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## Priority Queue

Main operations:

- add(*x*)
- removeMin()
- findMin()

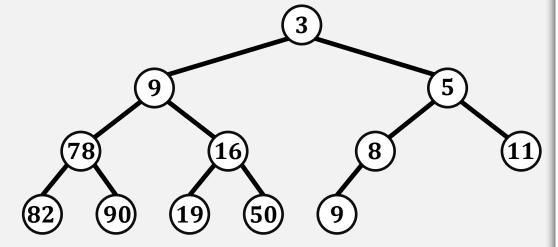
Additional operations:

- remove(u)
- decreaseKey(u)

u is the pointer to a node/element

# Binary Heap

every level (except possibly the bottom) is completely full



Binary Heap is a complete binary tree with the heap property.

The minimum value is stored at the root.

Heap/(priority queue) allows duplicate values.

for each non-root node u:  $u \cdot x \ge u$ . parent. x Min-Heap

#### Supports:

- **insert** an element *x* to the heap
- extract minimum element:
  - return the smallest element
  - remove it from the heap



removeMin()

Heap is an implementation of a **priority queue** 

## Eytzinger Method

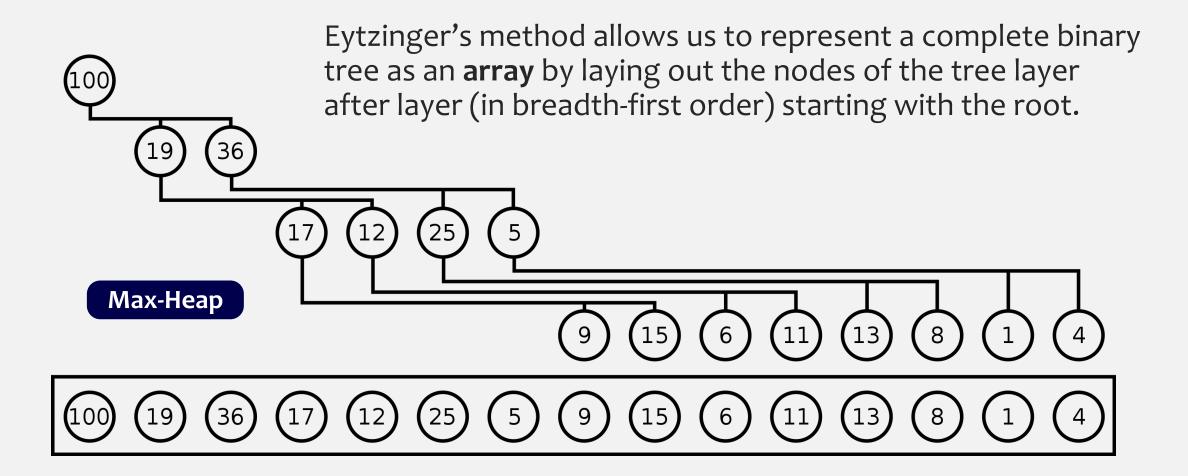
This method allows for the numbering of ancestors beginning with a descendant.

Allows to present family trees and pedigree charts in text format.



https://en.wikipedia.org/wiki/Ge nealogical numbering systems

## Heap using an Array



Min-Heap

a. heap size

Binary Heap Data Structure

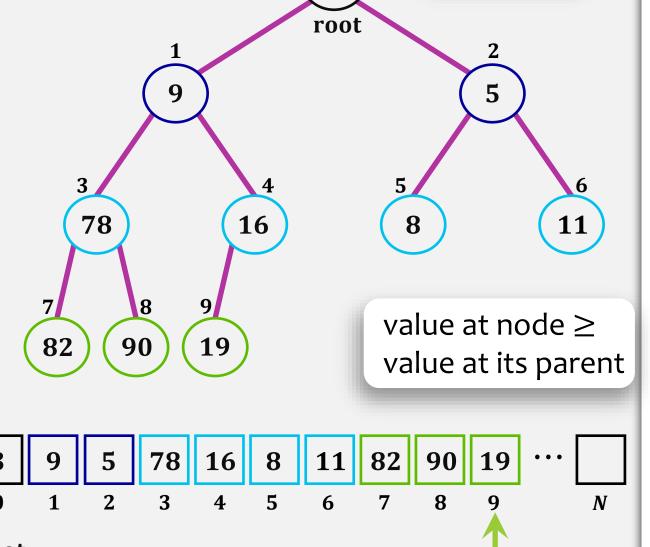
a:

It is an array that we can view as a nearly complete binary tree.

