

## Unit-V Heap Data Structure

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## What is a “heap”?

- Definitions of **heap**:
  - A large area of memory from which the programmer can allocate blocks as needed, and deallocate them (or allow them to be garbage collected) when no longer needed
  - A balanced, left-justified binary tree in which no node has a value greater than the value in its parent
- These two definitions have little in common
- Heapsort uses the second definition

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## Why study Heapsort?

- It is a well-known, traditional sorting algorithm you will be expected to know
- Heapsort is *always*  $O(n \log n)$ 
  - Quicksort is usually  $O(n \log n)$  but in the worst case slows to  $O(n^2)$
  - Quicksort is generally faster, but Heapsort is better in time-critical applications
- Heapsort is a *really cool* algorithm!

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## Balanced binary trees

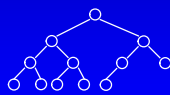
- Recall:
  - The **depth** of a **node** is its distance from the root
  - The **depth** of a **tree** is the depth of the deepest node
- A binary tree of depth  **$n$**  is **balanced** if all the nodes at depths **0** through  **$n-2$**  have two children



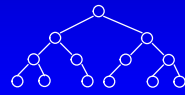
4

## Left-justified binary trees

- A balanced binary tree is **left-justified** if:
  - all the leaves are at the same depth, or
  - all the leaves at depth  $n+1$  are to the left of all the nodes at depth  $n$



Left-justified



Not left-justified

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## Heaps

A **heap** is a certain kind of complete binary tree.

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## Heaps

A **heap** is a certain kind of complete binary tree.

Root



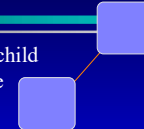
When a complete binary tree is built, its first node must be the root.

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## Heaps

Complete binary tree.

Left child of the root

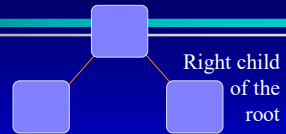


The second node is always the left child of the root.

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## Heaps

Complete  
binary tree.

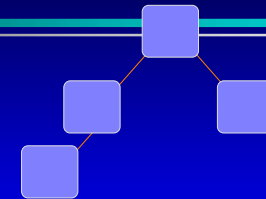


The third node is  
always the right child  
of the root.

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## Heaps

Complete  
binary tree.

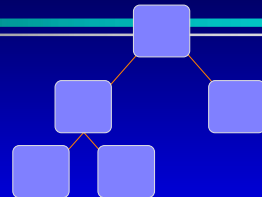


The next nodes  
always fill the next  
level from left-to-right.

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## Heaps

Complete  
binary tree.

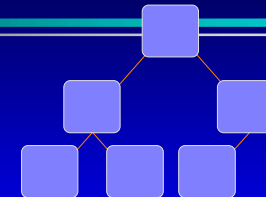


The next nodes  
always fill the next  
level from left-to-right.

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## Heaps

Complete  
binary tree.

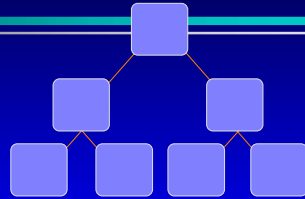


The next nodes  
always fill the next  
level from left-to-right.

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## Heaps

Complete  
binary tree.

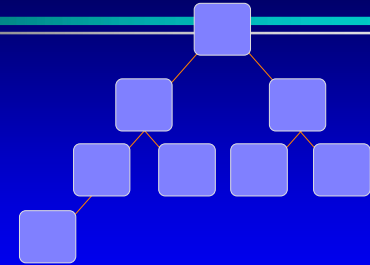


The next nodes  
always fill the next  
level from left-to-right.

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## Heaps

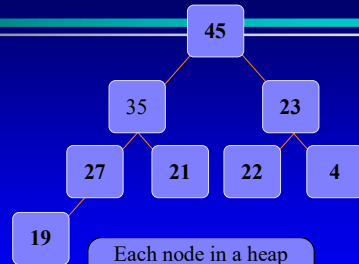
Complete  
binary tree.



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## Heaps

A heap is a  
**certain** kind  
of complete  
binary tree.

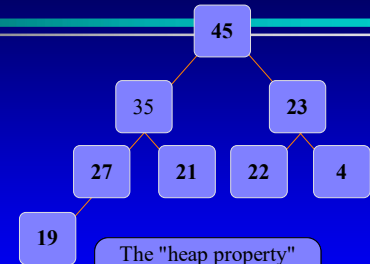


Each node in a heap  
contains a key that  
can be compared to  
other nodes' keys.

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## Heaps

A heap is a  
**certain** kind  
of complete  
binary tree.

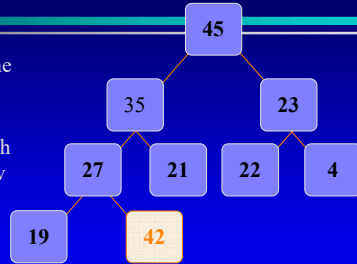


The "heap property"  
requires that each  
node's key is  $\geq$  the  
keys of its children

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## Adding a Node to a Heap

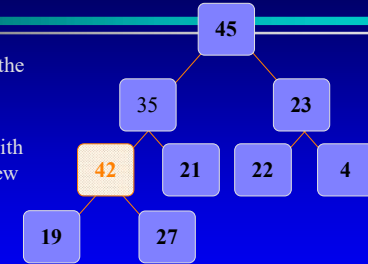
- Put the new node in the next available spot.
- Push the new node upward, swapping with its parent until the new node reaches an acceptable location.



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## Adding a Node to a Heap

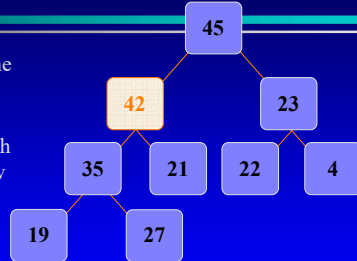
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## Adding a Node to a Heap

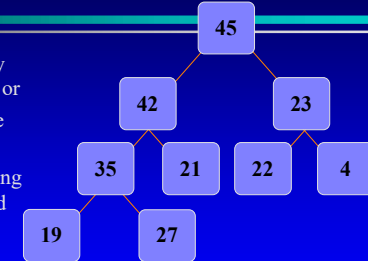
- Put the new node in the next available spot.
- Push the new node upward, swapping with its parent until the new node reaches an acceptable location.



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## Adding a Node to a Heap

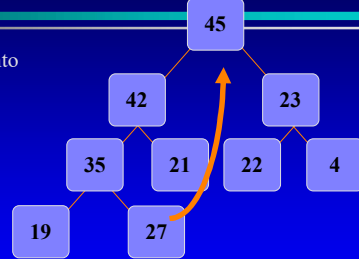
- The parent has a key that is  $\geq$  new node, or
- The node reaches the root.
- The process of pushing the new node upward is called **reheapification upward**.



