

CS217 - Data Structures & Algorithm Analysis (DSAA)

Lecture #4

HeapSort

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Reading: Chapter 6

➤ Aims of this lecture

- To introduce the **HeapSort** algorithm.
- To show how a **clever data structure**, a **heap**, can lead to a **fast** and **in place** sorting algorithm
 - In place: $O(1)$ additional space.
- To **practice the design and analysis of algorithms**.

➤ Idea behind HeapSort

- Idea:
 - Find the largest element.
 - Move it to the end of the array (put another one in its place).
 - Repeat with remaining elements.
- Like SelectionSort but ...
 - SelectionSort compares lots of elements to find the largest.
 - Can we store knowledge gained from these comparisons for the future?
 - Use this knowledge to make future iterations faster!

➤ Use your imagination...



Photo : Thomas Bresson

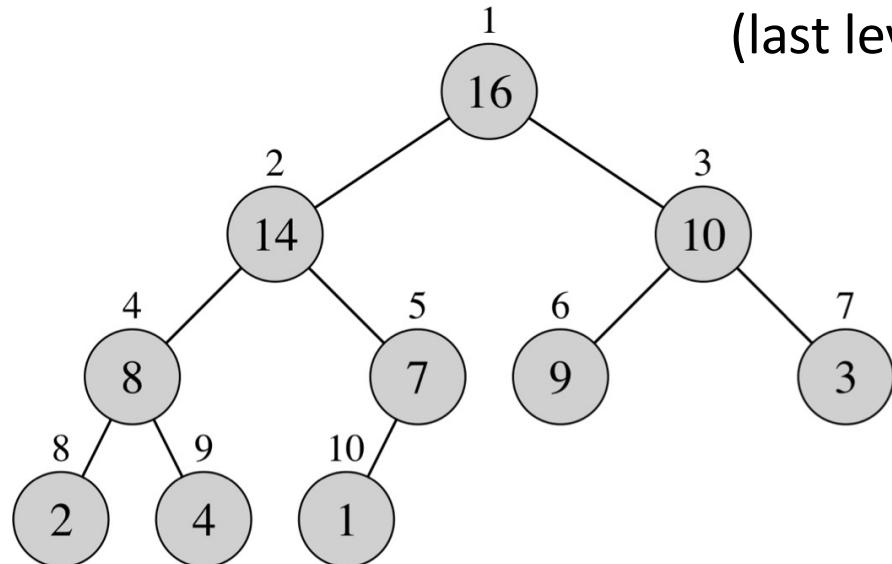
➤ A Heap



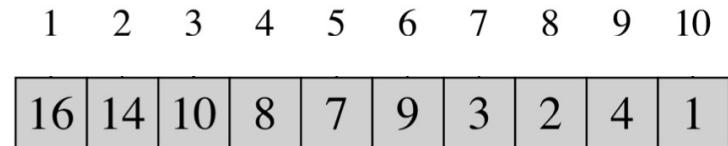
- Essentially an array **imagined** as being a **binary tree**!

- Elements are arranged row by row from left to right.

(last level may be incomplete)



(a)



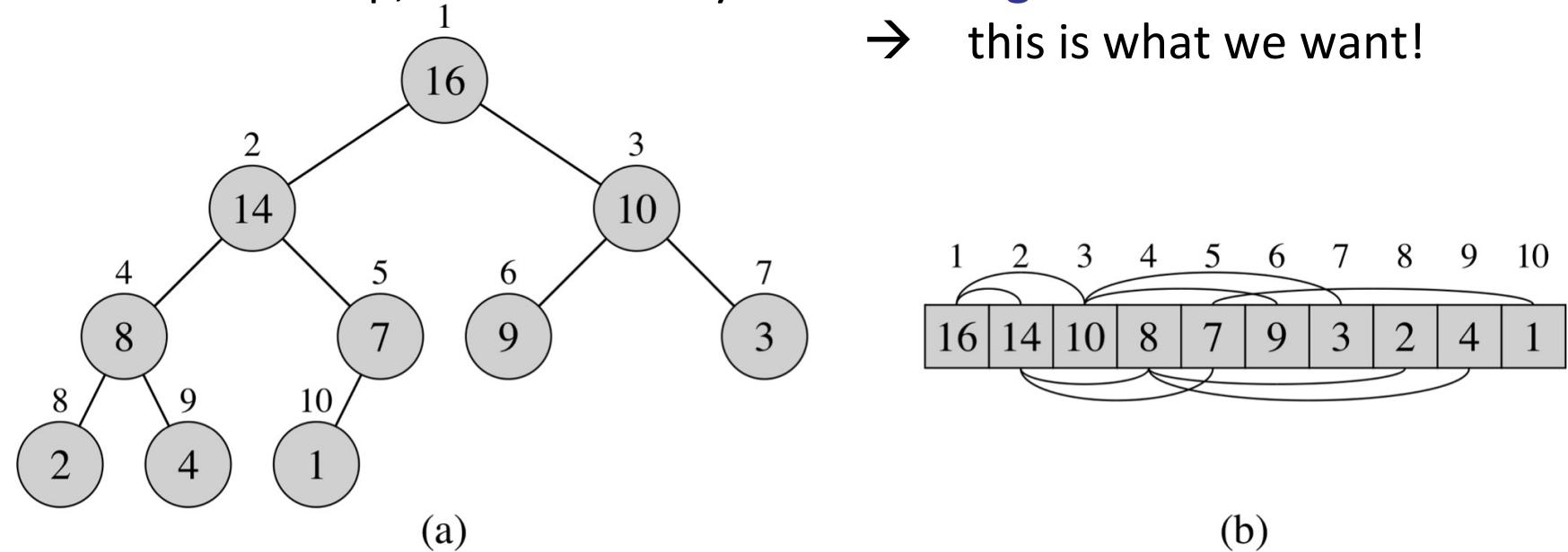
(b)

- Navigate through the array/imaginary tree using these operations:

- $\text{Parent}(i) = \left\lfloor \frac{i}{2} \right\rfloor$ ("floor of $i/2$ "), $\text{Left}(i) = 2i$, $\text{Right}(i) = 2i + 1$

➤ Heap Properties

- **Max-heap property**: for every node other than the root, the parent is no smaller than the node, $A[\text{Parent}(i)] \geq A[i]$.
- In a max-heap, the **root** always stores a **largest element**.
→ this is what we want!



- **Min-heap property**: for every node other than the root, the parent is no larger than the node, $A[\text{Parent}(i)] \leq A[i]$.