Do not confuse heap with BST.

Notice, a heap is not a sorted structure; it can be regarded as being **partially ordered**.

Heap can be used to sort elements.

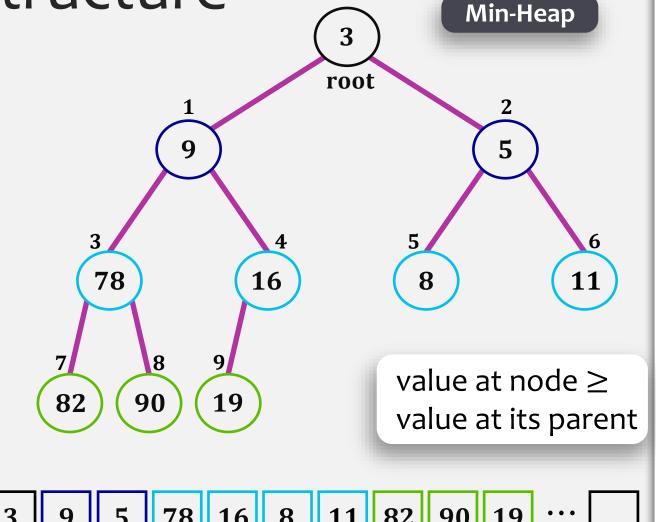


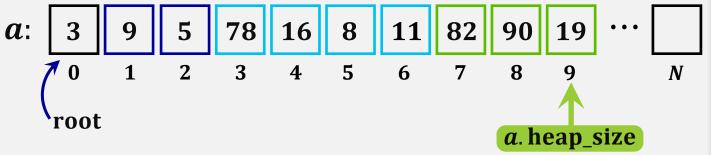
Binary Heap Data Structure

Array a[0..n-1] is a **heap** if for all  $0 < i \le n-1$ :  $a[parent(i)] \le a[i]$ 

#### Node with index *i*:

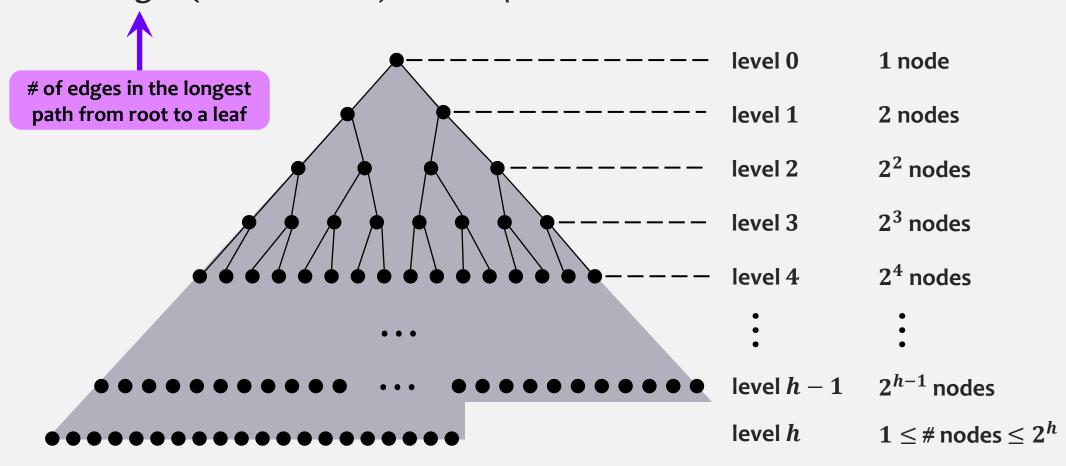
- Parent of i has index  $\left|\frac{i-1}{2}\right|$
- Left child of i has index 2i + 1
- Right child of i has index 2i + 2





### Binary Heap Data Structure

What is the **height** (let us call it h) of a heap with n nodes?