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## Removing the Top of a Heap

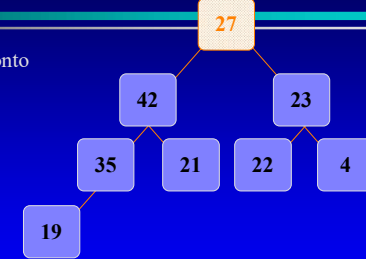
- Move the last node onto the root.



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## Removing the Top of a Heap

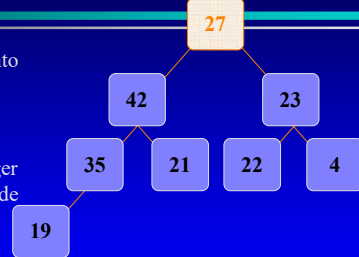
- Move the last node onto the root.



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## Removing the Top of a Heap

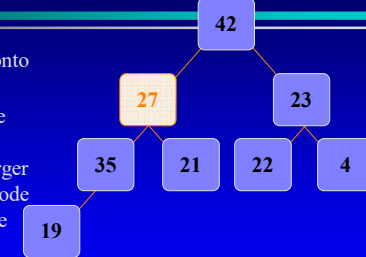
- Move the last node onto the root.
- Push the out-of-place node downward, swapping with its larger child until the new node reaches an acceptable location.



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## Removing the Top of a Heap

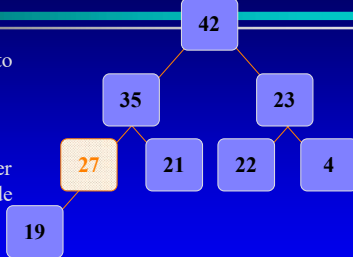
- Move the last node onto the root.
- Push the out-of-place node downward, swapping with its larger child until the new node reaches an acceptable location.



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## Removing the Top of a Heap

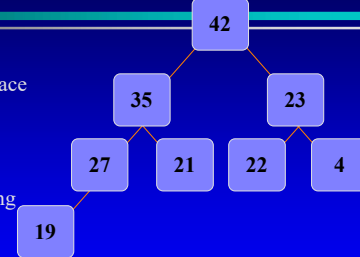
- Move the last node onto the root.
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## Removing the Top of a Heap

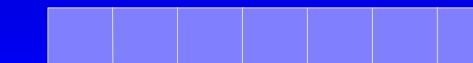
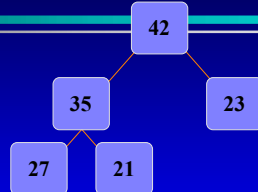
- The children all have keys  $\leq$  the out-of-place node, or
- The node reaches the leaf.
- The process of pushing the new node downward is called reheapification downward.



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## Implementing a Heap

- We will store the data from the nodes in a partially-filled array.

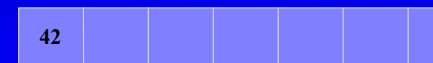
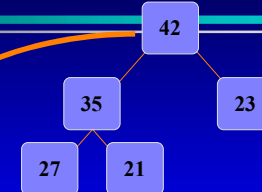


An array of data

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## Implementing a Heap

- Data from the root goes in the first location of the array.

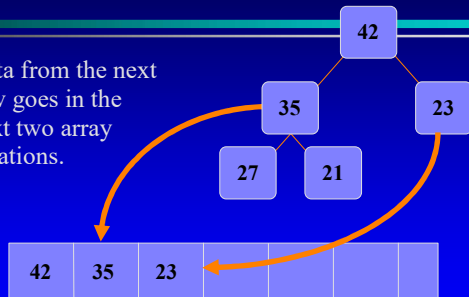


An array of data

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## Implementing a Heap

- Data from the next row goes in the next two array locations.

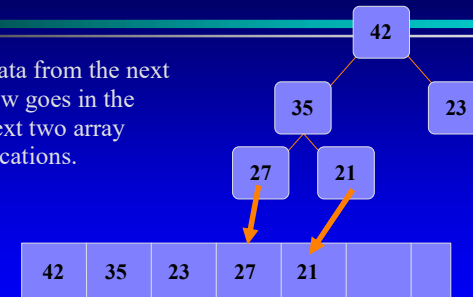


An array of data

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## Implementing a Heap

- Data from the next row goes in the next two array locations.

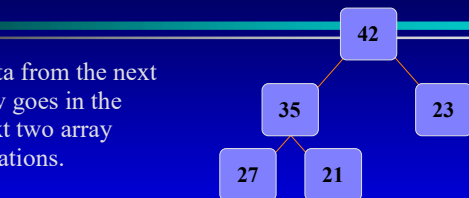


An array of data

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## Implementing a Heap

- Data from the next row goes in the next two array locations.



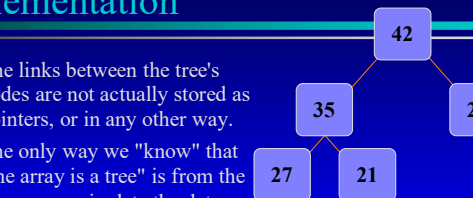
An array of data

We don't care what's in this part of the array.

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## Important Points about the Implementation

- The links between the tree's nodes are not actually stored as pointers, or in any other way.
- The only way we "know" that "the array is a tree" is from the way we manipulate the data.



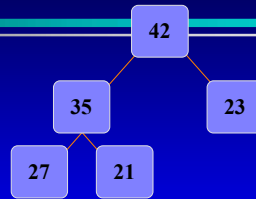
An array of data

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## Important Points about the Implementation

- If you know the index of a node, then it is easy to figure out the indexes of that node's parent and children. Formulas are given in the book.



42	35	23	27	21		
[1]	[2]	[3]	[4]	[5]		

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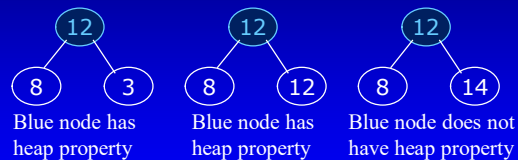
## Summary

- A heap is a complete binary tree, where the entry at each node is greater than or equal to the entries in its children.
- To add an entry to a heap, place the new entry at the next available spot, and perform a reheapification upward.
- To remove the biggest entry, move the last node onto the root, and perform a reheapification downward.

