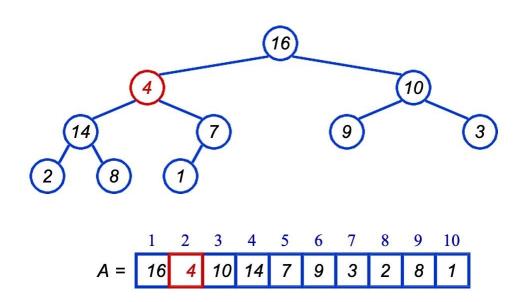
8 if largest \neq i

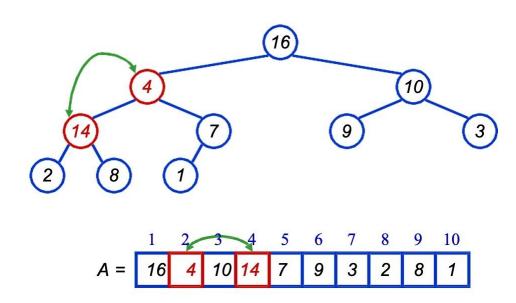
9 then exchange A[i] \leftrightarrow A[largest]

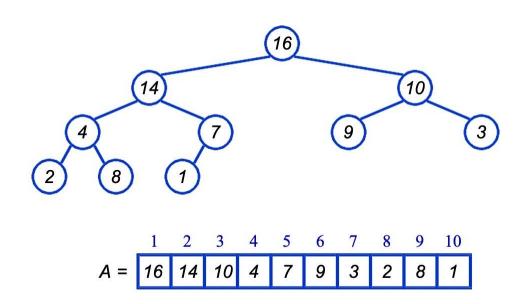
10 MAX-HEAPIFY (A, largest)
```

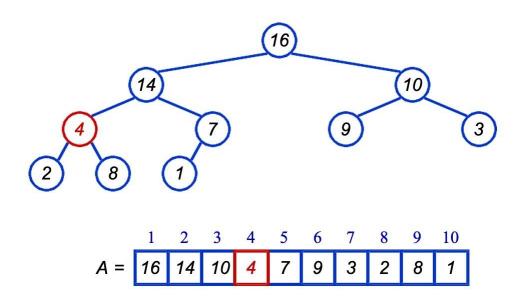


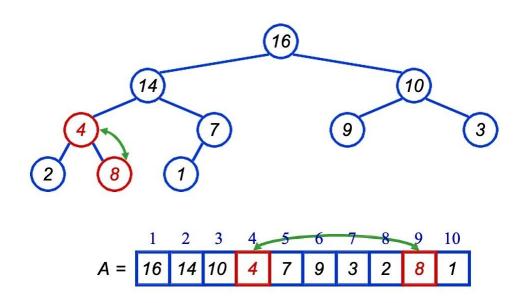


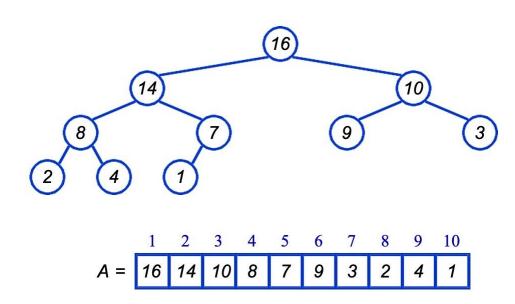


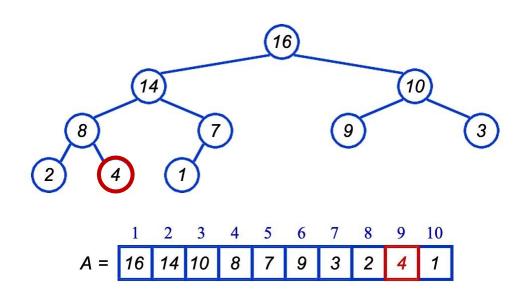


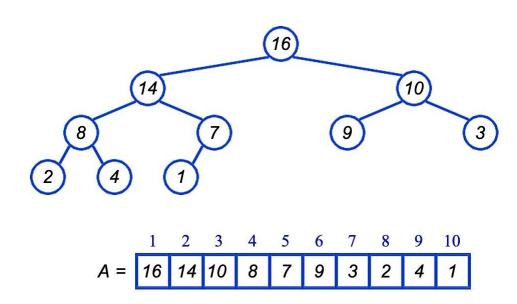






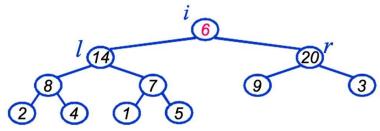






Analyzing Heapify()

- Fixing up relationships among the elements A[i], A[I], and A[r] takes O(1) time
- •If the heap at i has n elements, at most how many elements can the subtrees at I or r have?



- Answer: 2n/3 (worst case: bottom row half full)
- •So time taken by **Heapify()** is given by $T(n) \le T(2n/3) + \Theta(1)$