### Binary Heap Data Structure

$$2^{0} + 2^{1} + 2^{2} + \dots + 2^{k} = 2^{k+1} - 1$$

What is the **height** (let us call it h) of a heap with n nodes?

$$1 + 2 + 2^{2} + \dots + 2^{h-1} + 1 \le n \le 1 + 2 + 2^{2} + \dots + 2^{h-1} + 2^{h}$$

$$(2^{h} - 1) + 1 \le n \le 2^{h+1} - 1$$

$$2^{h} \le n < 2^{h+1}$$

$$h \le \log n < h + 1$$

$$\log n - 1 < h \le \log n$$

$$h = \lfloor \log n \rfloor$$

### Heaps – Basic Procedures

**0**(1)

findMin()

– return the smallest element, which is a[0]

 $O(\log n)$ 

decreaseKey(i, x)

- for  $\mathbf{x} \leq a[i]$  decrease a[i] to  $\mathbf{x}$  and restore the heap property.

 $O(\log n)$ 

add(x)

- insert element  $\mathbf{x}$  and restore the heap property.

 $O(\log n)$ 

minHeapify(i)

– restore the heap property for element a[i].

O(n)

buildMinHeap(a)

– build a heap from an unordered input array a.

 $O(n\log n)$ 

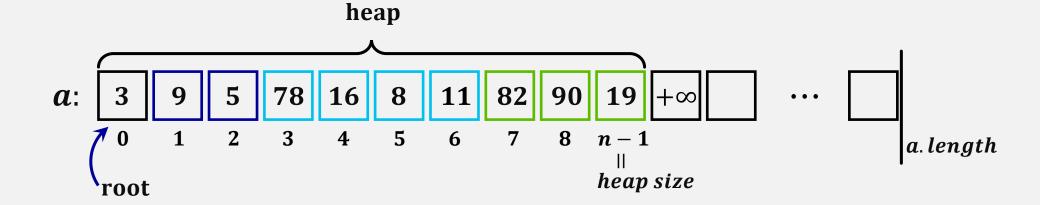
HeapSort(a)

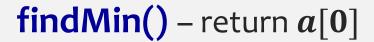
– sort an array a in place.

 $O(\log n)$ 

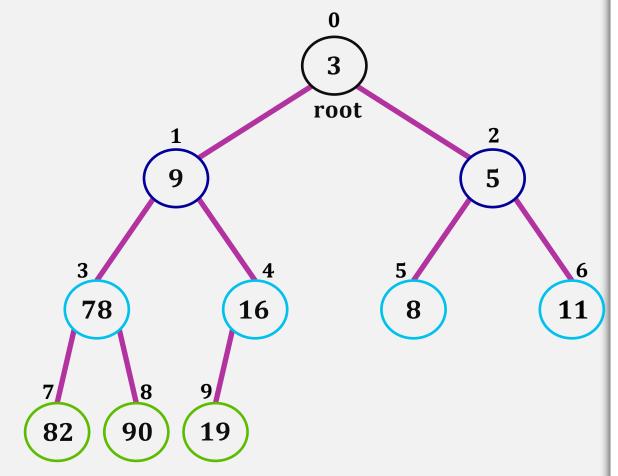
removeMin()

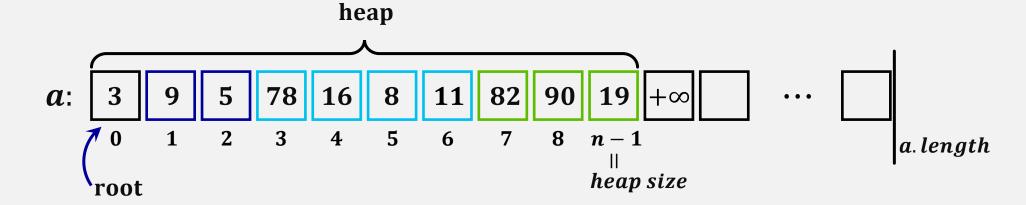
return and delete the smallest element,
 restore the heap property.