➤ Procedures (what do we need)

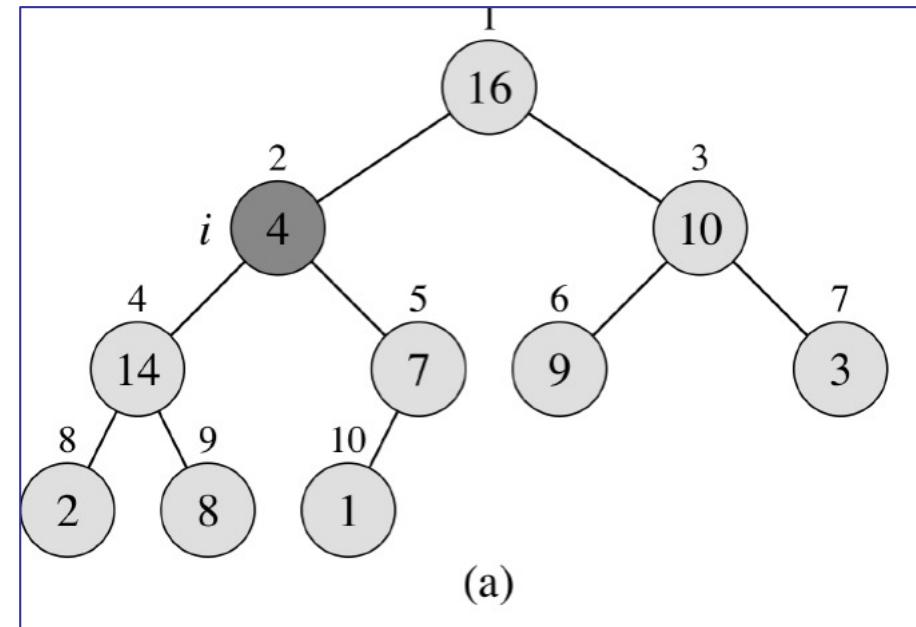
1. **Build-Max-Heap**: produces a Max-Heap from an unordered array
 2. **Max-Heapify**: maintains the max-heap property once the maximum has been removed
 3. **HeapSort**: sorts an array in place
-
- New variable `A.heap-size` indicates how many elements of `A` are stored in a heap: $0 \leq A.\text{heap-size} \leq A.\text{length}$.
 - Decreasing `A.heap-size` by 1 effectively removes the last element from the heap (we imagine a heap without it)
 - There are analogous operations for min-heaps:
Min-Heapify and Build-Min-Heap.

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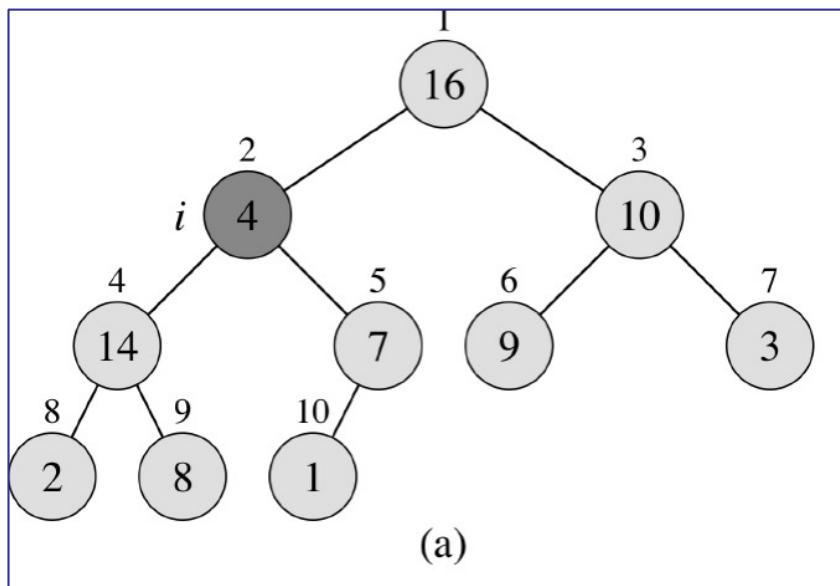
➤ Max-Heapify(A, i)

- Assumes subtrees $\text{Left}(i)$ and $\text{Right}(i)$ are max-heaps, but max-heap property might be violated in root of subtree at i .
 - “Subtree x ”: the part of the tree including x and everything below.
- Lets the value at $A[i]$ “float down” if necessary, to restore max-heap property at i
- At the end of Max-Heapify the subtree at i is a max-heap.



► Max-Heapify: informal and in pseudocode

- Compare $A[i]$ with all existing children
 - If **largest child** is larger than $A[i]$, swap and recurse on child



MAX-HEAPIFY(A, i)

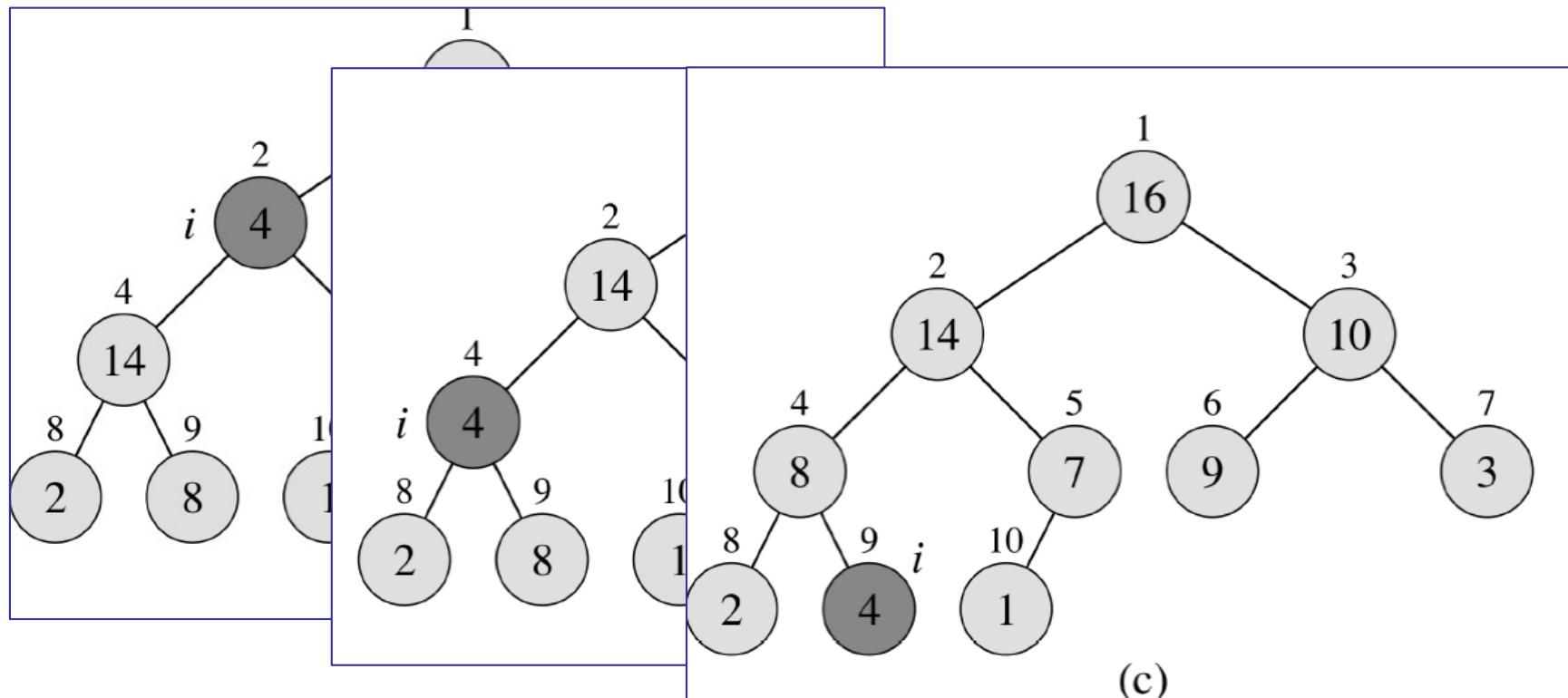
```

1: l = Left(i)
2: r = Right(i)
3: if l  $\leq$  A.heap-size and A[l] > A[i] then
4:     largest = l
5: else
6:     largest = i
7: if r  $\leq$  A.heap-size and A[r] > A[largest] then
8:     largest = r
9: if largest  $\neq$  i then
10:    exchange A[i] with A[largest]
11:    MAX-HEAPIFY(A, largest)

