

Algorithms

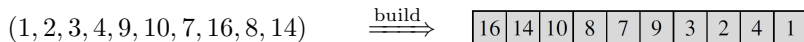
Heaps and priority queues

Emanuele Rodolà
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Heap

Data structures allow us to **organize data** for efficient use.



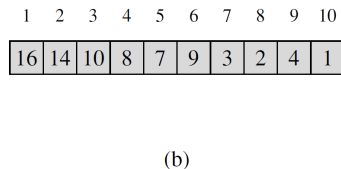
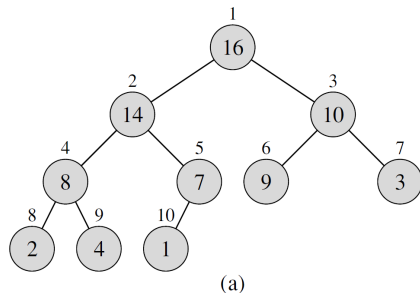
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1	2	3	4	5	6	7	8	9	10
16	14	10	8	7	9	3	2	4	1

Heap

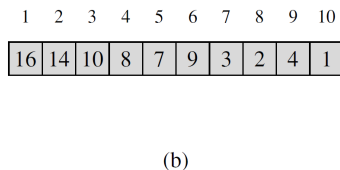
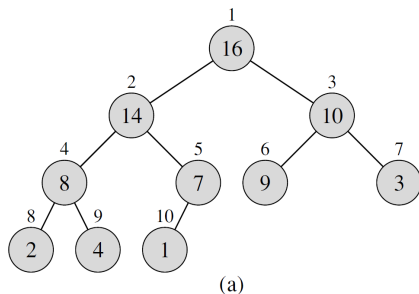
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- $\text{left}(i) = 2i$
- $\text{right}(i) = 2i + 1$

Heap

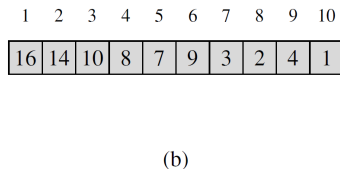
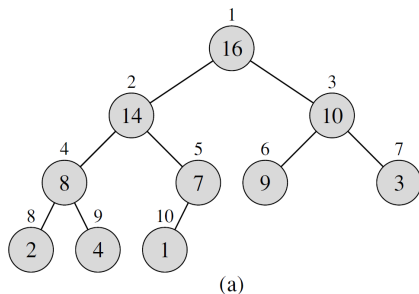
Data structures allow us to **organize data** for efficient use.



- $\text{left}(i) = 2i$
- $\text{right}(i) = 2i + 1$
- $\text{parent}(i) = \lfloor i/2 \rfloor$

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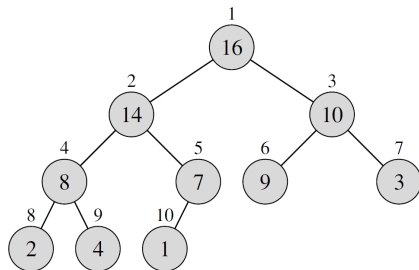
max-heap property:

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

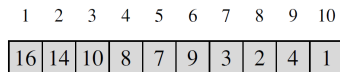
which means that **root** = **largest** element.

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(a)



(b)

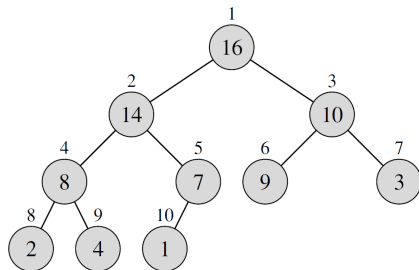
min-heap property:

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

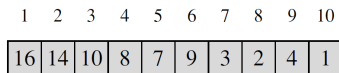
which means that **root** = **smallest** element.