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## The heap property

- A node has the **heap property** if the value in the node is as large as or larger than the values in its children

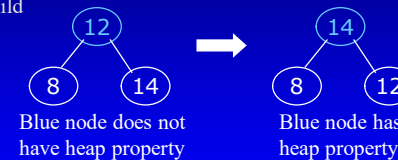


- All leaf nodes automatically have the heap property
- A binary tree is a **heap** if *all* nodes in it have the heap property

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## siftUp

- Given a node that does not have the heap property, you can give it the heap property by exchanging its value with the value of the larger child

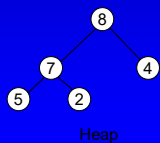


- This is sometimes called **sifting up**
- Notice that the child may have *lost* the heap property

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## The Heap Data Structure

- Def: A **heap** is a nearly complete binary tree with the following two properties:
  - Structural property:** all levels are full, except possibly the last one, which is filled from left to right
  - Order (heap) property:** for any node  $x$   
 $\text{Parent}(x) \geq x$



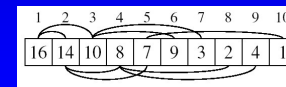
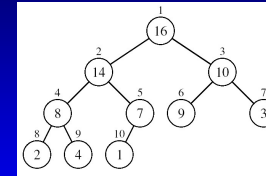
From the heap property, it follows that:  
 "The root is the maximum element of the heap!"

A heap is a binary tree that is filled in order

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## Array Representation of Heaps



- A heap can be stored as an array  $A$ .
  - Root of tree is  $A[1]$
  - Left child of  $A[i] = A[2i]$
  - Right child of  $A[i] = A[2i + 1]$
  - Parent of  $A[i] = A[\lfloor i/2 \rfloor]$
  - $\text{Heapsize}[A] \leq \text{length}[A]$
- The elements in the subarray  $A[(\lfloor n/2 \rfloor + 1) .. n]$  are leaves

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## Heap Types

- Max-heaps** (largest element at root), have the *max-heap property*:

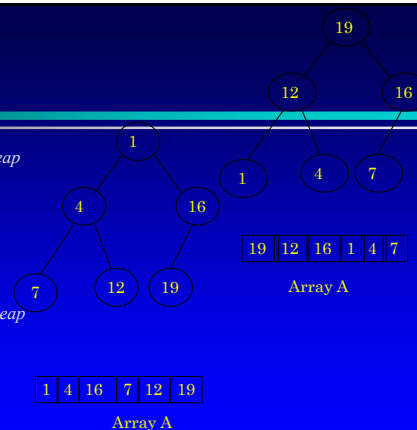
- for all nodes  $i$ , excluding the root:

$$A[\text{PARENT}(i)] \geq A[i]$$

- Min-heaps** (smallest element at root), have the *min-heap property*:

- for all nodes  $i$ , excluding the root:

$$A[\text{PARENT}(i)] \leq A[i]$$



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## Adding/Deleting Nodes

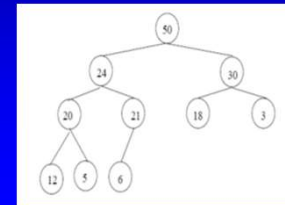
- New nodes are always inserted at the bottom level (left to right)
- Nodes are removed from the bottom level (right to left)

### Algorithm- Insertion

- Add the new element to the next available position at the lowest level
- Restore the max-heap property if violated
  - General strategy is percolate up (or bubble up): if the parent of the element is smaller than the element, then interchange the parent and child.

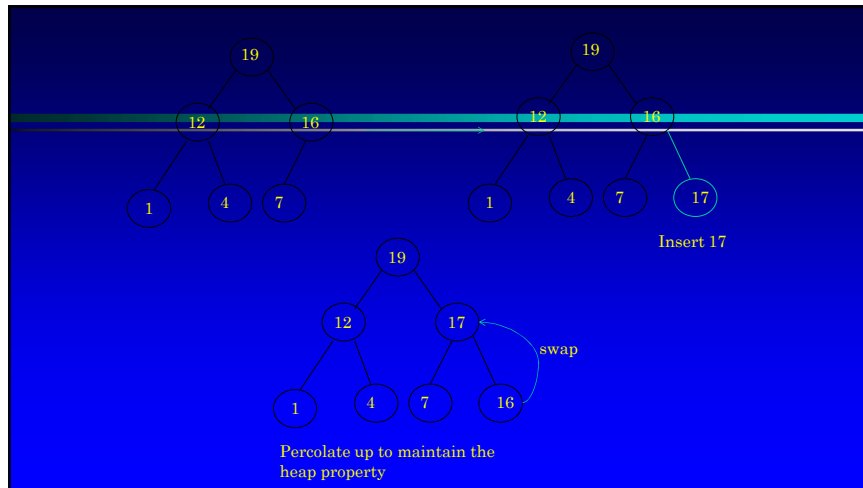
OR

- Restore the min-heap property if violated
- General strategy is percolate up (or bubble up): if the parent of the element is larger than the element, then interchange the parent and child.



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## Deletion

- ❑ Delete max
  - ❑ Copy the last number to the root ( overwrite the maximum element stored there ).
  - ❑ Restore the max heap property by percolate down.
- ❑ Delete min
  - ❑ Copy the last number to the root ( overwrite the minimum element stored there ).
  - ❑ Restore the min heap property by percolate down.

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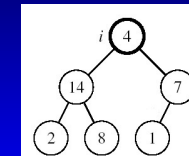
## Operations on Heaps

- ❑ Maintain/Restore the max-heap property
  - ❑ MAX-HEAPIFY
- ❑ Create a max-heap from an unordered array
  - ❑ BUILD-MAX-HEAP
- ❑ Sort an array in place
  - ❑ HEAPSORT
- ❑ Priority queues

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## Maintaining the Heap Property

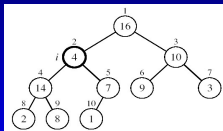
- ❑ Suppose a node is smaller than a child
  - ❑ Left and Right subtrees of  $i$  are max-heaps
- ❑ To eliminate the violation:
  - ❑ Exchange with larger child
  - ❑ Move down the tree
  - ❑ Continue until node is not smaller than children



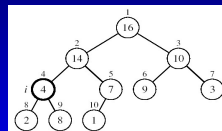
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## Example

MAX-HEAPIFY(A, 2, 10)

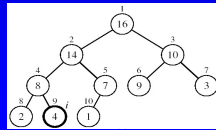


A[2]  $\leftrightarrow$  A[4]



A[4] violates the heap property

A[4]  $\leftrightarrow$  A[9]



Heap property restored

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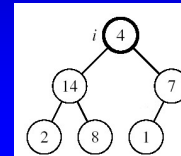
## Maintaining the Heap Property

### Assumptions:

- Left and Right subtrees of  $i$  are max-heaps
- $A[i]$  may be smaller than its children