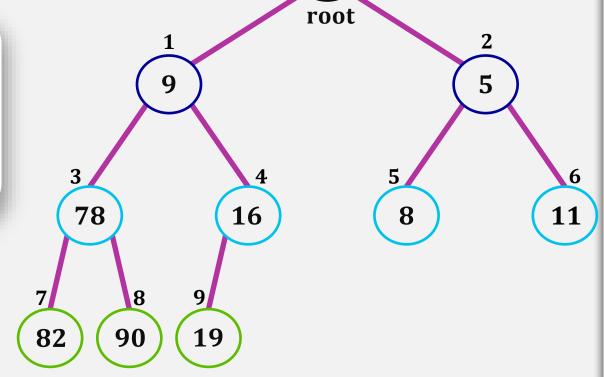


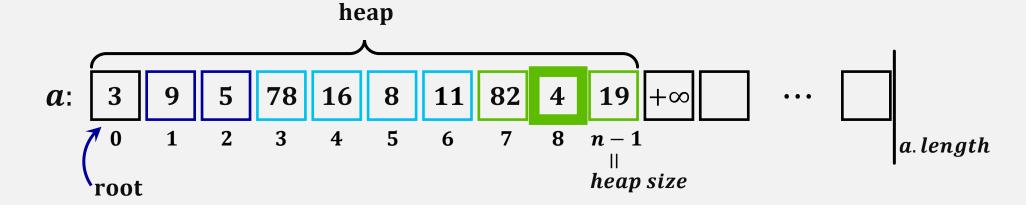
0(1)



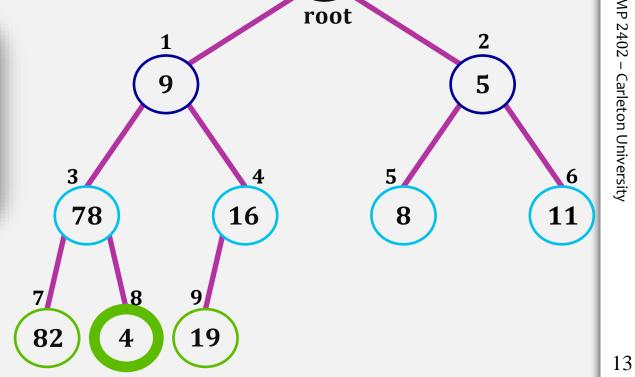


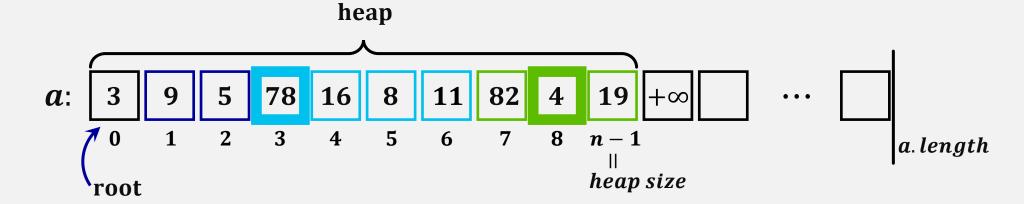
a[i] = x;while i > 0 and a[parent(i)] > a[i] do:
 swap a[i] and a[parent(i)]; i = parent(i);



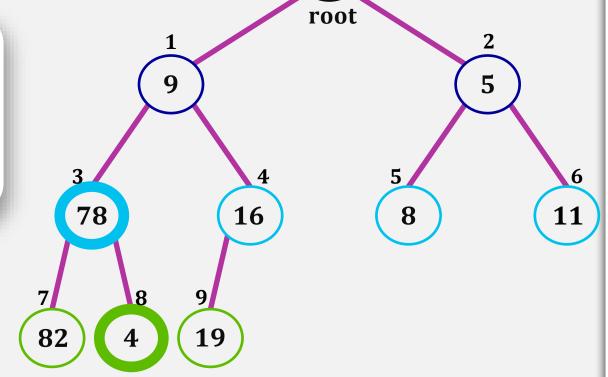


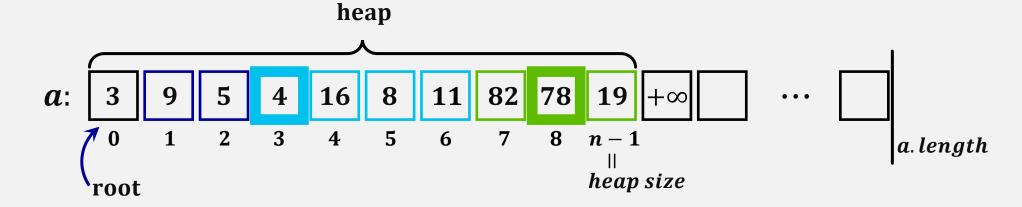
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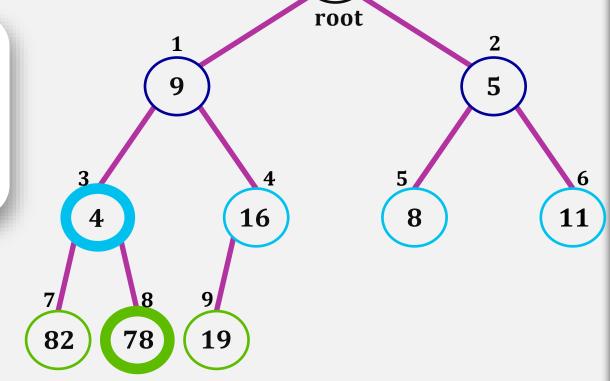