Analyzing Heapify()

- So we have
 - $T(n) \le T(2n/3) + \Theta(1)$
- Solving the recurrence, we have
 - $T(n) = O(\log n)$
- Thus, Heapify() takes O(h) time for a node at height h.

Heap Operations: BuildHeap()

- We can build a heap in a bottom-up manner by running Heapify() on successive subarrays
- Fact: for array of length n, all elements in the range $A[\lfloor n/2 \rfloor + 1 \dots n]$ already satisfies Heap Property (Why?)
- So
 - ◆Walk backwards through the array from *n*/2 to 1, calling **Heapify()**on each node.
 - ◆Order of processing guarantees that the children of node *i* are heaps when *i* is processed

Heap Operations: BuildHeap()

 Converts an unorganized array A into a max-heap.

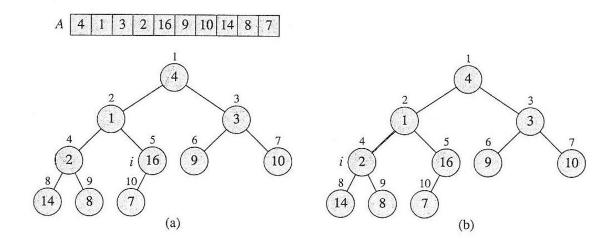
```
BUILD-MAX-HEAP(A)