Alg: MAX-HEAPIFY(A,  $i$ ,  $n$ )

1.  $l \leftarrow \text{LEFT}(i)$
2.  $r \leftarrow \text{RIGHT}(i)$
3. if  $l \leq n$  and  $A[l] > A[i]$
4.   then  $\text{largest} \leftarrow l$
5.   else  $\text{largest} \leftarrow i$
6. if  $r \leq n$  and  $A[r] > A[\text{largest}]$
7.   then  $\text{largest} \leftarrow r$
8. if  $\text{largest} \neq i$
9.   then exchange  $A[i] \leftrightarrow A[\text{largest}]$
10.   MAX-HEAPIFY(A,  $\text{largest}$ ,  $n$ )



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## MAX-HEAPIFY Running Time

### Intuitively:

It traces a path from the root to a leaf (longest path length  $h$ )  
At each level, it makes exactly 2 comparisons  
Total number of comparisons is  $2h$   
Running time is  $O(h)$  or  $O(\lg n)$

- Running time of MAX-HEAPIFY is  $O(\lg n)$
- Can be written in terms of the height of the heap, as being  $O(h)$
- Since the height of the heap is  $\lfloor \lg n \rfloor$

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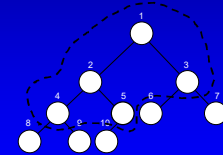
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## Building a Heap

- Convert an array  $A[1 \dots n]$  into a max-heap ( $n = \text{length}[A]$ )
- The elements in the subarray  $A[(\lfloor n/2 \rfloor + 1) \dots n]$  are leaves
- Apply MAX-HEAPIFY on elements between 1 and  $\lfloor n/2 \rfloor$

10 BUILD-MAX-HEAP(A)

1.  $n \leftarrow \text{length}[A]$
2. for  $i \leftarrow \lfloor n/2 \rfloor$  downto 1
3.   do MAX-HEAPIFY(A,  $i$ ,  $n$ )



A: 4 1 3 2 16 9 10 14 8 7

48

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Example:

A

4	1	3	2	16	9	10	14	8	7
---	---	---	---	----	---	----	----	---	---

Alg: MAX-HEAPIFY(A, i, n)

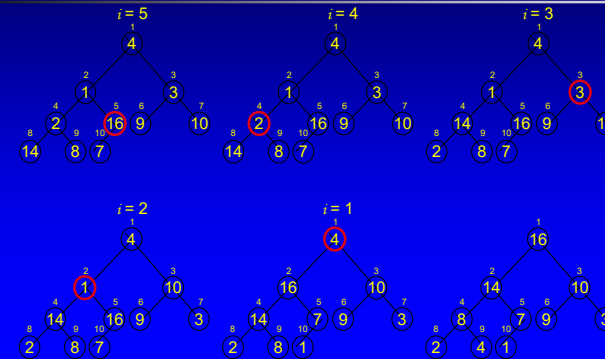
1.  $l \leftarrow \text{LEFT}(i)$
2.  $r \leftarrow \text{RIGHT}(i)$
3. if  $l \leq n$  and  $A[l] > A[i]$
4.   then  $\text{largest} \leftarrow l$
5.   else  $\text{largest} \leftarrow i$
6. if  $r \leq n$  and  $A[r] > A[\text{largest}]$
7.   then  $\text{largest} \leftarrow r$
8. if  $\text{largest} \neq i$
9.   then exchange  $A[i] \leftrightarrow A[\text{largest}]$
10.   MAX-HEAPIFY(A, largest, n)

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Example:

A

4	1	3	2	16	9	10	14	8	7
---	---	---	---	----	---	----	----	---	---



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## Running Time of BUILD MAX HEAP

Alg: BUILD-MAX-HEAP(A)

1.  $n = \text{length}[A]$
2. for  $i \leftarrow \lfloor n/2 \rfloor$  downto 1
3.   do MAX-HEAPIFY(A, i, n)

 $O(\lg n) \quad \left. \vphantom{\begin{matrix} 1 \\ 2 \\ 3 \end{matrix}} \right\} O(n)$ 
 $\Rightarrow$  Running time:  $O(n \lg n)$ 

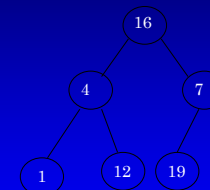
□ This is not an asymptotically tight upper bound

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Example: Convert the following array to a heap

16	4	7	1	12	19
----	---	---	---	----	----

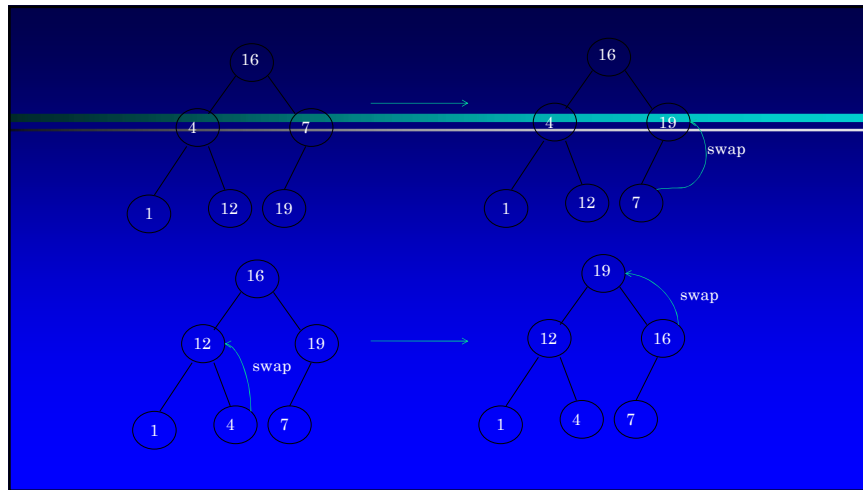
Picture the array as a complete binary tree:



Alg: MAX-HEAPIFY(A, i, n)

1.  $l \leftarrow \text{LEFT}(i)$
2.  $r \leftarrow \text{RIGHT}(i)$
3. if  $l \leq n$  and  $A[l] > A[i]$
4.   then  $\text{largest} \leftarrow l$
5.   else  $\text{largest} \leftarrow i$
6. if  $r \leq n$  and  $A[r] > A[\text{largest}]$
7.   then  $\text{largest} \leftarrow r$
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9.   then exchange  $A[i] \leftrightarrow A[\text{largest}]$
10.   MAX-HEAPIFY(A, largest, n)