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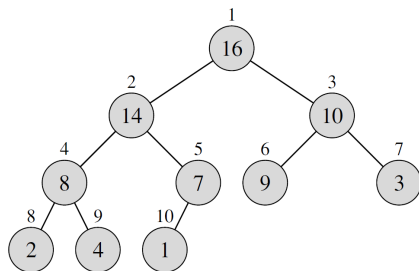


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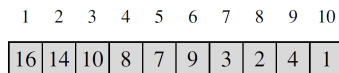
height of a node: # edges of the longest path to a leaf

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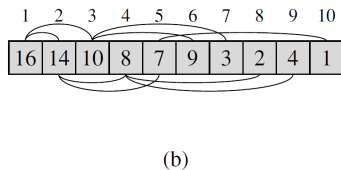
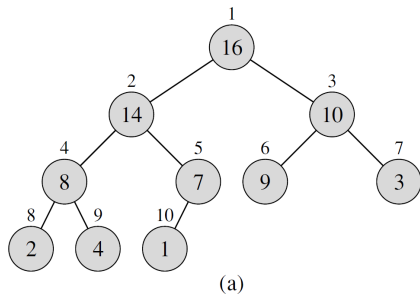
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height of a node: # edges of the longest path to a leaf

height of the heap = height of the root = $\Theta(\lg n)$

Heap

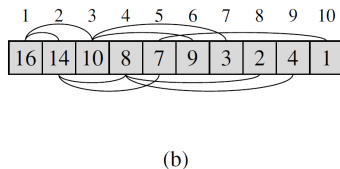
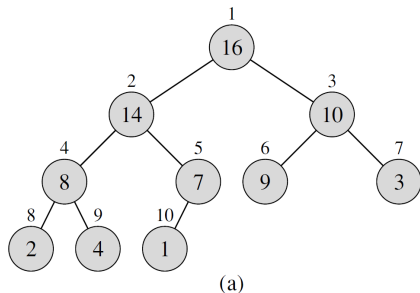
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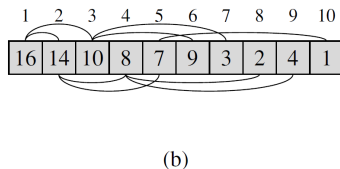
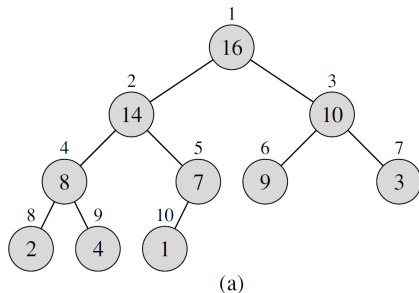


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Each level (except the last one) must be **filled completely** left-to-right.

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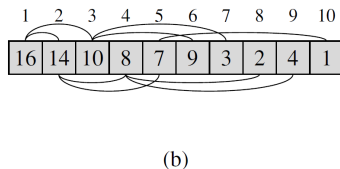
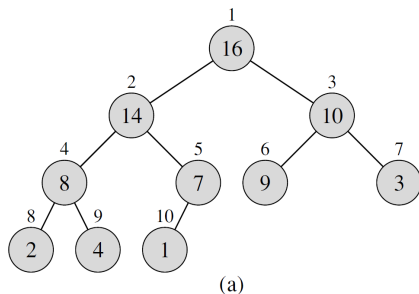
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The number of **leaves** is $n - \lfloor \frac{n}{2} \rfloor$.

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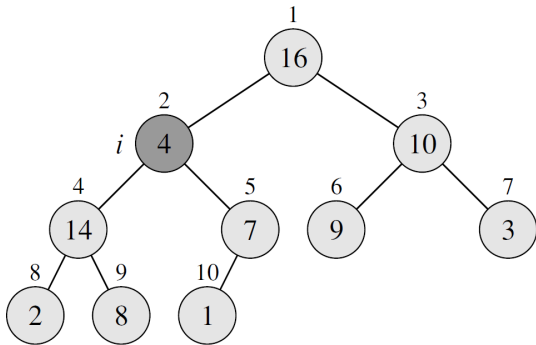


We will see the following operations:

- **Build** a (max-)heap from some input array.
- **Maintain** the (max-)heap property.
- Construct a **priority queue** on top of a heap.

