

## Design and Analysis of Algorithms

### Heap Sort

#### Lecture 9-10

Instructor: Dr. G P Gupta

## Heapsort

- Combines the better attributes of merge sort and insertion sort.
  - Like merge sort, but unlike insertion sort, running time is  $O(n \lg n)$ .
  - Like insertion sort, but unlike merge sort, **sorts in place**.
- Introduces an algorithm design technique
  - Create data structure (**heap**) to manage information during the execution of an algorithm.
- The **heap** has other applications beside sorting.
  - Priority Queues

## Binary heap

- A binary tree where the value of a parent is greater than or equal to the value of its children
- Additional restriction: all levels of the tree are **complete** except the last

### Max heap vs. Min heap

## Heap Property (Max and Min)

### Max-Heap

- For every node excluding the root,

$$A[\text{parent}[i]] \geq A[i]$$

- Largest element is stored at the root.**

- In any subtree, no values are **larger** than the value stored at subtree root.

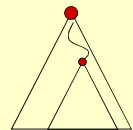
### Min-Heap

- For every node excluding the root,

$$A[\text{parent}[i]] \leq A[i]$$

- Smallest element is stored at the root.**

- In any subtree, no values are **smaller** than the value stored at subtree root



## Binary Heap

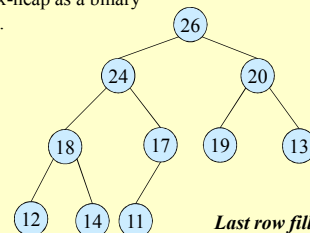
- Array viewed as a nearly complete binary tree.
  - Physically – linear array.
  - Logically – binary tree, filled on all levels (except lowest.)
- Map from array elements to tree nodes and vice versa
  - Root –  $A[1]$
  - Left[ $i$ ] –  $A[2i]$
  - Right[ $i$ ] –  $A[2i+1]$
  - Parent[ $i$ ] –  $A[\lfloor i/2 \rfloor]$
- $\text{length}[A]$  – number of elements in array  $A$ .
- $\text{heap-size}[A]$  – number of elements in heap stored in  $A$ .
  - $\text{heap-size}[A] \leq \text{length}[A]$

## Heaps – Example

26	24	20	18	17	19	13	12	14	11
1	2	3	4	5	6	7	8	9	10

Max-heap as an array.

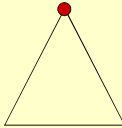
Max-heap as a binary tree.



Last row filled from left to right.

## Height

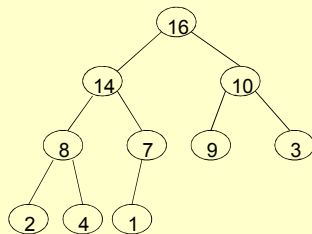
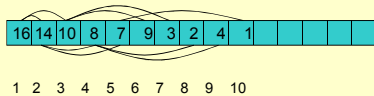
- ♦ **Height of a node in a tree:** the number of edges on the longest simple downward path from the node to a leaf.
- ♦ **Height of a tree:** the height of the root.
- ♦ **Height of a heap:**  $\lfloor \lg n \rfloor$ 
  - » Basic operations on a heap run in  $O(\lg n)$  time



## Binary heap - operations

- Maximum(S) - return the largest element in the set
- ExtractMax(S) - Return and remove the largest element in the set
- Insert(S, val) - insert val into the set
- IncreaseElement(S, x, val) - increase the value of element x to val
- BuildHeap(A) - build a heap from an array of elements

## Binary heap representations



## Heapify

Assume left and right children are heaps, turn current set into a valid heap

```

HEAPIFY(A, i)
1  l ← LEFT(i)
2  r ← RIGHT(i)
3  largest ← i
4  if l ≤ heap-size[A] and A[l] > A[i]
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```

find out which is  
largest: current,  
left of right

## Heapify

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## Heapify

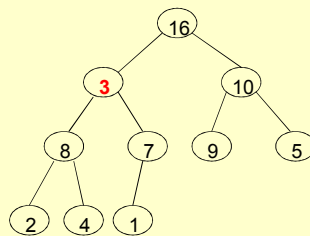
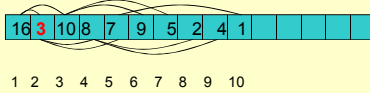
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```

if a child is  
larger, swap and  
recurse

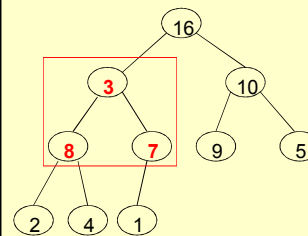
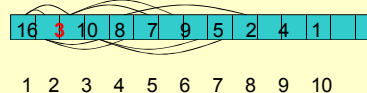
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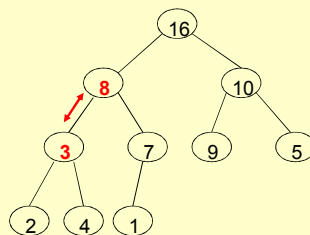
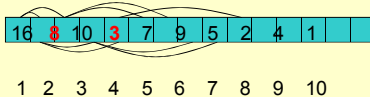
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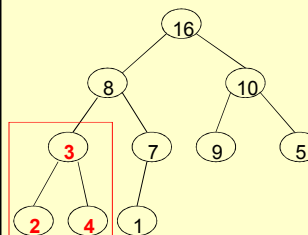
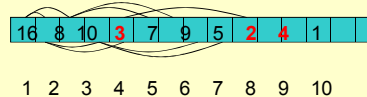
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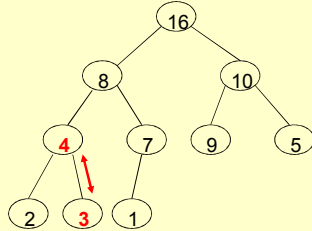
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## Heapify

16 8 10 4 7 9 5 2 3 1

1 2 3 4 5 6 7 8 9 10



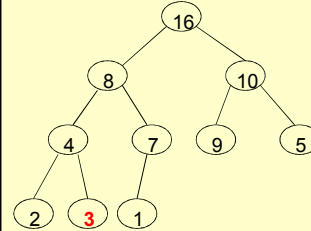
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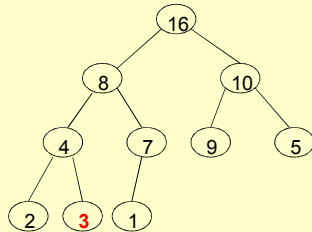
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## Correctness of Heapify

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```

## Correctness of Heapify

Base case:

- Heap with a single element
- Trivially a heap