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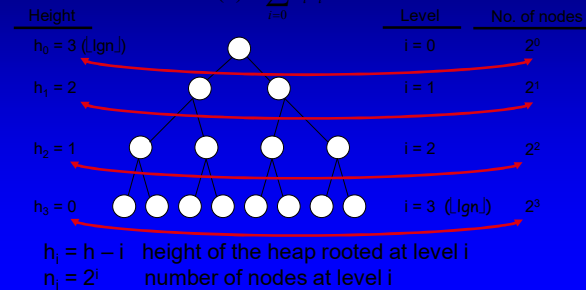
53

## Running Time of BUILD MAX HEAP

- HEAPIFY takes  $O(h) \Rightarrow$  the cost of HEAPIFY on a node  $i$  is proportional to the height of the node  $i$  in the tree

$$\Rightarrow T(n) = \sum_{i=0}^h n_i h_i$$

$$= \sum_{i=0}^h \approx (\lg(n))$$



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## Running Time of BUILD MAX HEAP

$$T(n) = \sum_{i=0}^h n_i h_i \quad \text{Cost of HEAPIFY at level } i * \text{number of nodes at that level}$$

$$= \sum_{i=0}^h 2^i (h - i) \quad \text{Replace the values of } n_i \text{ and } h_i \text{ computed before}$$

$$= \sum_{i=0}^h \frac{h - i}{2^{h-i}} 2^h \quad \text{Multiply by } 2^h \text{ both at the nominator and denominator and write } 2^i \text{ as } \frac{1}{2^{h-i}}$$

$$= 2^h \sum_{k=0}^h \frac{k}{2^k} \quad \text{Change variables: } k = h - i$$

$$\leq n \sum_{k=0}^{\infty} \frac{k}{2^k} \quad \text{The sum above is smaller than the sum of all elements to } \infty \text{ and } h = \lg n$$

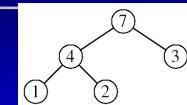
$$= O(n) \quad \text{The sum above is smaller than 2}$$

Running time of BUILD-MAX-HEAP:  $T(n) = O(n)$

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## Heapsort

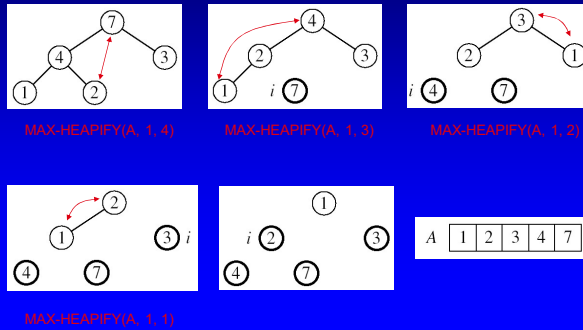
- Goal:
  - Sort an array using heap representations
- Idea:
  - Build a **max-heap** from the array
  - Swap the root (the maximum element) with the last element in the array
  - "Discard" this last node by decreasing the heap size
  - Call MAX-HEAPIFY on the new root
  - Repeat this process until only one node remains



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Example:  $A=[7, 4, 3, 1, 2]$



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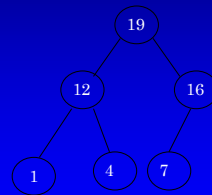
$\mathcal{Alg}$ : HEAPSORT( $A$ )

1. BUILD-MAX-HEAP( $A$ )  $O(n)$
  2. for  $i \leftarrow \text{length}[A]$  downto 2
  3.   do exchange  $A[1] \leftrightarrow A[i]$
  4.   MAX-HEAPIFY( $A, 1, i - 1$ )  $O(\lg n)$
- }  $n-1$  times
- Running time:  $O(n \lg n)$  --- Can be shown to be  $\Theta(n \lg n)$

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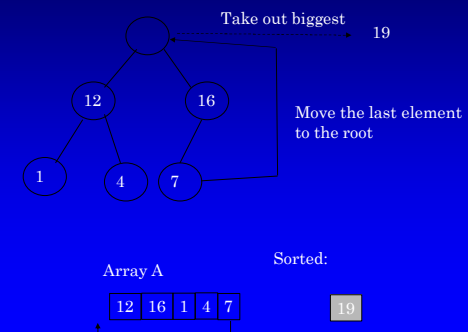
## Heap Sort

- The heapsort algorithm consists of two phases:
  - build a heap from an arbitrary array
  - use the heap to sort the data
- To sort the elements in the decreasing order, use a min heap
- To sort the elements in the increasing order, use a max heap

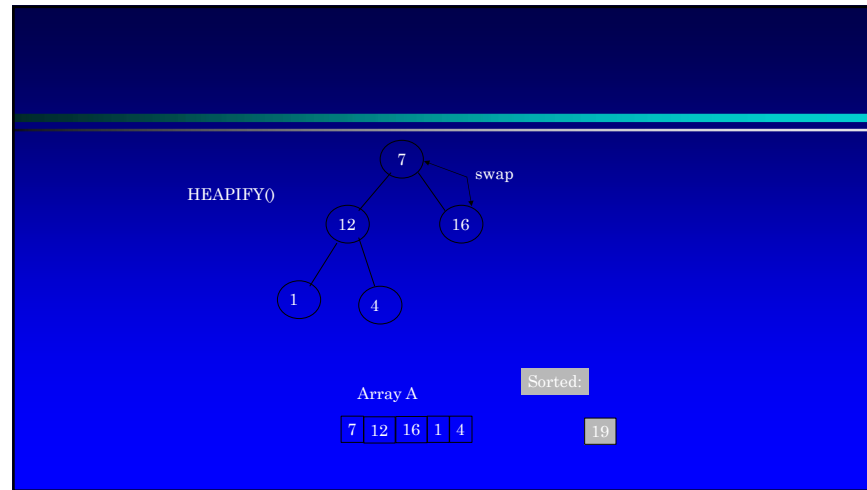


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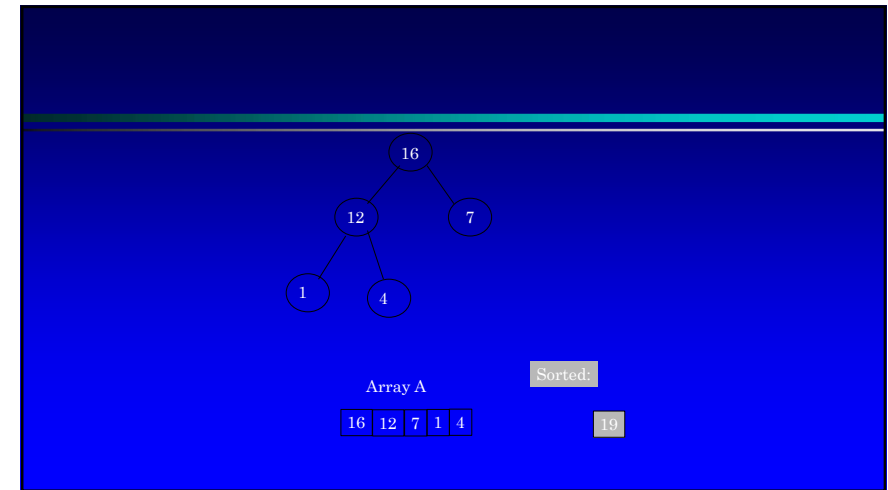
## Example of Heap Sort