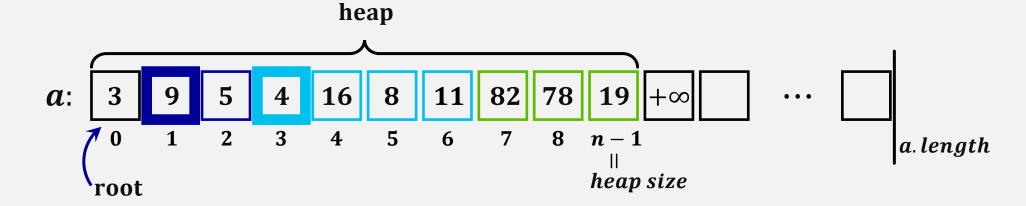
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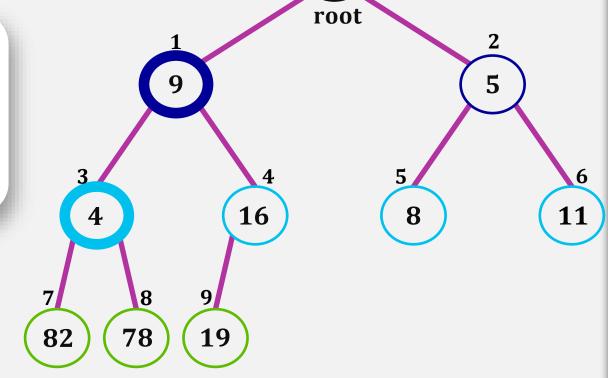


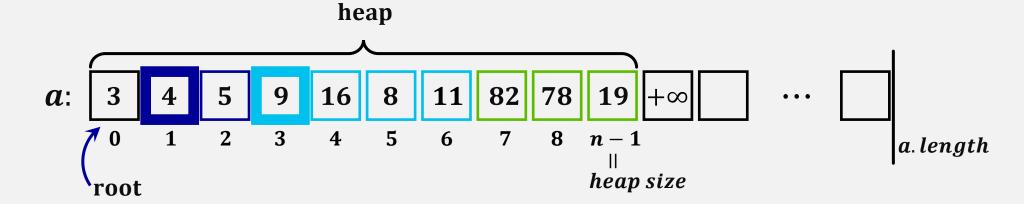
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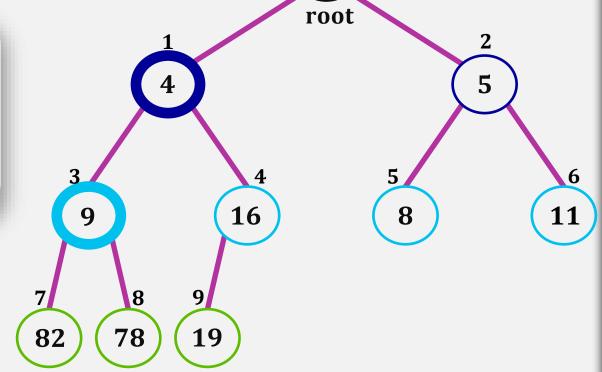


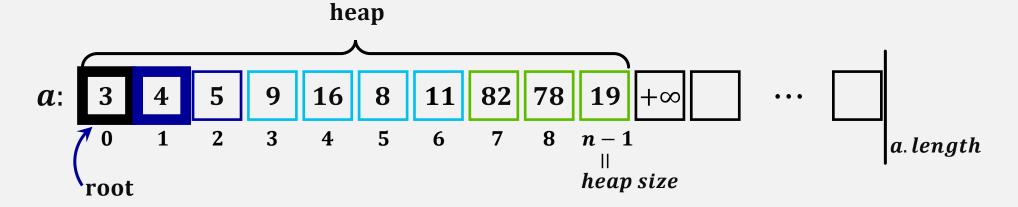
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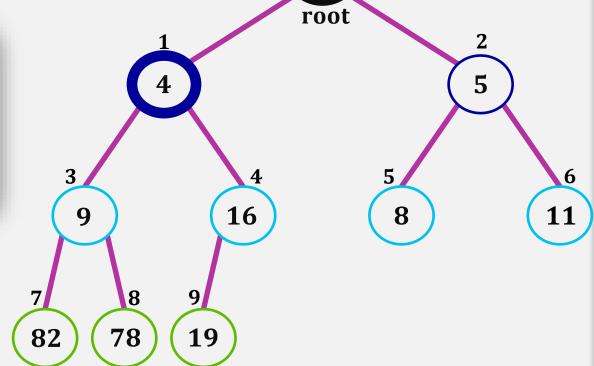