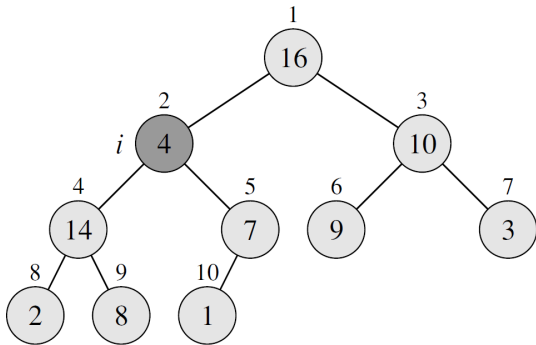
“Heapify”: Maintaining the heap property

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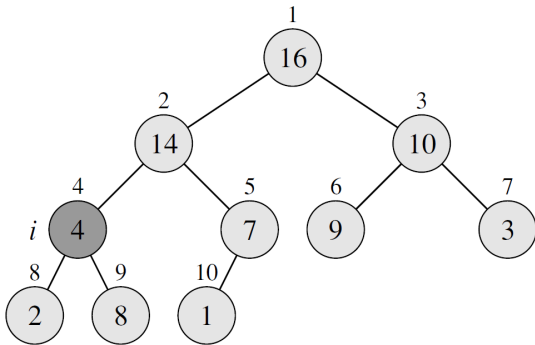
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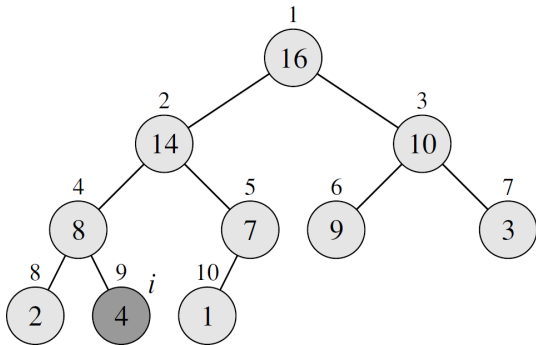


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Maintaining the heap property

MAX-HEAPIFY(A, i)

```
1   $l \leftarrow \text{LEFT}(i)$ 
2   $r \leftarrow \text{RIGHT}(i)$ 
3  if  $l \leq \text{heap-size}[A]$  and  $A[l] > A[i]$ 
4      then  $\text{largest} \leftarrow l$ 
5      else  $\text{largest} \leftarrow i$ 
6  if  $r \leq \text{heap-size}[A]$  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( $A, \text{largest}$ )
```

If $A[i]$ is the largest, then the sub-tree is already a max heap.

Maintaining the heap property

MAX-HEAPIFY(A, i)

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

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$$T(n) \leq T(2n/3) + \Theta(1)$$

Maintaining the heap property

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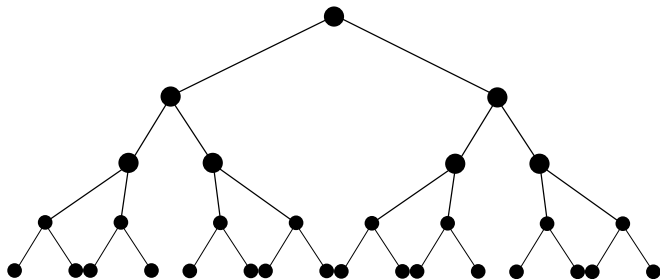
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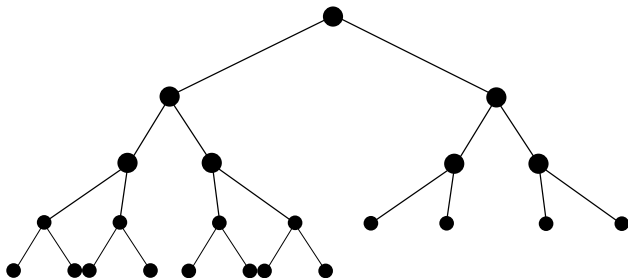
$$T(n) = O(\lg n)$$

by case (2) of the [master theorem](#).