```

➤ Max-Heapify: Example

- Compare $A[i]$ with all existing children
- If **largest child** is larger than $A[i]$, swap and recurse on child

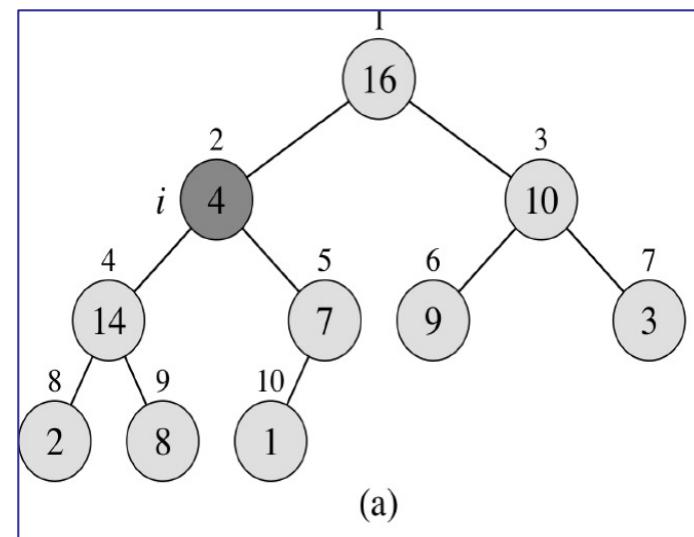


➤ Runtime of Max-Heapify

- Define the **height** of a node as the longest number of simple downward edges from the node to a **leaf**.
- **Leaf**: a node without children.
- Max-Heapify takes constant time, $\Theta(1)$, on each level.
- Running time of Max-Heapify on a node of height h is $O(h)$.
- It's not $\Omega(h)$ as Max-Heapify may stop early, e.g. if heap-property holds at i .
- For leaves $h = 0$ and the time is $O(1)$.

MAX-HEAPIFY(A, i)

```
1:  $l = \text{Left}(i)$ 
2:  $r = \text{Right}(i)$ 
3: if  $l \leq A.\text{heap-size}$  and  $A[l] > A[i]$  then
4:      $\text{largest} = l$ 
5: else
6:      $\text{largest} = i$ 
7: if  $r \leq A.\text{heap-size}$  and  $A[r] > A[\text{largest}]$  then
8:      $\text{largest} = r$ 
9: if  $\text{largest} \neq i$  then
10:    exchange  $A[i]$  with  $A[\text{largest}]$ 
11:    MAX-HEAPIFY( $A, \text{largest}$ )
```



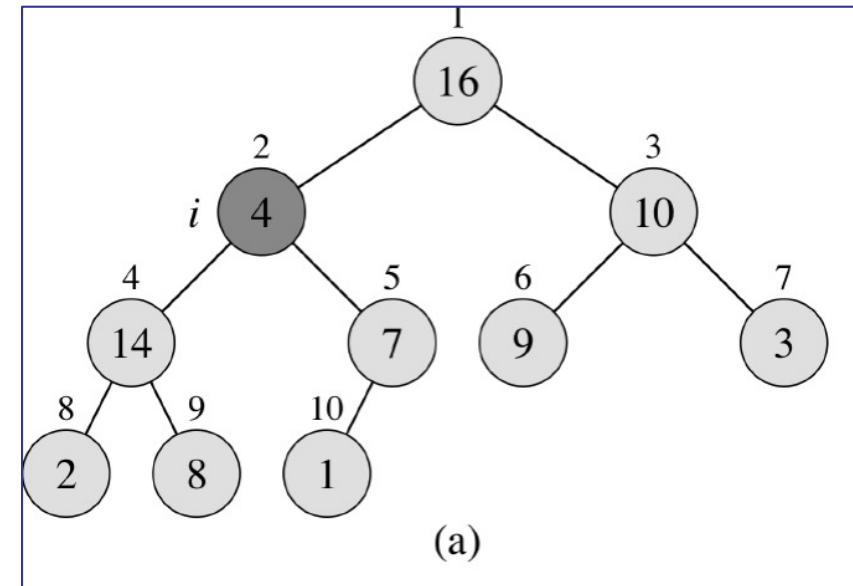
➤ Bounding the height of a heap

- **Claim:** the height of a heap = height of the root is at most $\log n$.
- **Proof:** the number n of elements in a heap of height h is

- Doubling on each level
- At least 1 node on the last level
- Hence in total at least