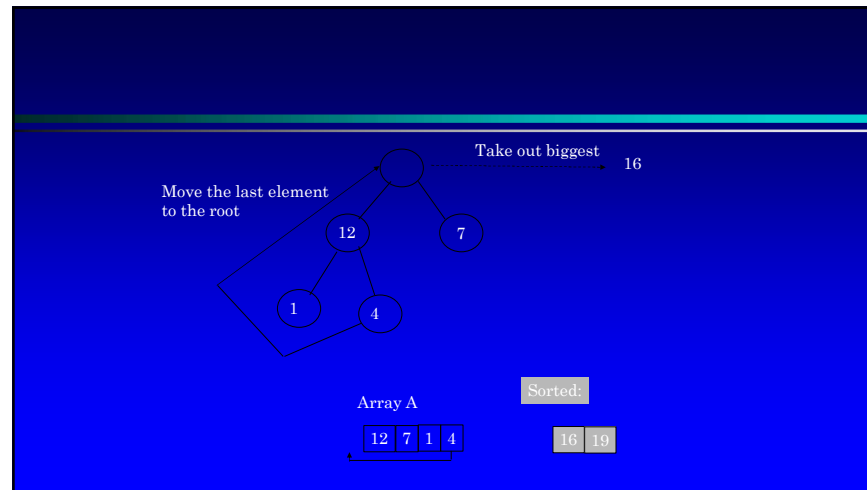
60



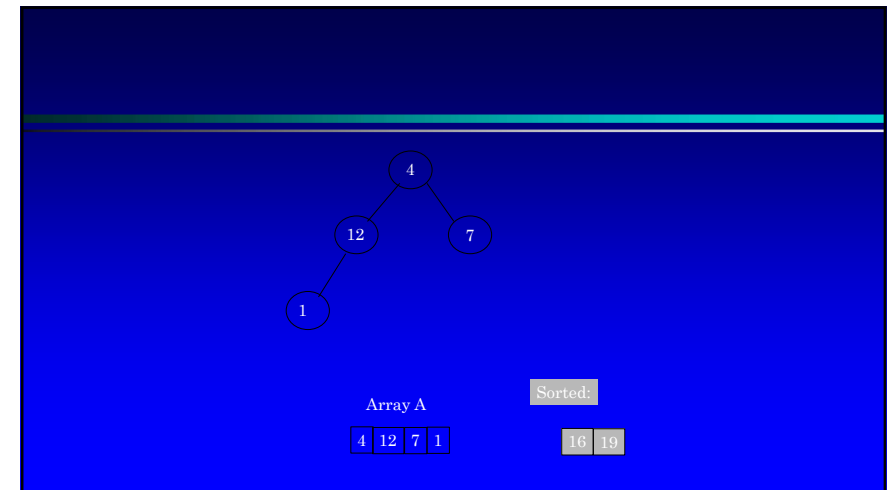
61



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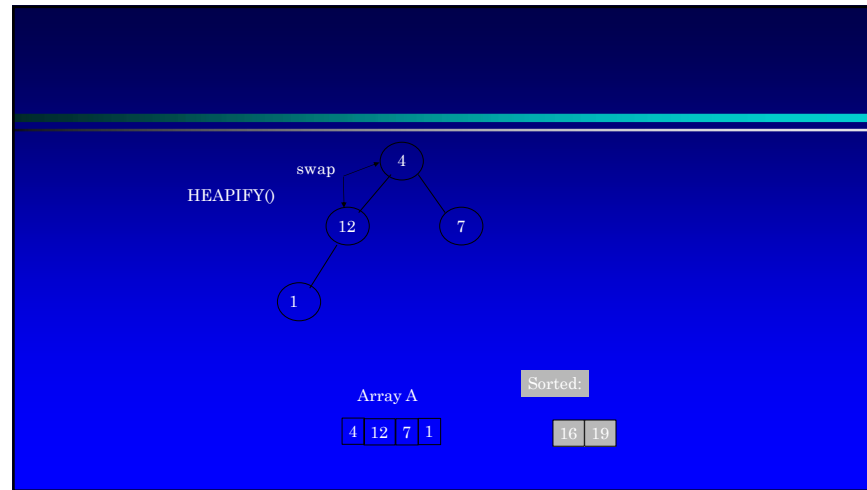