```

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```

## Correctness of Heapify

Both children are valid heaps

Three cases:

**Case 1:**  $A[i]$  (current node) is the largest

```

8 if largest ≠ i
9     swap A[i] and A[largest]
10    HEAPIFY(A, largest)
    
```

- parent is greater than both children
- both children are heaps
- current node is a valid heap

## Correctness of Heapify

### Case 2: left child is the largest

```

8  if largest ≠ i
9      swap A[i] and A[largest]
10     HEAPIFY(A, largest)

```

- When Heapify returns:
  - Left child is a valid heap
  - Right child is unchanged and therefore a valid heap
  - Current node is larger than both children since we selected the largest node of current, left and right
  - current node is a valid heap

### Case 3: right child is largest

- similar to above

## Running time of Heapify

What is the cost of each individual call to Heapify (not counting recursive calls)?

»  $\Theta(1)$

How many calls are made to Heapify?

»  $O(\text{height of the tree})$

What is the height of the tree?

» Complete binary tree, except for the last level

$$2^h \leq n$$

$$h \leq \log_2 n$$

$$O(\log n)$$

```

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## Binary heap - operations

Maximum(S) - return the largest element in the set

ExtractMax(S) – Return and remove the largest element in the set

Insert(S, val) – insert val into the set

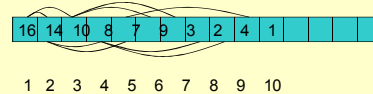
IncreaseElement(S, x, val) – increase the value of element x to val

BuildHeap(A) – build a heap from an array of elements

## Maximum

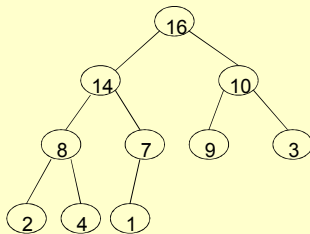
Return the largest element from the set

Return A[1]



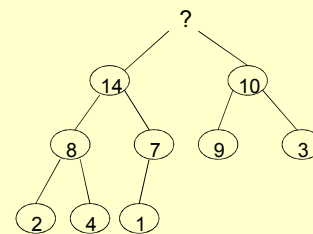
## ExtractMax

Return and remove the largest element in the set



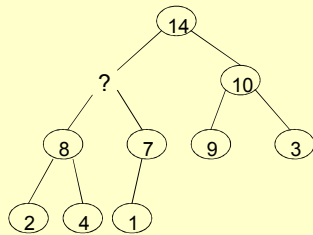
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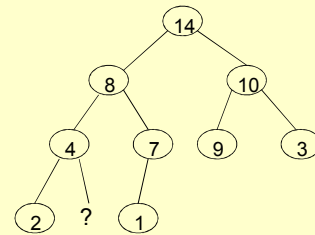
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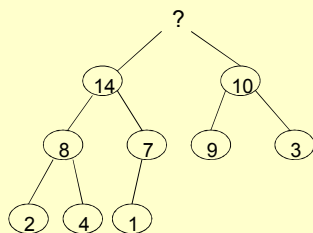
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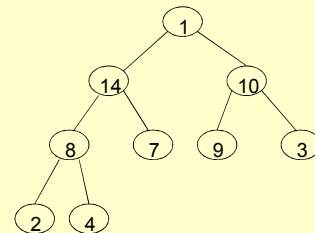
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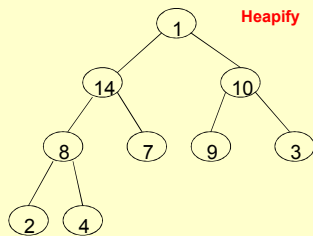
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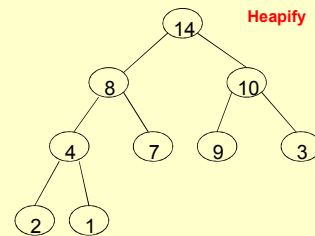
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Return and remove the largest element in the set



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## ExtractMax

Return and remove the largest element in the set

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

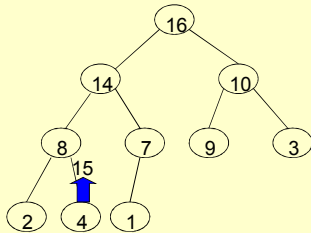
## ExtractMax running time

Constant amount of work plus one call to Heapify –  $O(\log n)$

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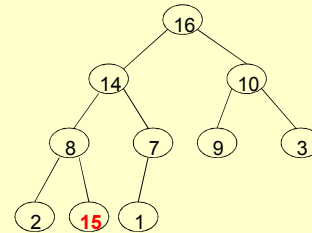
## IncreaseElement

Increase the value of element *x* to *val*



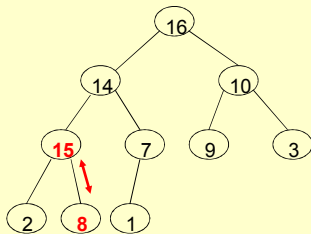
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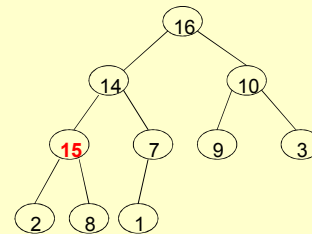
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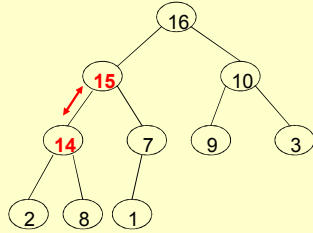
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Increase the value of element *x* to *val*



## IncreaseElement

Increase the value of element  $x$  to  $val$



## IncreaseElement

Increase the value of element  $x$  to  $val$