Maximum size of sub-tree in a heap

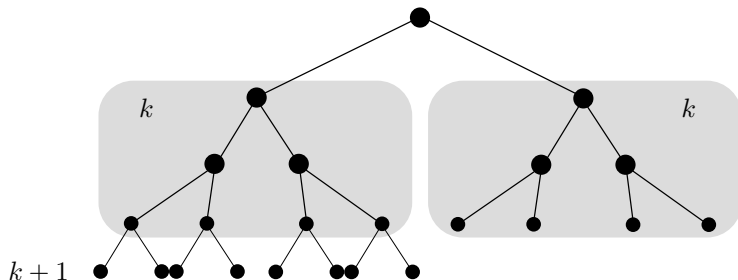


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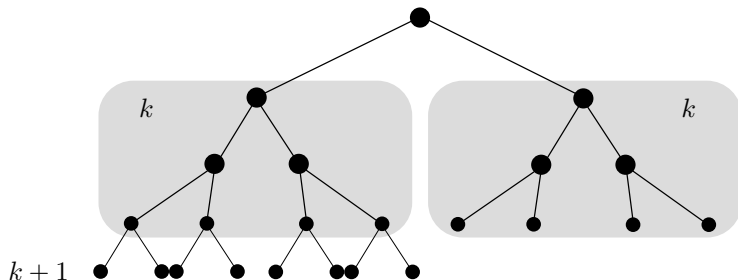
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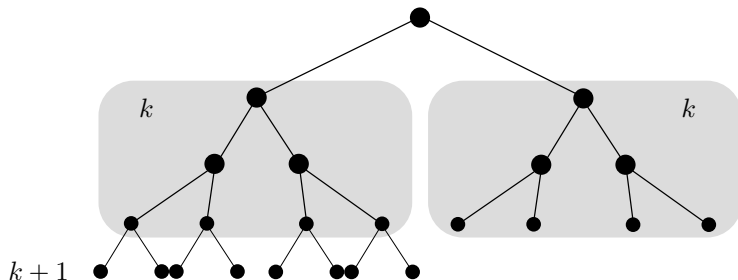
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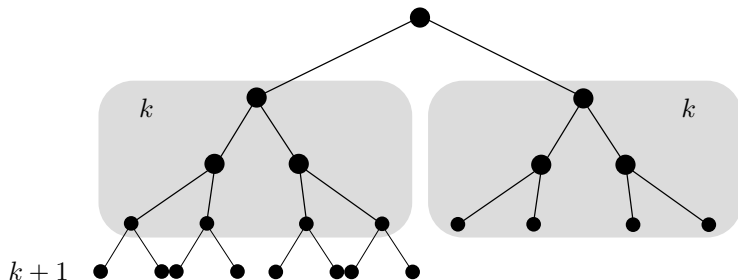
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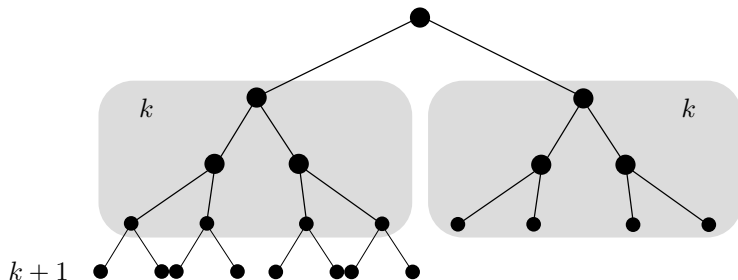
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- For arbitrary k , we get $\lim_{k \rightarrow \infty} \frac{2k+1}{3k+2}n = \frac{2}{3}n$.

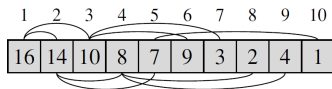
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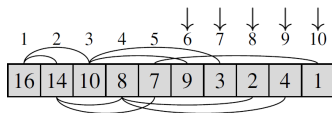
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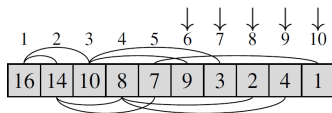
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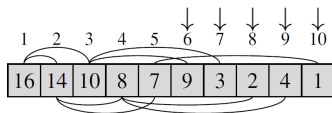


General idea: run **max-heapify** on the internal nodes.

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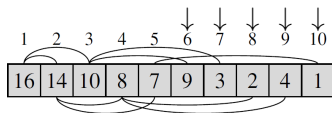
BUILD-MAX-HEAP(A)

- 1 $heap\text{-}size[A] \leftarrow length[A]$
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- 3 **do** **MAX-HEAPIFY**(A, i)

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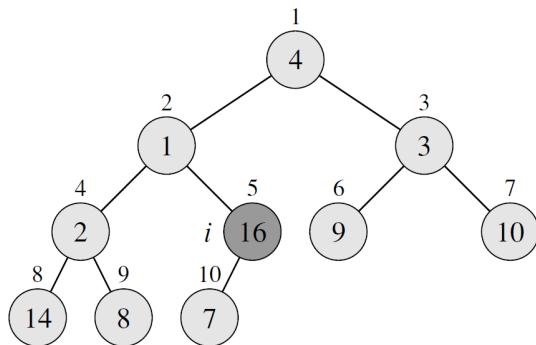
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$O(n)$ calls to a $O(\lg n)$ algorithm: $T(n) = O(n \lg n)$.