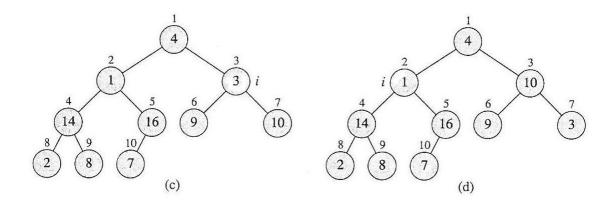
1 heap-size[A] \leftarrow length[A]

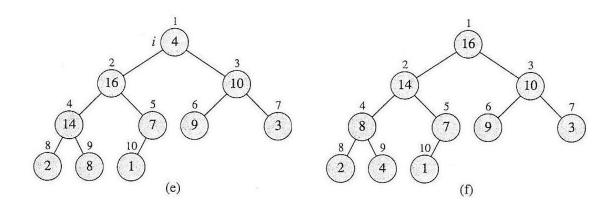
2 \mathbf{for}\ i \leftarrow \lfloor length[A]/2 \rfloor \mathbf{downto}\ 1

3 \mathbf{do}\ \mathrm{MAX}-HEAPIFY(A, i)
```

Work through example
A = {4, 1, 3, 2, 16, 9, 10, 14, 8, 7}







Work through example

 $A = \{14, 11, 33, 22, 56, 49, 30, 24, 18, 37\}$

Construct a Min-Heap from the given array using Build-Heap function.

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)
 - ■How can this be? Is there a flaw in the above reasoning?

Analyzing BuildHeap(): Tight

- To Heapify() a subtree takes O(h) time, where h is the height of the subtree
 - $\blacksquare h = O(\log m), m = \# \text{ nodes in the subtree}$
 - ■The height of most subtrees is small
- Fact: an *n*-element heap has at most $\lceil n/2^{h+1} \rceil$ nodes of height *h*
- Prove that **BuildHeap()** takes O(n) time

HeapSort

- Given **BuildHeap()**, a sorting algorithm can easily be constructed:
 - ■Maximum element is at A[1]
 - Discard by swapping with element at A[n]
 - ◆Decrement heap_size[A]
 - ◆A[n] now contains correct value
 - ■Restore heap property at A[1] by calling **Heapify()**
 - ■Repeat, always swapping A[1] for A[heap_size(A)]

HeapSort

```
Heapsort(A)
{
    BuildHeap(A);
    for (i = length(A) downto 2)
    {
        Swap(A[1], A[i]);
        heap_size(A) = heap_size(A) - 1;
        Heapify(A, 1);
    }
}
```

HeapSort

Work through example

 $A = \{14, 11, 33, 22, 56, 49, 30, 24, 18, 37\}$

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()

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

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

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

Priority Queue

- A queue that is ordered according to some priority value
- The heap data structure is incredibly useful for implementing priority queues
 - ■A data structure for maintaining a set S of elements, each with an associated value or key
 - Supports the operations Insert(), Maximum(), and ExtractMax()

- Insert(S, x) Inserts element x into set S, according to its priority
- Maximum(S) Returns, but does not remove, the element of S with the largest key
- Extract-Max(S) Removes and returns the element of S with the largest key
- Increase-Key(S, x, k) Increases the value of element x's key to the new value k
 - •How could we implement these operations using a heap?

```
HEAP-EXTRACT-MAX(A)

1 if heap-size[A] < 1

2 then error "heap underflow"

3 max \leftarrow A[1]

4 A[1] \leftarrow A[heap-size[A]]

5 heap-size[A] \leftarrow heap-size[A] - 1

6 MAX-HEAPIFY(A, 1)

7 return max
```

HEAP-MAXIMUM(A)

1 return A[1]

```
HEAP-INCREASE-KEY (A, i, key)

1 if key < A[i]

2 then error "new key is smaller than current key"

3 A[i] \leftarrow key

4 while i > 1 and A[PARENT(i)] < A[i]

5 do exchange A[i] \leftrightarrow A[PARENT(i)]

6 i \leftarrow PARENT(i)
```

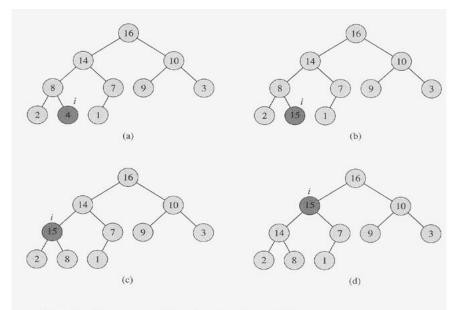


Figure 6.5 The operation of HEAP-INCREASE-KEY. (a) The max-heap of Figure 6.4(a) with a node whose index is i heavily shaded. (b) This node has its key increased to 15. (c) After one iteration of the while loop of lines 4–6, the node and its parent have exchanged keys, and the index i moves up to the parent. (d) The max-heap after one more iteration of the while loop. At this point, $A[PARENT(i)] \ge A[i]$. The max-heap property now holds and the procedure terminates.

```
MAX-HEAP-INSERT(A, key)
```

- 1 heap- $size[A] \leftarrow heap$ -size[A] + 1
- 2 $A[heap-size[A]] \leftarrow -\infty$
- 3 HEAP-INCREASE-KEY (A, heap-size[A], key)

THANK YOU