```
INCREASE-ELEMENT( $A, i, val$ )
1  if  $val < A[i]$ 
2      error
3   $A[i] \leftarrow val$ 
4  while  $i > 1$  and  $A[PARENT(i)] < A[i]$ 
5      swap  $A[i]$  and  $A[PARENT(i)]$ 
6       $i \leftarrow PARENT(i)$ 
```

## Correctness of IncreaseElement

Why is it ok to swap values with parent?

```
INCREASE-ELEMENT( $A, i, val$ )
1  if  $val < A[i]$ 
2      error
3   $A[i] \leftarrow val$ 
4  while  $i > 1$  and  $A[PARENT(i)] < A[i]$ 
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```

## Correctness of IncreaseElement

Stop when heap property is satisfied

```
INCREASE-ELEMENT( $A, i, val$ )
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2      error
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4  while  $i > 1$  and  $A[PARENT(i)] < A[i]$ 
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6       $i \leftarrow PARENT(i)$ 
```

## Running time of IncreaseElement

Follows a path from a node to the root

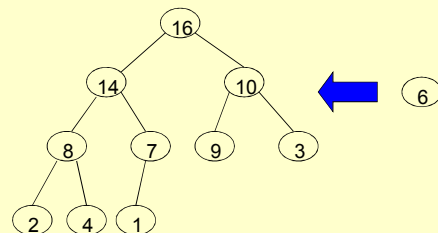
Worst case  $O(\text{height of the tree})$

$O(\log n)$

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## Insert

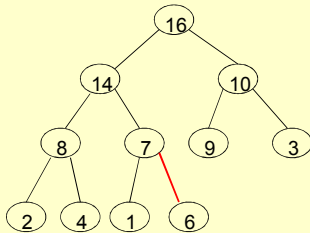
Insert  $val$  into the set





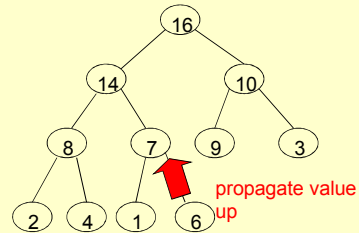
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Insert  $val$  into the set



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## Insert

```
INSERT( $A, val$ )  
1  $heap-size[A] \leftarrow heap-size[A] + 1$   
2  $A[heap-size[A]] \leftarrow -\infty$   
3 INCREASE-ELEMENT( $A, heap-size[A], val$ )
```

## Running time of Insert

Constant amount of work plus one call to  
IncreaseElement –  $O(\log n)$

```
INSERT( $A, val$ )  
1  $heap-size[A] \leftarrow heap-size[A] + 1$   
2  $A[heap-size[A]] \leftarrow -\infty$   
3 INCREASE-ELEMENT( $A, heap-size[A], val$ )
```

## **Building a heap**

Can we build a heap using the functions we have  
so far?

- Maximum( $S$ )
- ExtractMax( $S$ )
- Insert( $S, val$ )
- IncreaseElement( $S, x, val$ )

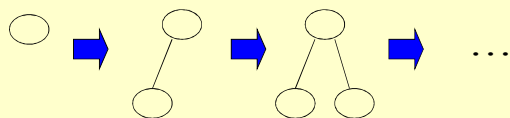
## Building a heap

```
BUILD-HEAP1( $A$ )  
1 copy  $A$  to  $B$   
2  $heap-size[A] \leftarrow 0$   
3 for  $i \leftarrow 1$  to  $length[B]$   
4     INSERT( $A, B[i]$ )
```

## Running time of BuildHeap1

$n$  calls to Insert –  $O(n \log n)$

Can we do better?