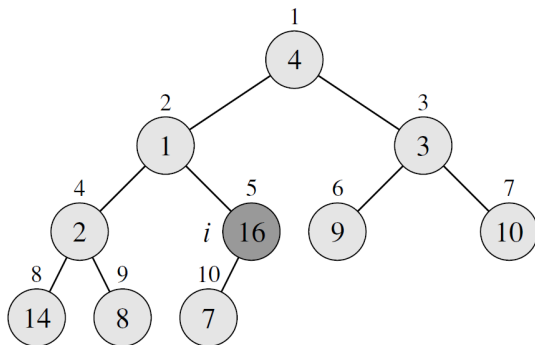
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-----	---	---	---	---	----	---	----	----	---	---

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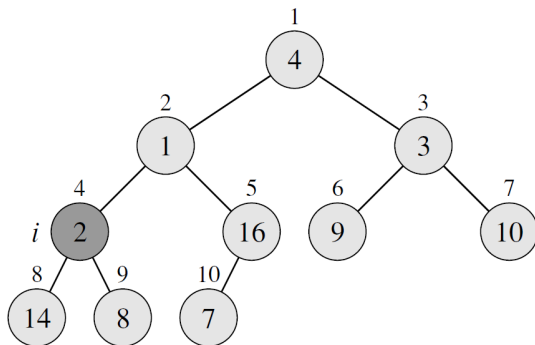


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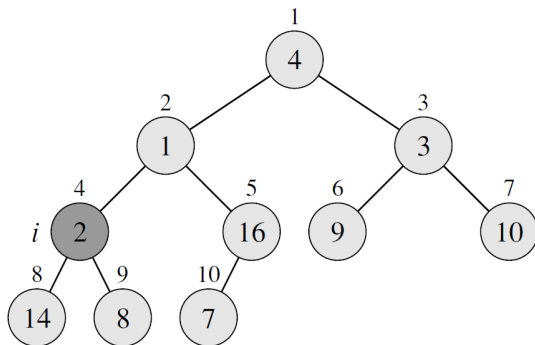


The sub-tree is already a heap.

Building a heap

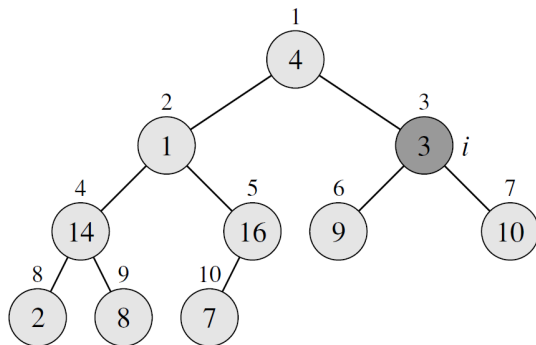


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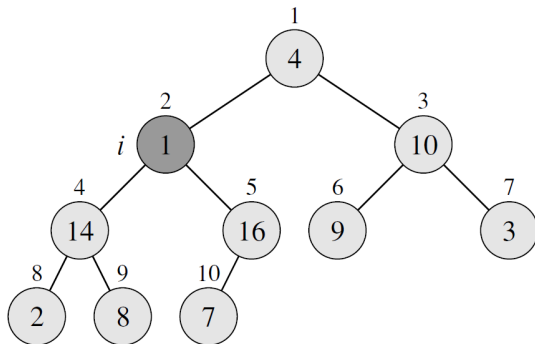


The sub-tree does **not** satisfy the heap property – run **heapify**.