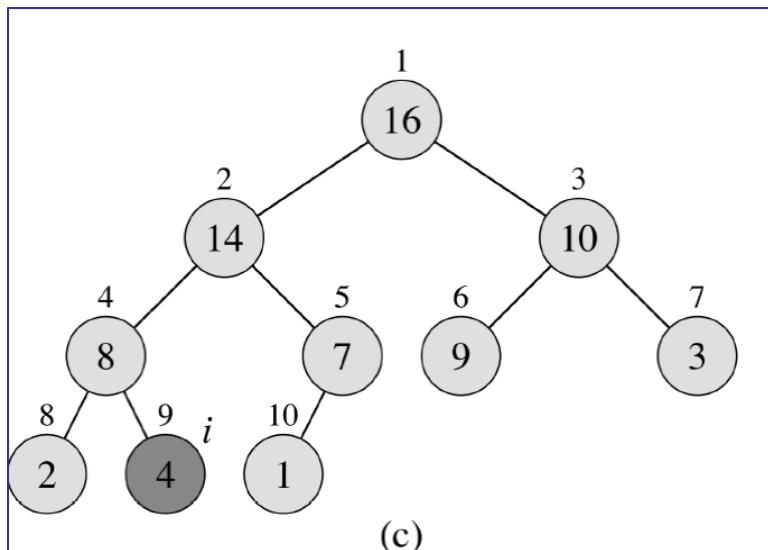
$$1 + 2 + 4 + \dots + 2^{h-1} + 1 = 2^h$$

(we used $\sum_{i=0}^{k-1} 2^i = 2^k - 1$)



- So size and height are related as $n \geq 2^h \Leftrightarrow \log n \geq h$
- “the height of the root is at most $\log n$ ”
- So the runtime of Max-Heapify is $O(\log n)$

► Max-Heapify: Correctness

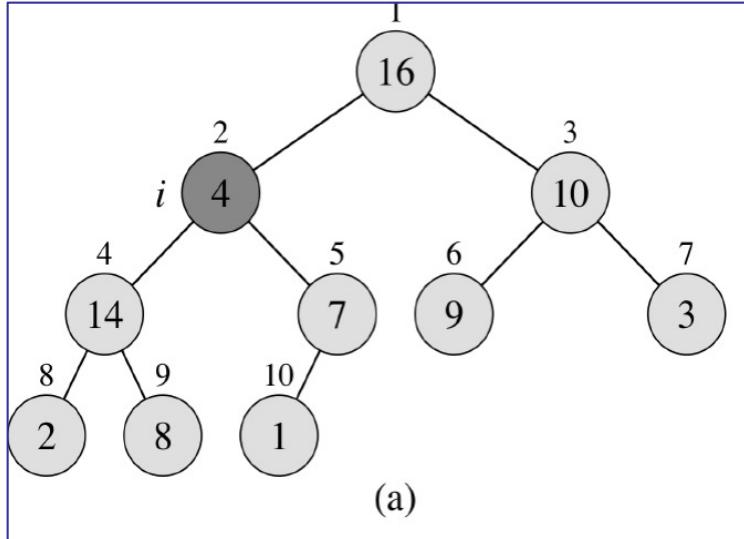


MAX-HEAPIFY(A, i)

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5: else
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```

- By induction (on the height):
- **Base case:** height = 0 (i is a leaf)
- Then $\text{left}(i)$ and $\text{right}(i)$ are larger than $A.\text{heap-size}$ and the algorithm returns a heap!

► Max-Heapify: Correctness



- By induction (on the height):
- **Inductive case:** assume it works for height $h = i - 1$ and show it works for $h = i$
- Then the algorithm swaps $A[i]$ with the larger between $\text{Left}(i)$ and $\text{Right}(i)$ (if any) and one subtree was already a heap and the other will be by inductive hypothesis.