## Building a heap: take 2

```
BUILD-HEAP2(A)
1  heap-size[A] ← (length)[A]
2  for i ← ⌊(length)[A]/2⌋ to 1
3    HEAPIFY(A, i)
```

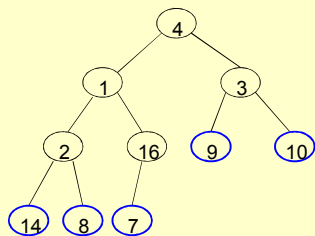
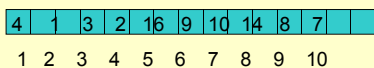
Start with  $n/2$  “simple” heaps

call Heapify on element  $n/2-1$ ,  $n/2-2$ ,  $n/2-3$  ...

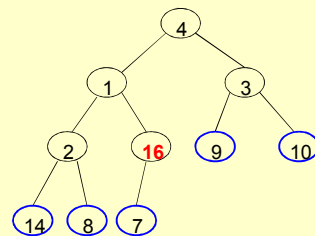
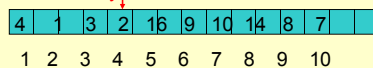
all children have smaller indices

building from the bottom up, makes sure that all the children are heaps

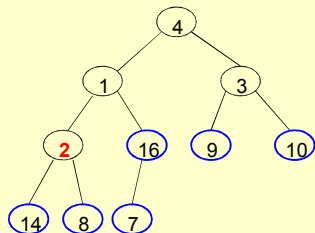
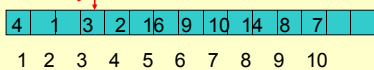
```
BUILD-HEAP2(A)
1  heap-size[A] ← (length)[A]
2  for i ← ⌊(length)[A]/2⌋ to 1
3    HEAPIFY(A, i)
```



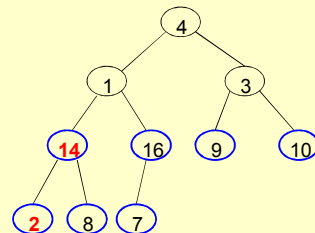
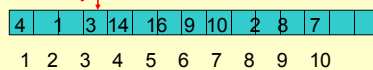
```
BUILD-HEAP2(A)
1  heap-size[A] ← (length)[A]
2  for i ← ⌊(length)[A]/2⌋ to 1
3    HEAPIFY(A, i)
```

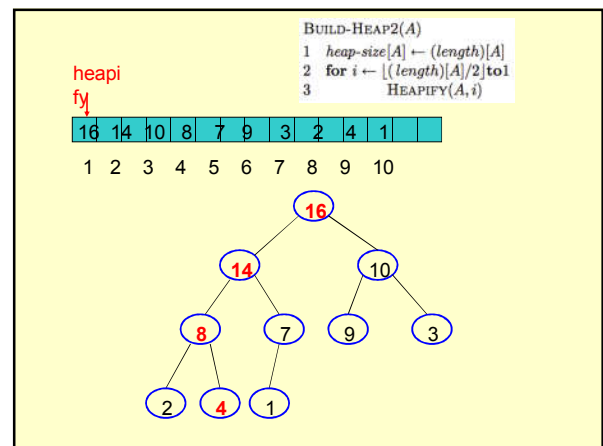
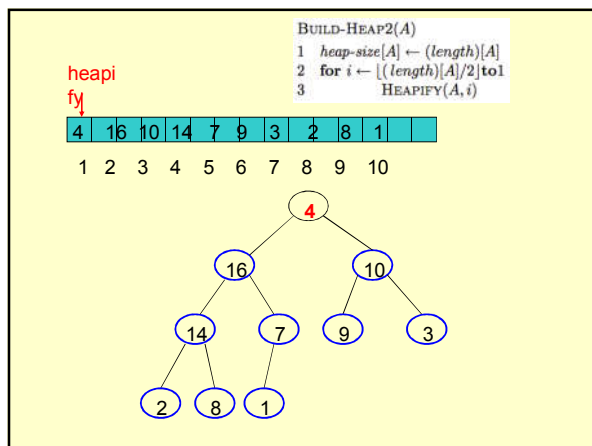
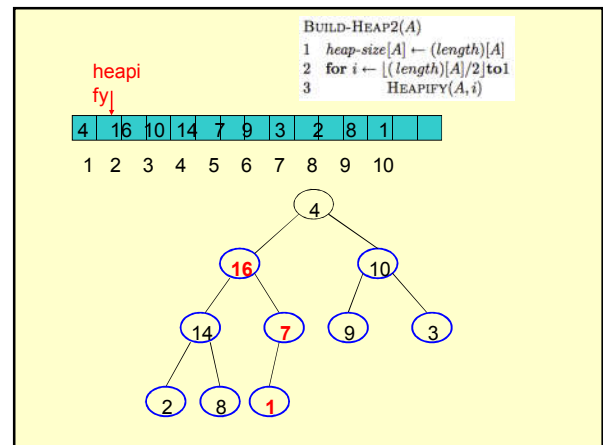
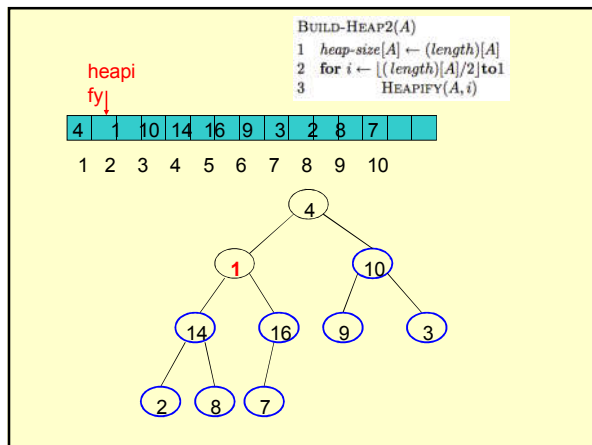
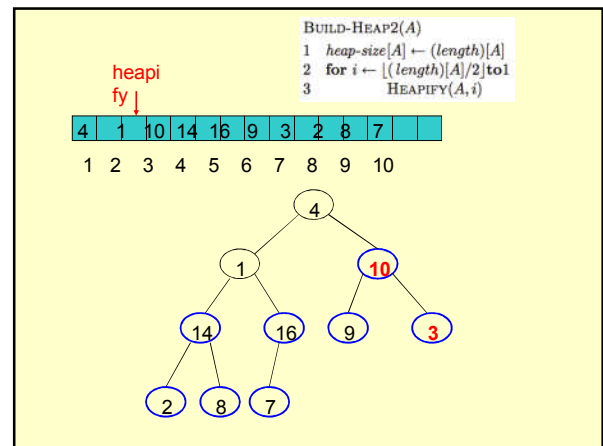
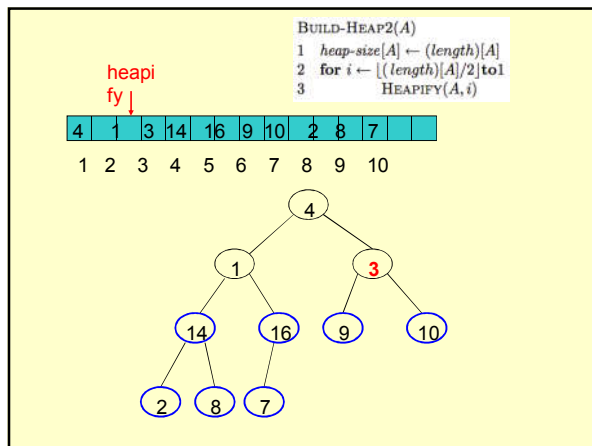


```
BUILD-HEAP2(A)
1  heap-size[A] ← (length)[A]
2  for i ← ⌊(length)[A]/2⌋ to 1
3    HEAPIFY(A, i)
```



```
BUILD-HEAP2(A)
1  heap-size[A] ← (length)[A]
2  for i ← ⌊(length)[A]/2⌋ to 1
3    HEAPIFY(A, i)
```