Building a heap



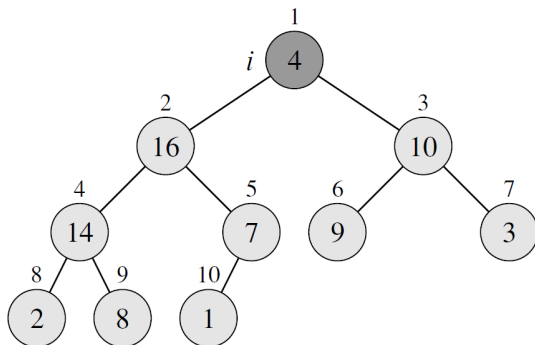
Building a heap



By proceeding **bottom-up**, we can always run **heapify**, which assumes that the sub-trees rooted at $\text{left}(i)$ and $\text{right}(i)$ are valid.

This way, after heapify we can guarantee that i is the root of a heap.

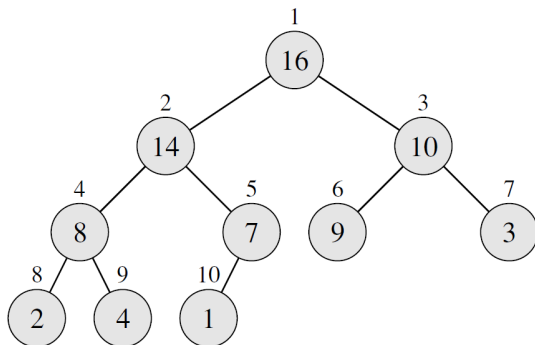
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Priority queues

Example:

Let us be given a set of tasks, each with a **priority**.

When a task is done, pass to the next with highest priority.

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- **Increase** the priority of a given element.

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We will **implement** the priority queue **on top of a heap**.

Priority queues

The heap stores the **priority** values of each element.

In practice the element itself is also attached, but we ignore it here.

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Extract the highest-priority element in $O(\lg n)$ time:

HEAP-EXTRACT-MAX(A)

1 **if** $\text{heap-size}[A] < 1$

2 **then error** “heap underflow”

3 $\text{max} \leftarrow A[1]$

discard
the root || 4 $A[1] \leftarrow A[\text{heap-size}[A]]$

5 $\text{heap-size}[A] \leftarrow \text{heap-size}[A] - 1$

6 MAX-HEAPIFY($A, 1$)

7 **return** max

Priority queues

The heap stores the **priority** values of each element.

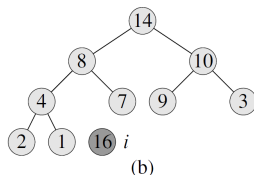
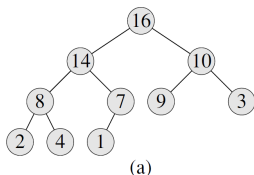
In practice the element itself is also attached, but we ignore it here.

Find the highest priority element in $\Theta(1)$ time:

HEAP-MAXIMUM(A)

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

Extract the highest-priority element in $O(\lg n)$ time:



discarding the root

Priority queues

Increase the priority of a given element in $O(\lg n)$ time:

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  $A[\text{PARENT}(i)] < A[i]$ 
5      do exchange  $A[i] \leftrightarrow A[\text{PARENT}(i)]$ 
6       $i \leftarrow \text{PARENT}(i)$ 
```

Priority queues

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```

Line 3 can violate the max-heap property, depending on the value of key .

Priority queues

Increase the priority of a given element in $O(\lg n)$ time:

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

```
1  if  $key < A[i]$ 
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3   $A[i] \leftarrow key$ 
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5      do exchange  $A[i] \leftrightarrow A[\text{PARENT}(i)]$ 
6       $i \leftarrow \text{PARENT}(i)$ 
```

Line 3 can violate the max-heap property, depending on the value of key .

If so, the correct index for key must be **up** in the tree.

Lines 4-6 find the correct index by moving the value in an upward path.

Priority queues

Insert a new element + priority in $O(\lg n)$ time:

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($A, heap-size[A], key$)

Priority queues

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1 $heap-size[A] \leftarrow heap-size[A] + 1$

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The value of $-\infty$ ensures that line 1 of **increase** is false.

Priority queues

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The value of $-\infty$ ensures that line 1 of **increase** is false.

To summarize:

operation	cost
find	$\Theta(1)$
extract	$O(\lg n)$
increase	$O(\lg n)$
insert	$O(\lg n)$

Suggested reading

Chapters 6.1, 6.2, 6.3 and 6.5 of:

“Introduction to Algorithms – 2nd Ed.”, Cormen et al.