MAX-HEAPIFY(A, i)

```
1:  $l = \text{Left}(i)$ 
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➤ Procedures (what do we need)

1. **Build-Max-Heap**: produces a Max-Heap from an unordered array
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➤ Building a Heap

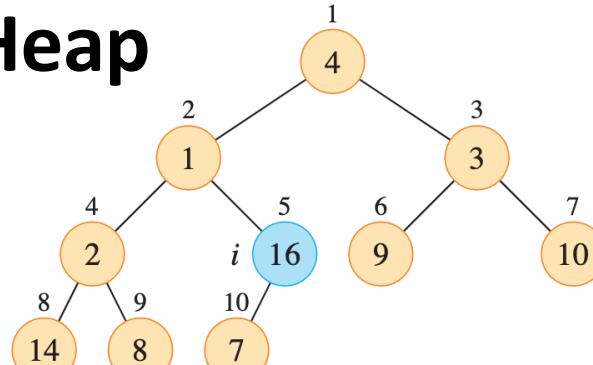
- Idea: use Max-Heapify repeatedly to create a heap.
- Which order of nodes: top-down or bottom-up?
- Answer: **bottom-up** – Max-Heapify assumes Left(i) and Right(i) are heaps. Top-down wouldn't work, bottom-up does.
- Note: nodes in $A\left[\left(\left\lfloor \frac{n}{2} \right\rfloor + 1\right), \dots, n\right]$ are all leaves. Leaves are max-heaps, so no work required.

BUILD-MAX-HEAP(A, n)

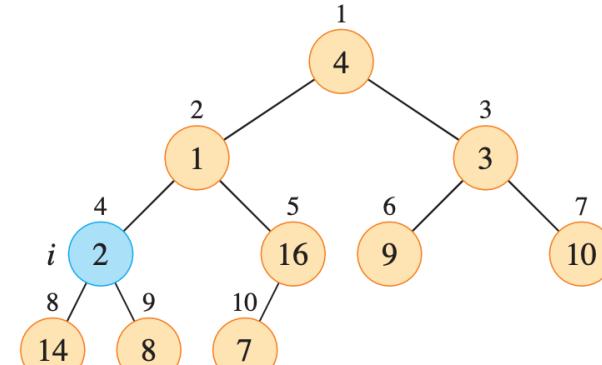
```
1   $A.\text{heap-size} = n$ 
2  for  $i = \lfloor n/2 \rfloor$  downto 1
3      MAX-HEAPIFY( $A, i$ )
```

<i>A</i>	4	1	3	2	16	9	10	14	8	7
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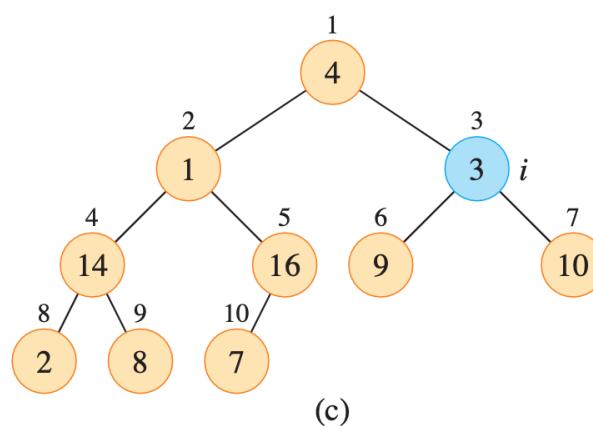
➤ Build-Max-Heap



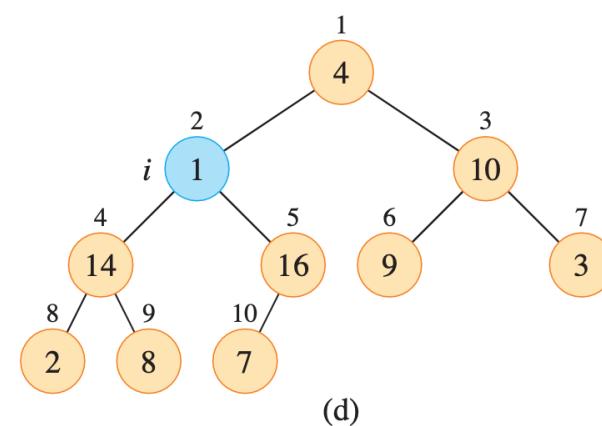
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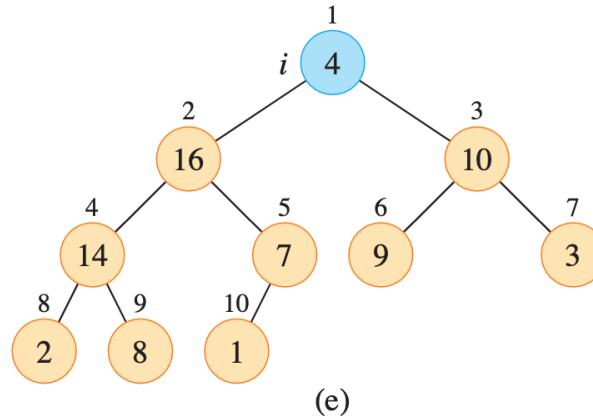
(b)



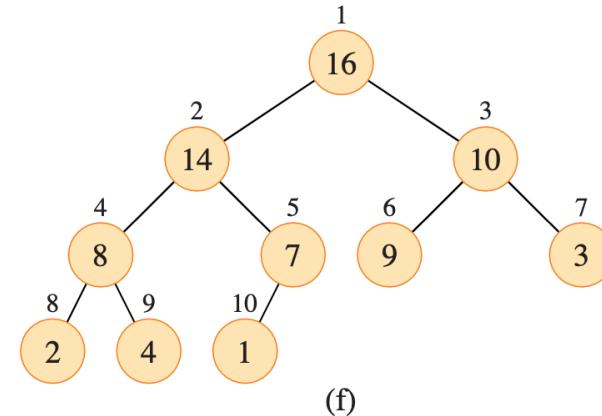
(c)



(d)



(e)



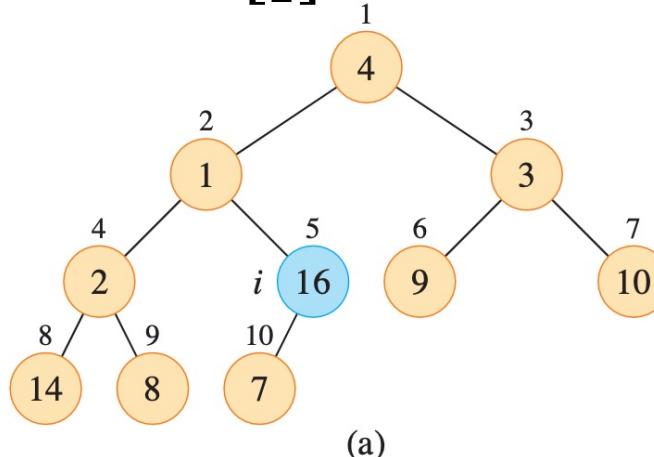
(f)

➤ Correctness of Build-Max-Heap

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BUILD-MAX-HEAP( $A, n$ )
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1    $A.\text{heap-size} = n$ 
2   for  $i = \lfloor n/2 \rfloor$  downto 1
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```

- **Loop invariant:** At the start of each iteration i of the for loop, each node $i + 1, i + 2, \dots, n$ is the root of a max-heap.
- **Initialisation:** true for leaves $\left\lfloor \frac{n}{2} \right\rfloor + 1, \dots, n$.

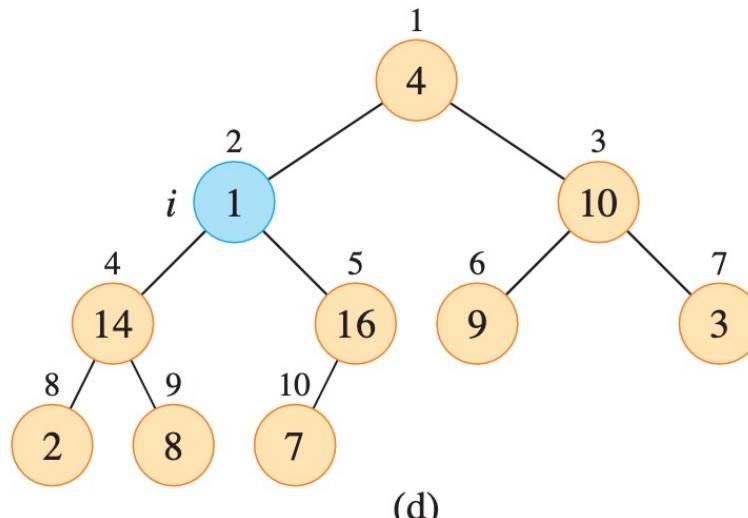


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```

- **Loop invariant:** At the start of each iteration i of the for loop, each node $i + 1, i + 2, \dots, n$ is the root of a max-heap.
- **Maintenance:** by loop invariant, all children of i are roots of max-heaps (as their numbers are larger than i).
Then Max-Heapify(A, i) turns the subtree at i into a max-heap.

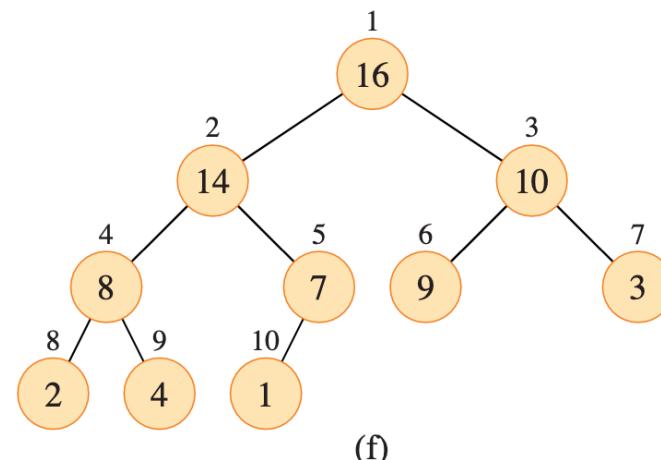
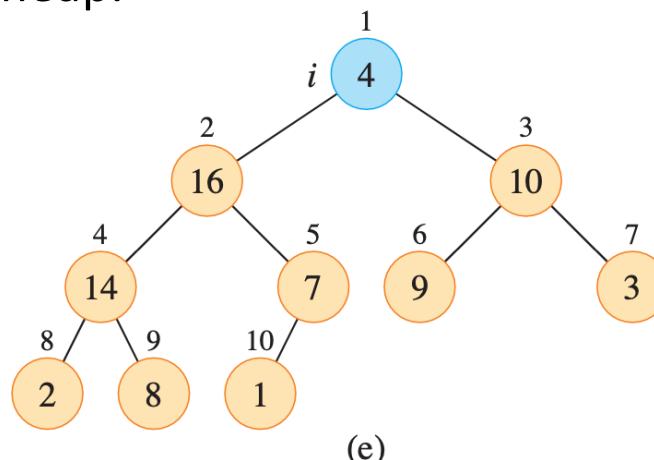


➤ Correctness of Build-Max-Heap

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- **Loop invariant:** At the start of each iteration i of the for loop, each node $i + 1, i + 2, \dots, n$ is the root of a max-heap.
- **Termination:** the loop terminates at $i = 0$, hence node 1 is the root of a max-heap.



➤ Runtime of Build-Max-Heap

- The **height of a heap** = height of the root is at most $\log n$.
- So all nodes have height at most $\log n$.
- Every call to Max-Heapify takes time $O(\log n)$.
- Build-Max-Heap calls Max-Heapify $O(n)$ times.
- Total time is at most $O(n) \cdot O(\log n) = O(n \log n)$.