## Correctness of BuildHeap2

Invariant:

```
BUILD-HEAP2(A)
1  heap-size[A] ← (length)[A]
2  for i ← ⌊(length)[A]/2⌋ to 1
3      HEAPIFY(A, i)
```

## Correctness of BuildHeap2

```
BUILD-HEAP2(A)
1  heap-size[A] ← (length)[A]
2  for i ← ⌊(length)[A]/2⌋ to 1
3      HEAPIFY(A, i)
```

Invariant: elements  $A[i+1 \dots n]$  are all heaps

**Base case:**  $i = \text{floor}(n/2)$ . All elements  $i+1, i+2, \dots, n$  are “simple” heaps

**Inductive case:** We know  $i+1, i+2, \dots, n$  are all heaps, therefore the call to  $\text{Heapify}(A, i)$  generates a heap at node  $i$

Termination?

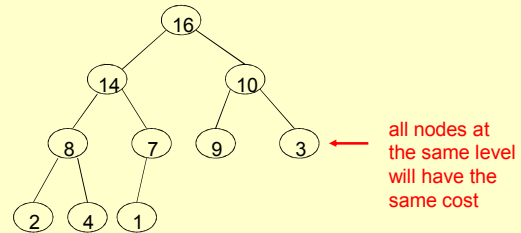
## Running time of BuildHeap2

$n/2$  calls to Heapify –  $O(n \log n)$

Can we get a tighter bound?

```
BUILD-HEAP2(A)
1  heap-size[A] ← (length)[A]
2  for i ← ⌊(length)[A]/2⌋ to 1
3      HEAPIFY(A, i)
```

## Running time of BuildHeap2



How many nodes are at level  $d$ ?  $2^d$

## Running time of BuildHeap2

$$T(n) = \sum_{d=0}^{\log n} 2^d O(d)$$

?

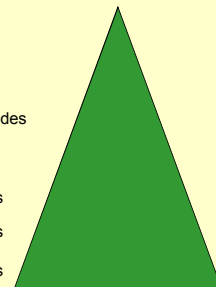
## Nodes at height $h$

$h$   $< \text{ceil}(n/2^{h+1})$  nodes

$h=2$   $< \text{ceil}(n/8)$  nodes

$h=1$   $< \text{ceil}(n/4)$  nodes

$h=0$   $< \text{ceil}(n/2)$  nodes



## Running time of BuildHeap2

$$\begin{aligned}
 T(n) &= \sum_{h=0}^{\log n} \left\lceil \frac{n}{2^{h+1}} \right\rceil O(h) \\
 &= O\left(n \sum_{h=0}^{\log n} \left\lceil \frac{1}{2^{h+1}} \right\rceil h\right) \\
 &= O\left(n \sum_{h=0}^{\log n} \frac{h}{2^h}\right) \\
 &= O\left(n \sum_{h=0}^{\infty} \frac{h}{2^h}\right) \\
 &= O(n) \quad \sum_{h=0}^{\infty} \frac{h}{2^h} = \frac{1/2}{(1-1/2)^2} = 2
 \end{aligned}$$