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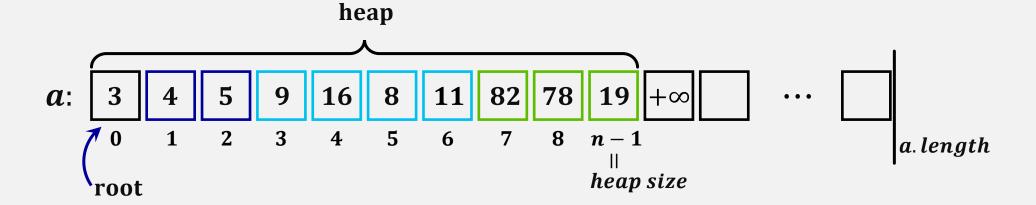




a[i] = x;while i > 0 and a[parent(i)] > a[i] do: swap a[i] and a[parent(i)]; i = parent(i); $O(\log n)$ 



root



add(x) – add element x to the bottom level of the heap at the leftmost open space and restore the heap property.

if (n+1>a.length) then resize(); n++;  $a[n] = \text{Integer.MAX_VALUE};$  decreaseKey(n-1,x);  $a[n] = \text{Integer.MAX_VALUE};$   $a[n] = \text{Integer.MAX_VAL$ 

**82** 

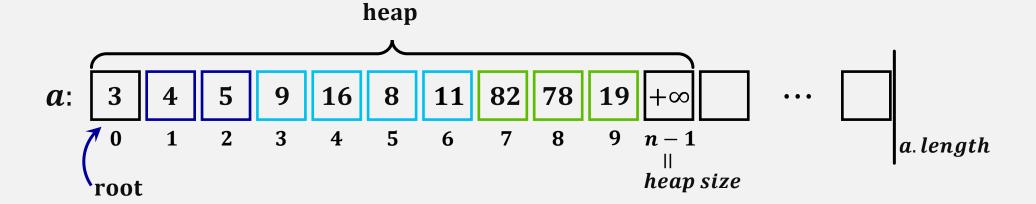
**78** 

add(12)

**11** 

root

10



**82** 

**78** 

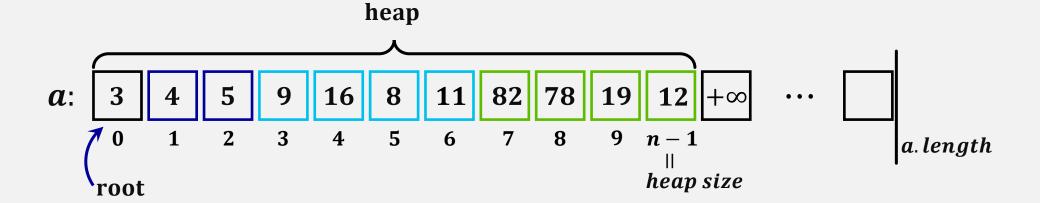
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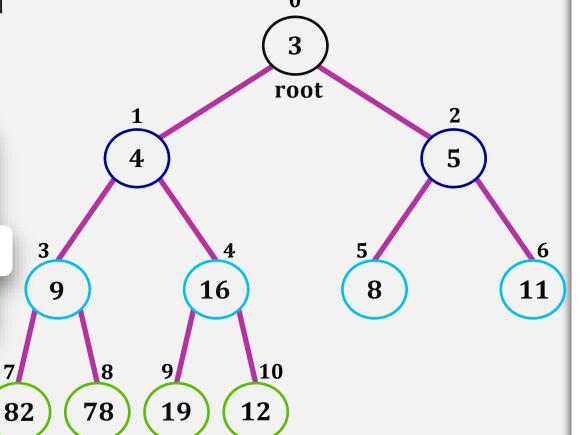
**11** 