63

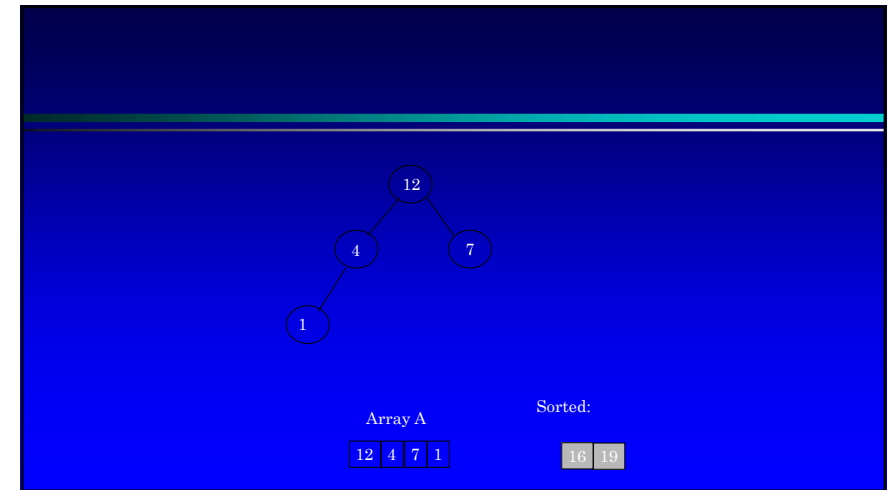


64

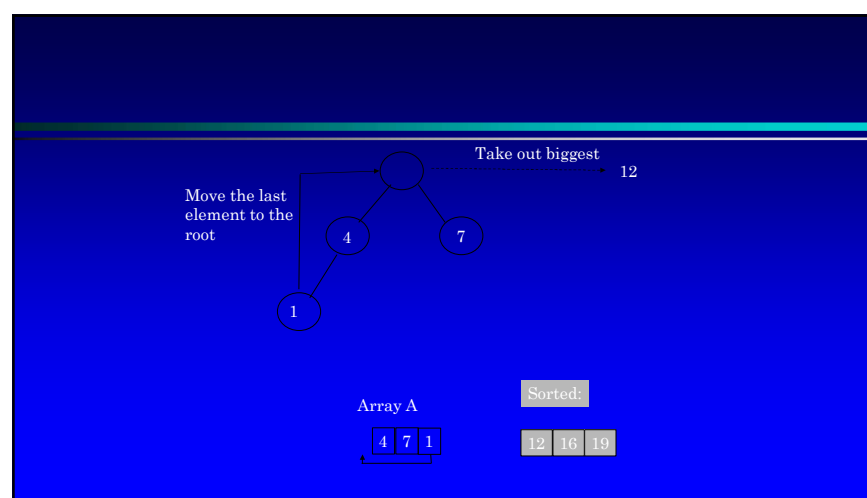




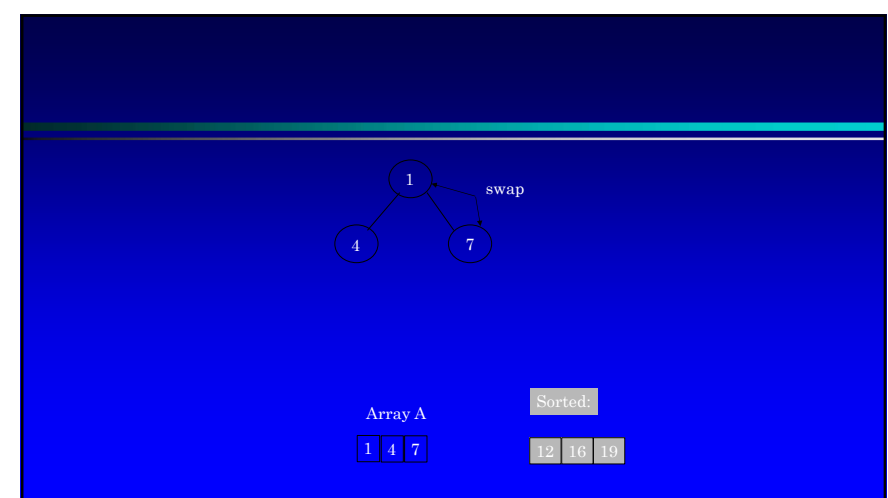
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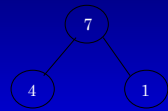
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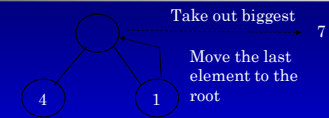
Array A

7	4	1
---	---	---

Sorted:

12	16	19
----	----	----

69



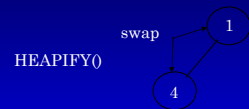
Array A

1	4
---	---

Sorted:

7	12	16	19
---	----	----	----

70



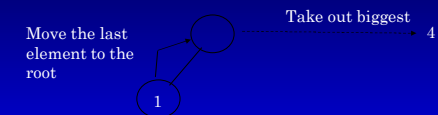
Array A

4	1
---	---

Sorted:

7	12	16	19
---	----	----	----

71



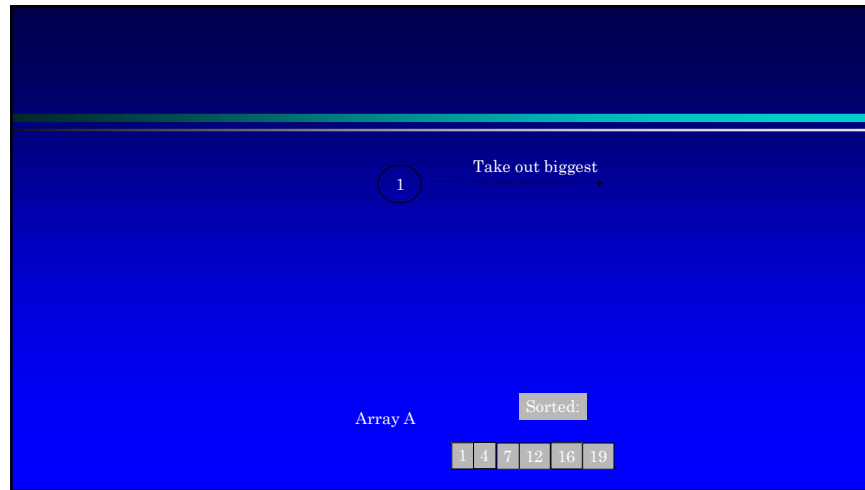
Array A

1
---

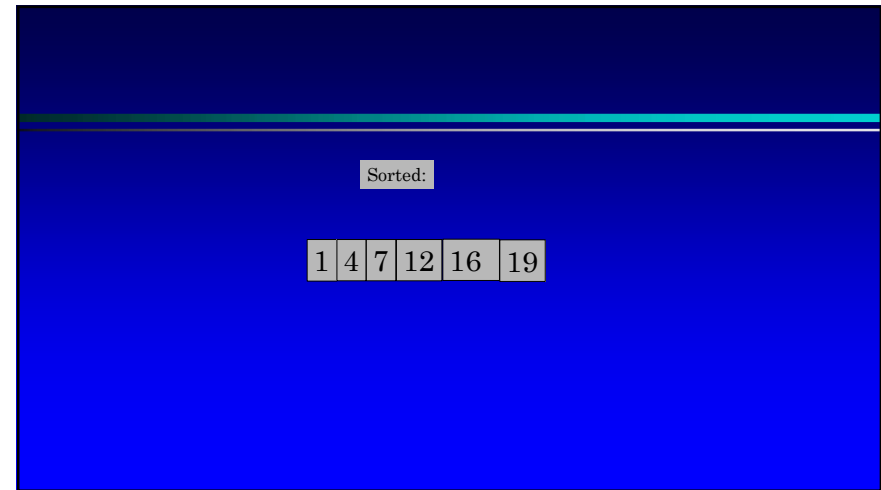
Sorted:

4	7	12	16	19
---	---	----	----	----

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## Time Analysis

- Build Heap Algorithm will run in  $O(n)$  time
- There are  $n-1$  calls to Heapify each call requires  $O(\log n)$  time
- Heap sort program combine Build Heap program and Heapify, therefore it has the running time of  $O(n \log n)$  time
- Total time complexity:  $O(n \log n)$

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## Heapify– Using location 0 in Array