## BuildHeap1 vs. BuildHeap2

```

BUILD-HEAP1(A)
1  copy A to B
2  heap-size[A] ← 0
3  for i ← 1 to length[B]
4      INSERT(A, B[i])
    
```

```

BUILD-HEAP2(A)
1  heap-size[A] ← (length)[A]
2  for i ← ⌊(length)[A]/2⌋ to 1
3      HEAPIFY(A, i)
    
```

### Runtime

- $O(n)$  vs.  $O(n \log n)$

### Memory

- Both  $O(n)$
- BuildHeap1 requires an additional array, i.e.  $2n$  memory

### Complexity/Ease of implementation

## Heap uses

Could we use a heap to sort?

## Heap uses

### Heapsort

- Build a heap
- Call ExtractMax for all the elements
- $O(n \log n)$  running time

### Priority queues

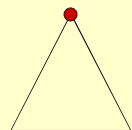
- scheduling tasks: jobs, processes, network traffic
- A\* search algorithm

## Heaps in Sorting

- Use **max-heaps for sorting**.
- The array representation of max-heap is not sorted.
- **Steps in sorting**
  - » Convert the given array of size  $n$  to a max-heap (*BuildMaxHeap*)
  - » Swap the first and last elements of the array.
    - Now, the largest element is in the last position – where it belongs.
    - That leaves  $n - 1$  elements to be placed in their appropriate locations.
    - However, the array of first  $n - 1$  elements is no longer a max-heap.
    - Float the element at the root down one of its subtrees so that the array remains a max-heap (*MaxHeapify*)
    - Repeat step 2 until the array is sorted.

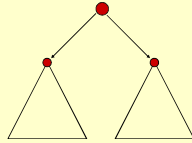
## Heap Characteristics

- ♦ *Height*  $= \lceil \lg n \rceil$
- ♦ No. of *leaves*  $= \lceil n/2 \rceil$
- ♦ No. of nodes of height  $h \leq \lceil n/2^{h+1} \rceil$



## Maintaining the heap property

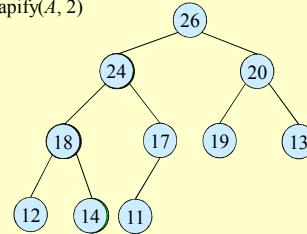
- Suppose two subtrees are max-heaps, but the root violates the max-heap property.



- Fix the offending node by exchanging the value at the node with the larger of the values at its children.
  - May lead to the subtree at the child not being a heap.
- Recursively fix the children until all of them satisfy the max-heap property.

## MaxHeapify – Example

MaxHeapify( $A, 2$ )



## Procedure MaxHeapify

MaxHeapify( $A, i$ )

- $l \leftarrow \text{left}(i)$
- $r \leftarrow \text{right}(i)$
- if  $l \leq \text{heap-size}[A]$  and  $A[l] > A[i]$
- then  $\text{largest} \leftarrow l$
- else  $\text{largest} \leftarrow i$
- if  $r \leq \text{heap-size}[A]$  and  $A[r] > A[\text{largest}]$
- then  $\text{largest} \leftarrow r$
- if  $\text{largest} \neq i$
- then exchange  $A[i] \leftrightarrow A[\text{largest}]$
- MaxHeapify( $A, \text{largest}$ )

Assumption:  
Left( $i$ ) and Right( $i$ )  
are max-heaps.

## Running Time for MaxHeapify

MaxHeapify( $A, i$ )

- $l \leftarrow \text{left}(i)$
- $r \leftarrow \text{right}(i)$
- if  $l \leq \text{heap-size}[A]$  and  $A[l] > A[i]$
- then  $\text{largest} \leftarrow l$
- else  $\text{largest} \leftarrow i$
- if  $r \leq \text{heap-size}[A]$  and  $A[r] > A[\text{largest}]$
- then  $\text{largest} \leftarrow r$
- if  $\text{largest} \neq i$
- then exchange  $A[i] \leftrightarrow A[\text{largest}]$
- MaxHeapify( $A, \text{largest}$ )

Time to fix node  $i$  and  
its children =  $\Theta(1)$

PLUS

Time to fix the  
subtree rooted at one  
of  $i$ 's children =  
 $T(\text{size of subtree at } \text{largest})$

## Running Time for MaxHeapify( $A, n$ )

- $T(n) = T(\text{largest}) + \Theta(1)$
- $\text{largest} \leq 2n/3$  (worst case occurs when the last row of tree is exactly half full)
- $T(n) \leq T(2n/3) + \Theta(1) \Rightarrow T(n) = O(\lg n)$
- Alternately, MaxHeapify takes  $O(h)$  where  $h$  is the height of the node where MaxHeapify is applied

## Building a heap

- Use *MaxHeapify* to convert an array  $A$  into a max-heap.
- How?
- Call MaxHeapify on each element in a bottom-up manner.