 - The time can be improved to $O(n)$ since most nodes have small height.
 - $O(n \log n)$ is sufficient for us, though.

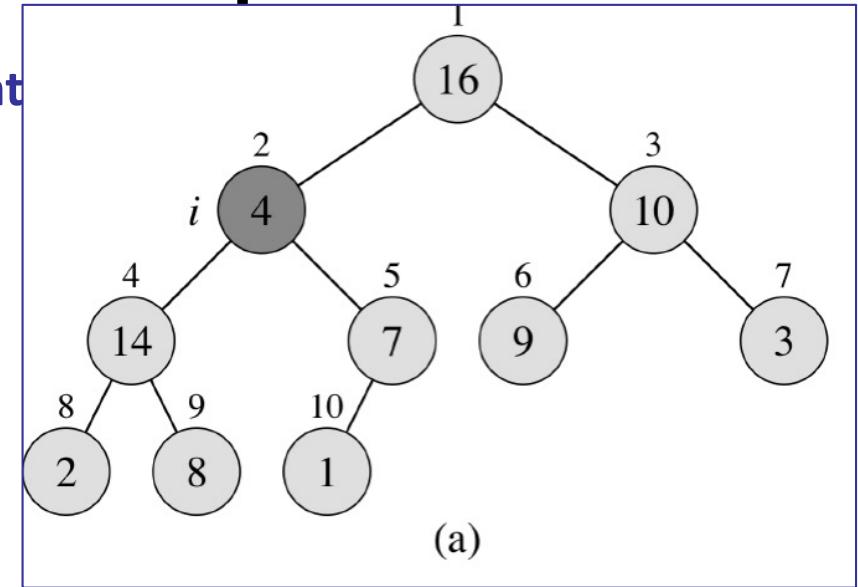
➤ Refined Analysis of Build-Max-Heap

- **Observation: most nodes have small height**
- One can show: there are at most $\left\lceil \frac{n}{2^{h+1}} \right\rceil$ nodes of height h .
- $O(\log n)$ time bound is correct, but crude for most nodes.
- A better bound:

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

as the infinite series of $\frac{h}{2^h}$ is 2.

- 1st equality, we used that: $\lceil x \rceil \leq 2x$ for $x \geq 1/2$
 \Rightarrow for $h \leq \log n$, $\frac{n}{2^{h+1}} \geq 1/2$ because $n \geq 2^h$ (see slide 13)
- 2nd equality, we used that $\sum_{k=0}^{\infty} kx^k = \frac{x}{(1-x)^2}$ for $|x| < 1$

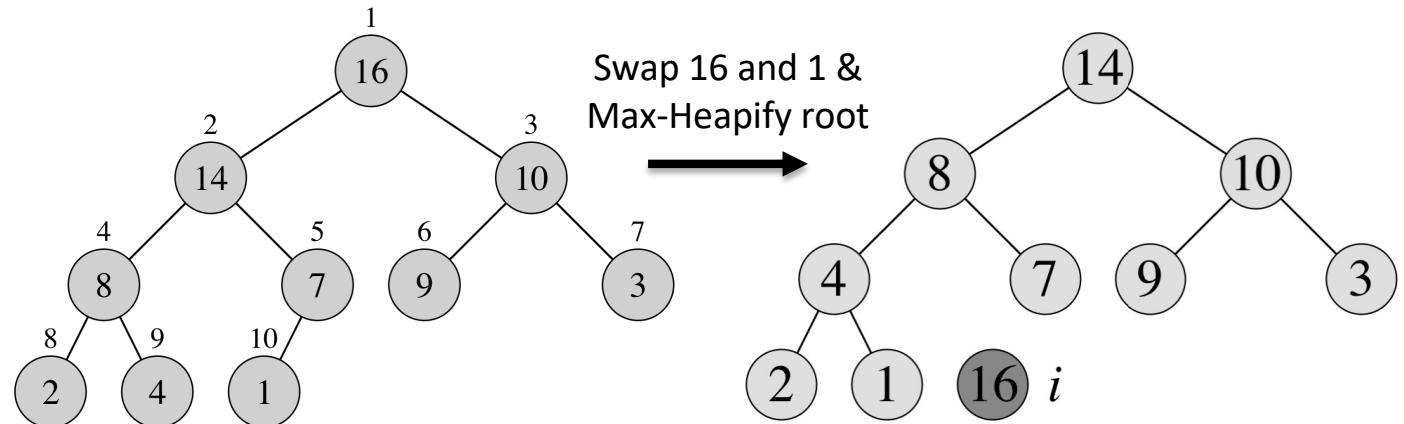


➤ Procedures (what do we need)

1. **Build-Max-Heap:** produces a Max-Heap from an unordered array ✓
2. **Max-Heapify:** maintains the max-heap property once the maximum has been removed ✓
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➤ HeapSort

- Ideas:

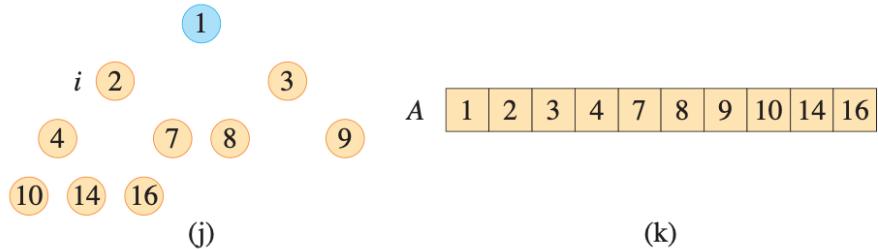
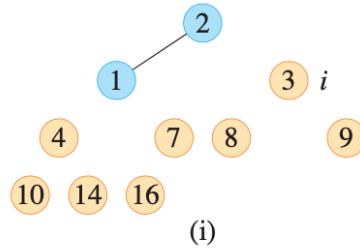
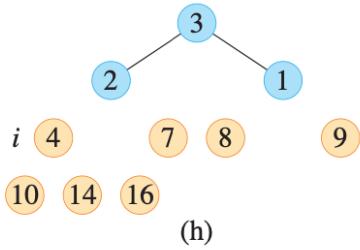
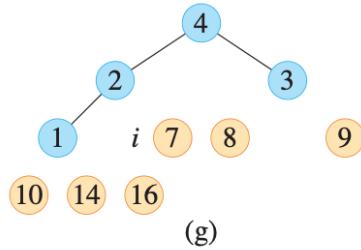
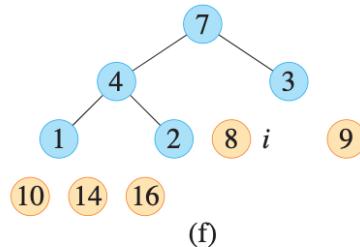
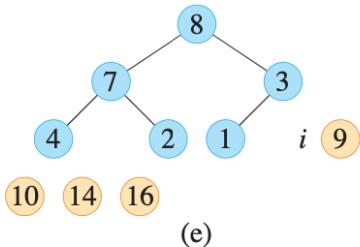
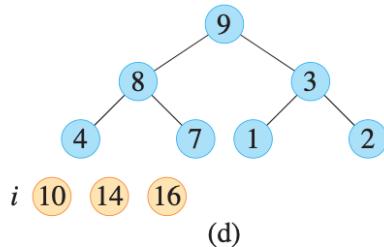
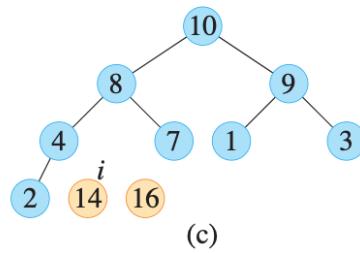
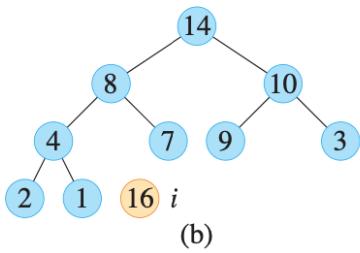
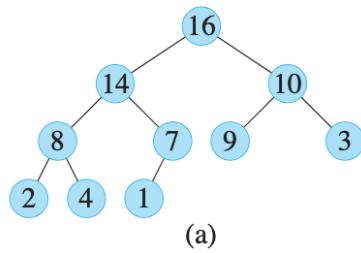


1. Build a max-heap, such that the root contains largest element.
2. Swap the root with the last element of the heap/array.
3. Discard the last element from the heap by reducing heap.size.
(We simply imagine a smaller heap.)
4. Call Max-Heapify($A, 1$) to restore heap property at the root.

HEAPSORT(A)

- 1: BUILD-MAX-HEAP(A)
- 2: **for** $i = A.length$ downto 2 **do**
- 3: exchange $A[1]$ with $A[i]$
- 4: $A.heap-size = A.heap-size - 1$
- 5: MAX-HEAPIFY($A, 1$)

➤ HeapSort: Example

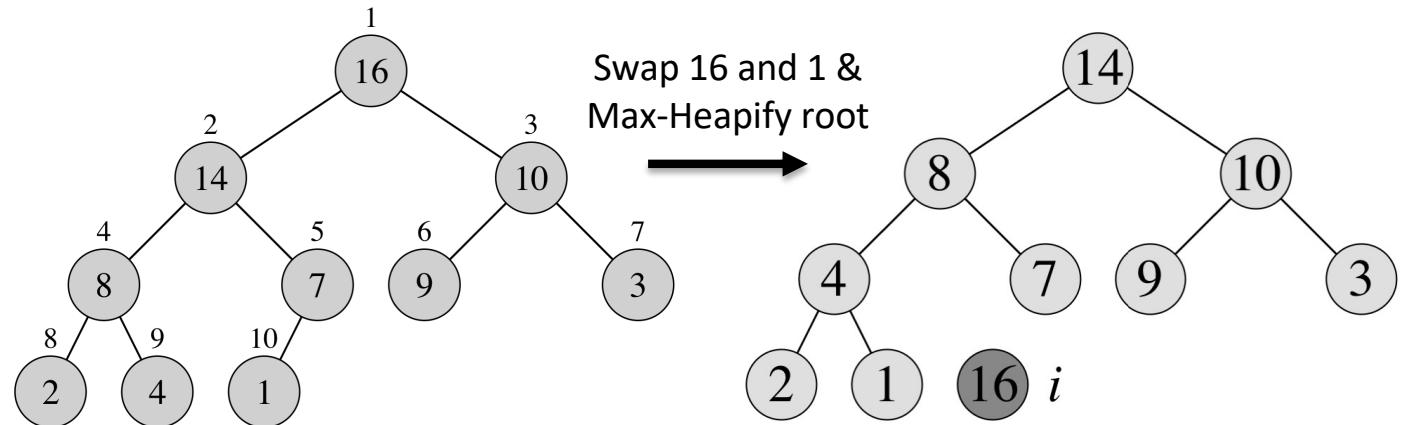


A	1 2 3 4 7 8 9 10 14 16
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(k)

➤ HeapSort

- Ideas:



1. Build a max-heap, such that the root contains largest element.
2. Swap the root with the last element of the heap/array.
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(We simply imagine a smaller heap.)
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Runtime:

$$\begin{aligned} & O(n \log n) \\ & + (n - 1) \cdot O(\log n) \\ & = O(n \log n) \end{aligned}$$

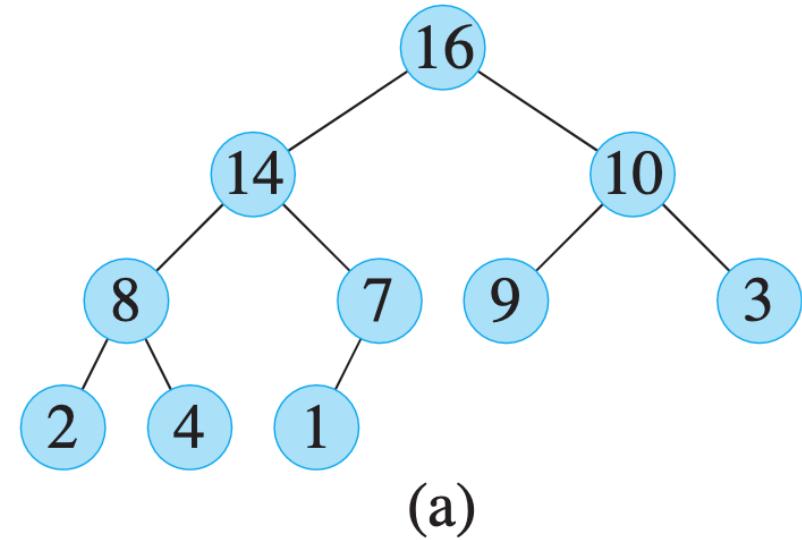
➤ Correctness of HeapSort

Loop Invariant: “At the start of each iteration of the for loop of lines 2-5, the subarray $A[1..i]$ is a max-heap containing the i smallest elements of $A[1..n]$, and the subarray $A[i+1..n]$ contains the $n-i$ largest elements of $A[1..n]$, sorted.”

- **Initialization:** The subarray $A[i+1..n]$ is empty, thus the invariant holds.

HEAPSORT(A)