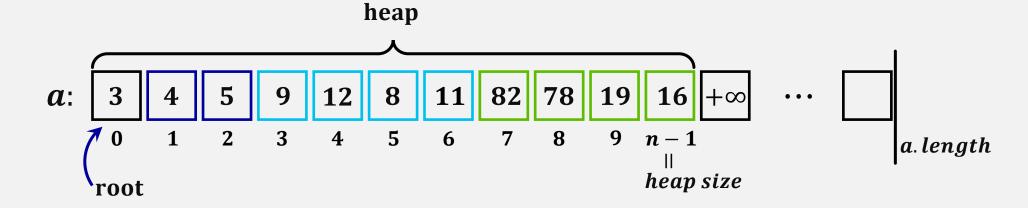


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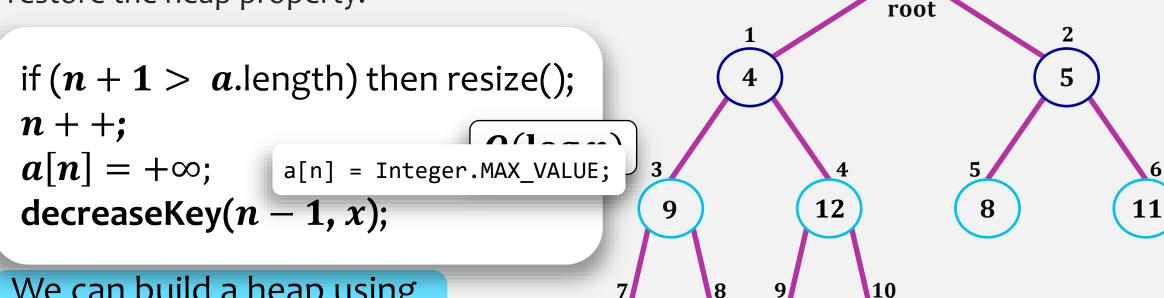
if (n+1>a.length) then resize(); n++;  $a[n]=+\infty;$   $a[n]=\text{Integer.MAX_VALUE};$  decreaseKey(n-1,x);

add(12)





add(x) – add element x to the bottom level of the heap at the leftmost open space and restore the heap property.



We can build a heap using add(x) operation.  $O(n \log n)$ 

**82 78 16** 

## How to build a heap?

**Input:** array a[0..N-1].

add(x) operation.

**Output:** heap a[0..N-1], containing the same elements

```
n=0; while n< N: add(a[n]); of a[n+1>a.length) then resize(); a[n]=+\infty; decreaseKey(n-1,x);
```

Time required = 
$$log1 + log2 + ... + log(\frac{N}{2}-1) + log(\frac{N}{2}) + ... + logN = \sum_{n=1}^{N} log n$$

Lower bound: 
$$\sum_{n=1}^{N} \log n \geqslant \sum_{n=N/2}^{N} \log n \geqslant \sum_{n=N/2}^{N} \log \frac{N}{2} = \log \frac{N}{2} \cdot \left(N - \frac{N}{2} + 1\right) \geqslant$$

$$\frac{1}{N-1} = \frac{1}{N} = \frac{$$

# minHeapify(i)

ods book calls this function **trickleDown()** 

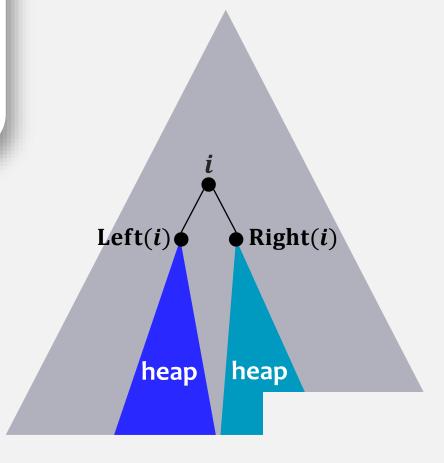
```
find the smallest of a[i], a[Left(i)] and a[Right(i)]; store its index in min; if min \neq i then swap a[i] and a[min]; minHeapify(min);
```

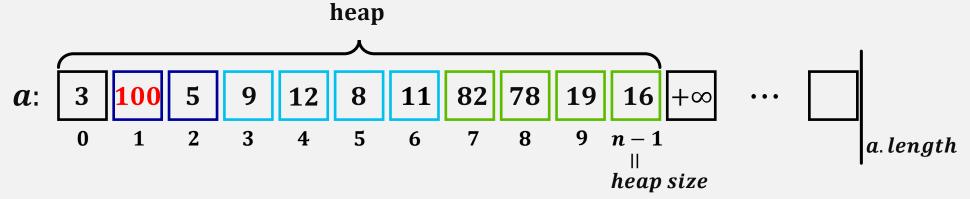
minHeapify(i) – restore the heap property for element a[i].