BuildMaxHeap( $A$ )

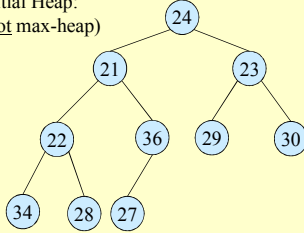
- $\text{heap-size}[A] \leftarrow \text{length}[A]$
- for  $i \leftarrow \lfloor \text{length}[A]/2 \rfloor$  downto 1
- do MaxHeapify( $A, i$ )

## BuildMaxHeap – Example

Input Array:

24 21 23 22 36 29 30 34 28 27

Initial Heap:  
(not max-heap)



## BuildMaxHeap – Example

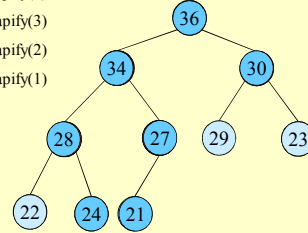
MaxHeapify( $\lfloor 10/2 \rfloor = 5$ )

MaxHeapify(4)

MaxHeapify(3)

MaxHeapify(2)

MaxHeapify(1)



## Correctness of BuildMaxHeap

- ♦ **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.
- ♦ **Initialization:**
  - » Before first iteration  $i = \lfloor n/2 \rfloor$
  - » Nodes  $\lfloor n/2 \rfloor + 1, \lfloor n/2 \rfloor + 2, \dots, n$  are leaves and hence roots of max-heaps.
- ♦ **Maintenance:**
  - » By LI, subtrees at children of node  $i$  are max heaps.
  - » Hence, MaxHeapify( $i$ ) renders node  $i$  a max heap root (while preserving the max heap root property of higher-numbered nodes).
  - » Decrementing  $i$  reestablishes the loop invariant for the next iteration.

## Running Time of BuildMaxHeap

- ♦ **Loose upper bound:**
  - » Cost of a MaxHeapify call  $\times$  No. of calls to MaxHeapify
  - »  $O(\lg n) \times O(n) = O(n \lg n)$
- ♦ **Tighter bound:**
  - » Cost of a call to MaxHeapify at a node depends on the height,  $h$ , of the node –  $O(h)$ .
  - » Height of most nodes smaller than  $n$ .
  - » Height of nodes  $h$  ranges from 0 to  $\lfloor \lg n \rfloor$ .
  - » No. of nodes of height  $h$  is  $\lceil n/2^{h+1} \rceil$

## Running Time of BuildMaxHeap

Tighter Bound for  $T(\text{BuildMaxHeap})$

$T(\text{BuildMaxHeap})$

$$\sum_{h=0}^{\lfloor \lg n \rfloor} \left\lceil \frac{n}{2^{h+1}} \right\rceil O(h)$$

$$= O\left(n \sum_{h=0}^{\lfloor \lg n \rfloor} \frac{h}{2^h}\right)$$

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

$$\sum_{h=0}^{\lfloor \lg n \rfloor} \frac{h}{2^h} \leq \sum_{h=0}^{\infty} \frac{h}{2^h}, \quad x = 1/2 \text{ in (A.8)}$$

$$= \frac{1/2}{(1 - 1/2)^2} = 2$$

Can build a heap from an unordered array in linear time

## Heapsort

- Sort by maintaining the as yet unsorted elements as a max-heap.
- Start by building a max-heap on all elements in  $A$ .
  - » Maximum element is in the root,  $A[1]$ .
- Move the maximum element to its correct final position.
  - » Exchange  $A[1]$  with  $A[n]$ .
- Discard  $A[n]$  – it is now sorted.
  - » Decrement heap-size[ $A$ ].
- Restore the max-heap property on  $A[1..n-1]$ .
  - » Call MaxHeapify( $A, 1$ ).
- Repeat until heap-size[ $A$ ] is reduced to 2.

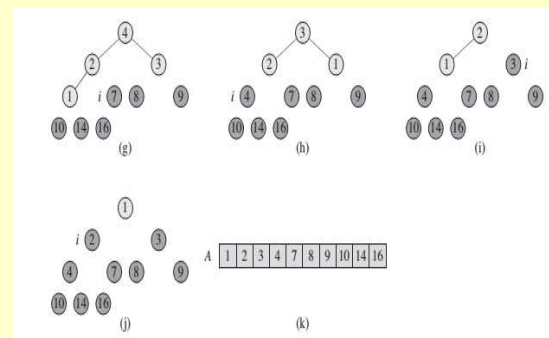
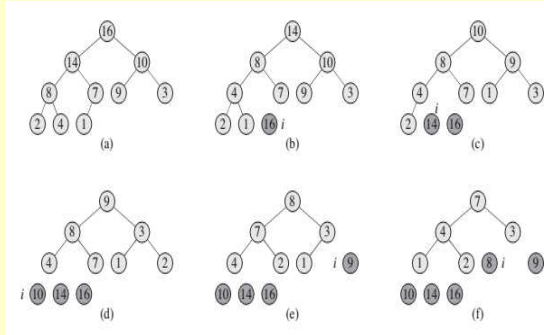
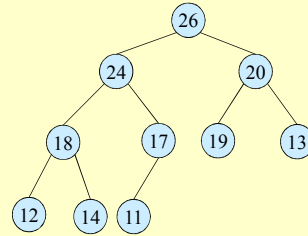
## Heapsort(*A*)

*HeapSort(A)*

1. Build-Max-Heap(*A*)
2. **for**  $i \leftarrow \text{length}[A]$  **downto** 2
3.     **do** exchange  $A[1] \leftrightarrow A[i]$
4.          $\text{heap-size}[A] \leftarrow \text{heap-size}[A] - 1$
5.         MaxHeapify(*A*, 1)