```
1: BUILD-MAX-HEAP( $A$ )
2: for  $i = A.length$  downto 2 do
3:   exchange  $A[1]$  with  $A[i]$ 
4:    $A.heap-size = A.heap-size - 1$ 
5:   MAX-HEAPIFY( $A, 1$ )
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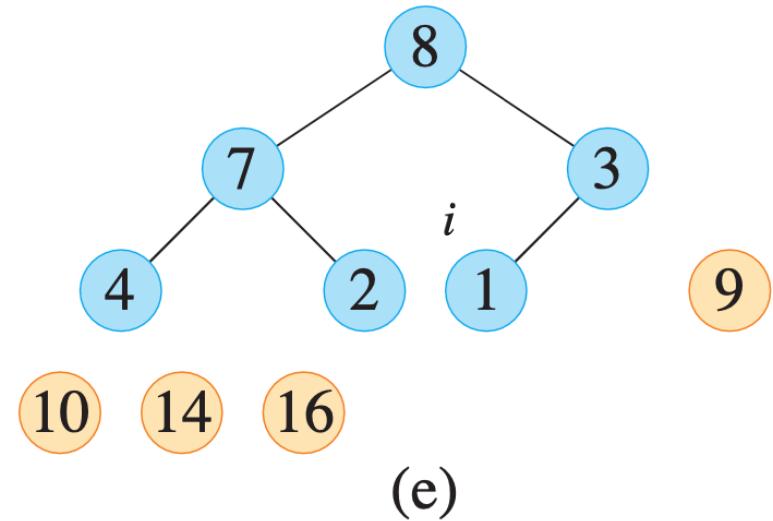
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Maintenance: $A[1]$ is the largest element in $A[1..i]$ and it is smaller than the elements in $A[i+1..n]$. When we put it in the i th position, then $A[i..n]$ contains the largest elements, sorted. Decreasing the heap size and calling Max-Heapify turns $A[1..i-1]$ into a max-heap. Decrementing i sets up the invariant for the next iteration.