Heapify function to construct a heap

```

void heapify( arr[], n, root)
{
    // largest = root // Initialize largest as root
    // left_child = 2*root + 1 // left = 2*i + 1
    // right_child = 2*root + 2 // right = 2*i + 2
    // If left child is larger than root
    if (left_child < n && arr[left_child] > arr[largest])
        largest = left_child
    // If right child is larger than largest so far
    if (right_child < n && arr[right_child] > arr[largest])
        largest = right_child
    // If largest is not root
    if (largest != root)
    {
        swap(arr[root], arr[largest])
        // Recursively heapify the affected
        // sub-tree
        heapify(arr, n, largest)
    }
}

```

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## Heap Sort– Using location 0 in Array

```

Heap sort algo
void heapSort(arr[], n)
{
    // Build heap (rearrange array)
    for (i = n / 2 - 1 to i >= 0)
        heapify(arr, n, i)
    // One by one extract an element from heap
    for (i=n-1 to 0)
    {
        // Move current root to end
        swap(arr[0], arr[i])
        // call max heapify on the reduced heap
        heapify(arr, i, 0)
    }
}

```

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## Comparison with Quick Sort and Merge Sort

- ❑ Quick sort is typically somewhat faster, due to better cache behavior and other factors, but the worst-case running time for quick sort is  $O(n^2)$ , which is unacceptable for large data sets and can be deliberately triggered given enough knowledge of the implementation, creating a security risk.
- ❑ The quick sort algorithm also requires  $\Omega(\log n)$  extra storage space, making it not a strictly in-place algorithm. This typically does not pose a problem except on the smallest embedded systems, or on systems where memory allocation is highly restricted. Constant space (in-place) variants of quick sort are possible to construct, but are rarely used in practice due to their extra complexity.

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## Comparison with Quick Sort and Merge Sort (cont)

- ❑ Thus, because of the  $O(n \log n)$  upper bound on heap sort's running time and constant upper bound on its auxiliary storage, embedded systems with real-time constraints or systems concerned with security often use heap sort.
- ❑ Heap sort also competes with merge sort, which has the same time bounds, but requires  $\Omega(n)$  auxiliary space, whereas heap sort requires only a constant amount. Heap sort also typically runs more quickly in practice. However, merge sort is simpler to understand than heap sort, is a stable sort, parallelizes better, and can be easily adapted to operate on linked lists and very large lists stored on slow-to-access media such as disk storage or network attached storage. Heap sort shares none of these benefits; in particular, it relies strongly on random access.

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## Possible Application

- ❑ When we want to know the task that carry the highest priority given a large number of things to do
- ❑ Interval scheduling, when we have a lists of certain task with start and finish times and we want to do as many tasks as possible
- ❑ Sorting a list of elements that needs and efficient sorting algorithm

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## Conclusion

- The primary advantage of the heap sort is its efficiency. The execution time efficiency of the heap sort is  $O(n \log n)$ . The memory efficiency of the heap sort, unlike the other  $n \log n$  sorts, is constant,  $O(1)$ , because the heap sort algorithm is not recursive.
- The heap sort algorithm has two major steps. The first major step involves transforming the complete tree into a heap. The second major step is to perform the actual sort by extracting the largest element from the root and transforming the remaining tree into a heap.

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## Priority Queues

### Properties

- Each element is associated with a value (priority)
- The key with the highest (or lowest) priority is extracted first



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## Operations on Priority Queues

- Max-priority queues support the following operations:
  - $\text{INSERT}(S, x)$ : inserts element  $x$  into set  $S$
  - $\text{EXTRACT-MAX}(S)$ : removes and returns element of  $S$  with largest key
  - $\text{MAXIMUM}(S)$ : returns element of  $S$  with largest key
  - $\text{INCREASE-KEY}(S, x, k)$ : increases value of element  $x$ 's key to  $k$  (Assume  $k \geq x$ 's current key value)

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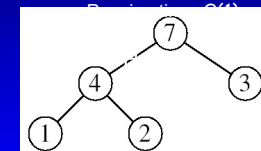
## HEAP-MAXIMUM

Goal:

- Return the largest element of the heap

*Alg:*  $\text{HEAP-MAXIMUM}(A)$

1. **return**  $A[1]$



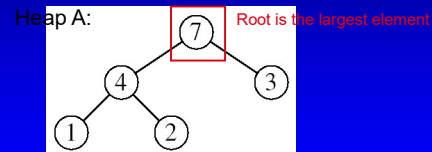
Heap-Maximum(A) returns 7

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## HEAP-EXTRACT-MAX

Goal:

- Extract the largest element of the heap (i.e., return the max value and also remove that element from the heap)

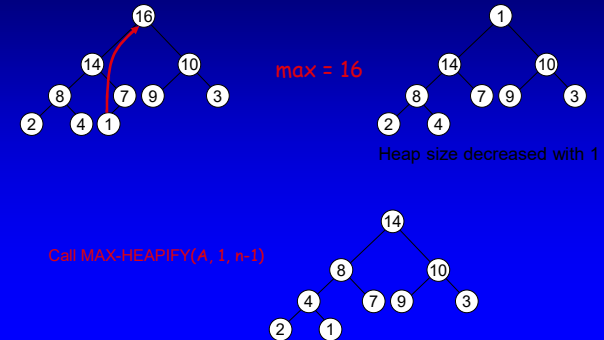


Idea:

- Exchange the root element with the last
- Decrease the size of the heap by 1 element
- Call MAX-HEAPIFY on the new root, on a heap of size  $n-1$

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## Example: HEAP-EXTRACT-MAX

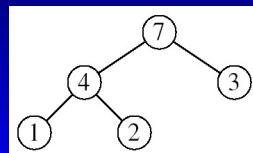


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## HEAP-EXTRACT-MAX

*Alg:* HEAP-EXTRACT-MAX( $A, n$ )

- if  $n < 1$
- then error "heap underflow"
- $\text{max} \leftarrow A[1]$
- $A[1] \leftarrow A[n]$
- MAX-HEAPIFY( $A, 1, n-1$ )
- return max



Running time:  $O(\lg n)$

remakes heap

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## Priority Queue Using Linked List



Remove a key:  $O(1)$

Insert a key:  $O(n)$

Increase key:  $O(n)$

Extract max key:  $O(1)$

Average:  $O(n)$

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## Problems

- (a) What is the maximum number of nodes in a max heap of height  $h$ ?
- (b) What is the maximum number of leaves?
- (c) What is the maximum number of internal nodes?

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## Problems

- Demonstrate, step by step, the operation of Build-Heap on the array

$A = [5, 3, 17, 10, 84, 19, 6, 22, 9]$

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## Problems

- Let  $A$  be a heap of size  $n$ . Give the most efficient algorithm for the following tasks:
- (a) Find the sum of all elements
- (b) Find the sum of the largest  $\lg n$  elements

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