## Heapsort – Example

26	24	20	18	17	19	13	12	14	11
1	2	3	4	5	6	7	8	9	10



## Algorithm Analysis

*HeapSort(A)*

1. Build-Max-Heap(*A*)
  2. **for**  $i \leftarrow \text{length}[A]$  **downto** 2
  3.     **do** exchange  $A[1] \leftrightarrow A[i]$
  4.          $\text{heap-size}[A] \leftarrow \text{heap-size}[A] - 1$
  5.         MaxHeapify(*A*, 1)
- ♦ In-place
  - ♦ Not Stable
  - ♦ Build-Max-Heap takes  $O(n)$  and each of the  $n-1$  calls to Max-Heapify takes time  $O(\lg n)$ .
  - ♦ Therefore,  $T(n) = O(n \lg n)$

## Heap Procedures for Sorting

- ♦ MaxHeapify  $O(\lg n)$
- ♦ BuildMaxHeap  $O(n)$
- ♦ HeapSort  $O(n \lg n)$



## Priority Queue

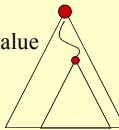
- ♦ Popular & important **application of heaps**.
- ♦ Max and min priority queues.
- ♦ Maintains a **dynamic** set  $S$  of elements.
- ♦ Each set element has a **key** – an associated value.
- ♦ Goal is to **support insertion and extraction efficiently**.
- ♦ **Applications:**
  - » Ready list of processes in operating systems by their priorities – the list is highly dynamic
  - » In event-driven simulators to maintain the list of events to be simulated in order of their time of occurrence.

## Basic Operations

- ♦ Operations on a max-priority queue:
  - » **Insert( $S, x$ )** - inserts the element  $x$  into the set  $S$ 
    - $S \leftarrow S \cup \{x\}$ .
  - » **Maximum( $S$ )** - returns 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  $k$ .
- ♦ **Min-priority queue** supports **Insert, Minimum, Extract-Min, and Decrease-Key**.
- ♦ Heap gives a good compromise between fast insertion but slow extraction and vice versa.

## Heap Property (Max and Min)

- ♦ **Max-Heap**
  - » For **every node** excluding the root, value is **at most** that of its parent:  $A[\text{parent}[i]] \geq A[i]$
- ♦ **Largest element is stored at the root.**
- ♦ In any subtree, no values are **larger** than the value stored at subtree root.
- ♦ **Min-Heap**
  - » For **every node** excluding the root, value is **at least** that of its parent:  $A[\text{parent}[i]] \leq A[i]$
- ♦ **Smallest element is stored at the root.**
- ♦ In any subtree, no values are **smaller** than the value stored at subtree root



## Heap-Extract-Max( $A$ )

Implements the **Extract-Max** operation.

```

Heap-Extract-Max( $A$ )
1. if heap-size[ $A$ ] < 1
2.   then error "heap underflow"
3.  $max \leftarrow A[1]$ 
4.  $A[1] \leftarrow A[\text{heap-size}[A]]$ 
5. heap-size[ $A$ ]  $\leftarrow$  heap-size[ $A$ ] - 1
6. MaxHeapify( $A, 1$ )
7. return max
  
```

Running time : Dominated by the running time of MaxHeapify  
 $= O(\lg n)$

## Heap-Insert( $A, key$ )

```

Heap-Insert( $A, key$ )
1. heap-size[ $A$ ]  $\leftarrow$  heap-size[ $A$ ] + 1
2.  $i \leftarrow$  heap-size[ $A$ ]
3. while  $i > 1$  and  $A[\text{Parent}(i)] < key$ 
4.   do  $A[i] \leftarrow A[\text{Parent}(i)]$ 
5.    $i \leftarrow \text{Parent}(i)$ 
6.  $A[i] \leftarrow key$ 
  
```

Running time is  $O(\lg n)$

The path traced from the new leaf to the root has length  $O(\lg n)$

## Heap-Increase-Key( $A, i, key$ )

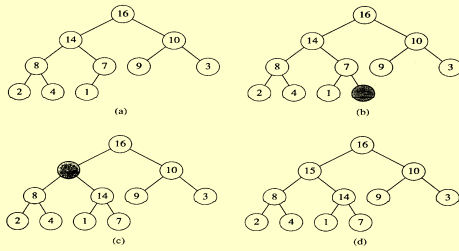
```

Heap-Increase-Key( $A, i, key$ )
1. If  $key < A[i]$ 
2.   then error "new key is smaller than the current key"
3.  $A[i] \leftarrow key$ 
4. while  $i > 1$  and  $A[\text{Parent}[i]] < A[i]$ 
5.   do exchange  $A[i] \leftrightarrow A[\text{Parent}[i]]$ 
6.    $i \leftarrow \text{Parent}[i]$ 
  
```

```

Heap-Insert( $A, key$ )
1. heap-size[ $A$ ]  $\leftarrow$  heap-size[ $A$ ] + 1
2.  $A[\text{heap-size}[A]] \leftarrow -\infty$ 
3. Heap-Increase-Key( $A, \text{heap-size}[A], key$ )
  
```

## Examples



**Figure 7.5** The operation of HEAP-INSERT. (a) The heap of Figure 7.4(a) before we insert a node with key 15. (b) A new leaf is added to the tree. (c) Values on the path from the new leaf to the root are copied down until a place for the key 15 is found. (d) The key 15 is inserted.