HEAPSORT(A)

- 1: BUILD-MAX-HEAP(A)
- 2: **for** $i = A.length$ downto 2 **do**
- 3: exchange $A[1]$ with $A[i]$
- 4: $A.heap-size = A.heap-size - 1$
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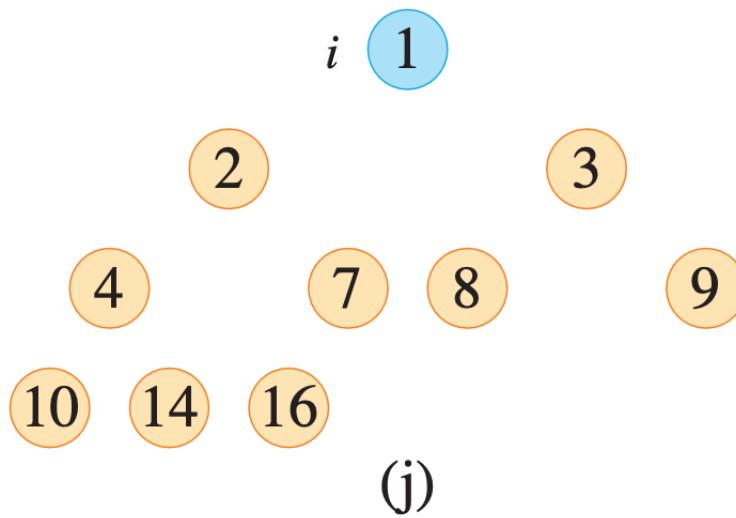
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- **Termination:** After the loop $i=1$. This means that $A[2..n]$ is sorted and $A[1]$ is the smallest element in the array, which makes the array sorted.

HEAPSORT(A)

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- 3: exchange $A[1]$ with $A[i]$
- 4: $A.heap-size = A.heap-size - 1$
- 5: MAX-HEAPIFY($A, 1$)



➤ Summary

- HeapSort sorts in place in time $O(n \log n)$.
 - Building a Heap in time $O(n)$.
 - Extracting the largest element and restoring the heap-property in total time $O(n \log n)$.
- The use of appropriate **data structures** can speed up computation (in contrast to SelectionSort).
 - The heap “memorises” information about comparisons of elements.
 - The heap is imaginary, no objects/pointers required!
- Heaps also play a role in **Priority Queues**.