This operation assumes:

- $0 \le i \le n-1$ ;
- Subtree rooted at Left(i) is a heap;
- Subtree rooted at Right(i) is a heap;

When minHeapify(*i*) terminates, the subtree rooted at *i* is a heap.



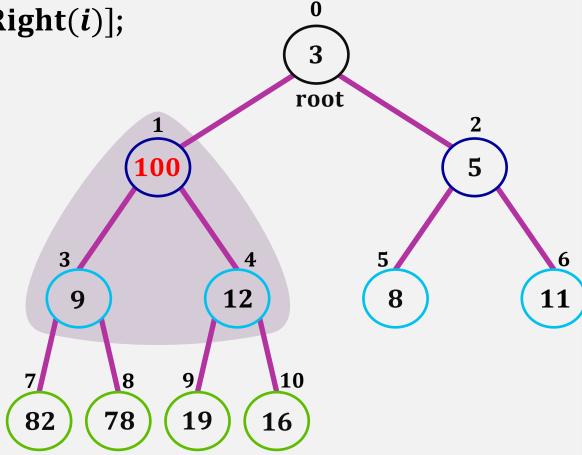


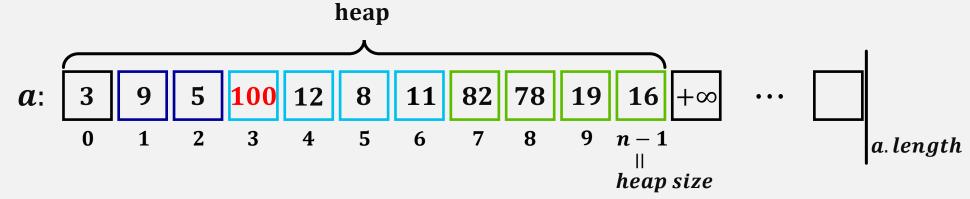
#### minHeapify(i):

find the smallest of a[i], a[Left(i)] and a[Right(i)]; store its index in min; if  $min \neq i$  then swap a[i] and a[min];

minHeapify(1)

minHeapify(min);



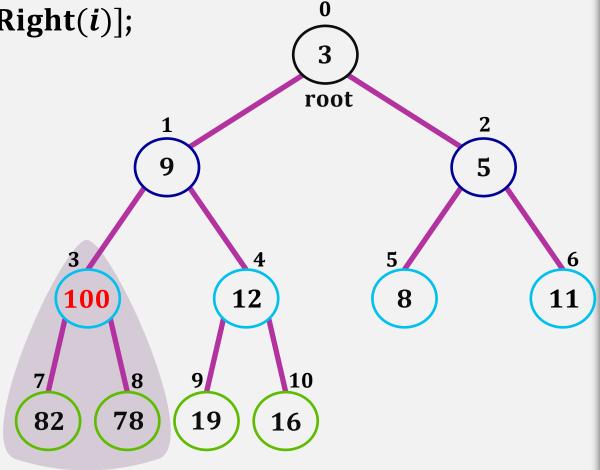


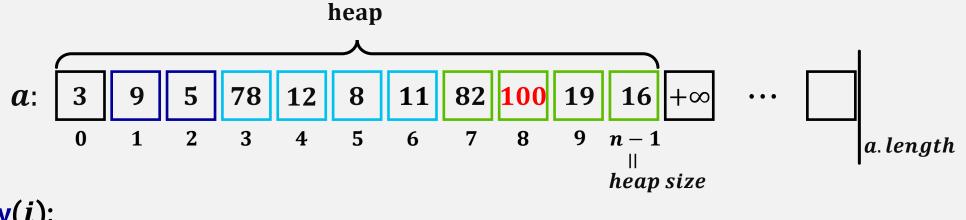
#### minHeapify(i):

find the smallest of a[i], a[Left(i)] and a[Right(i)]; store its index in min; if  $min \neq i$  then swap a[i] and a[min];

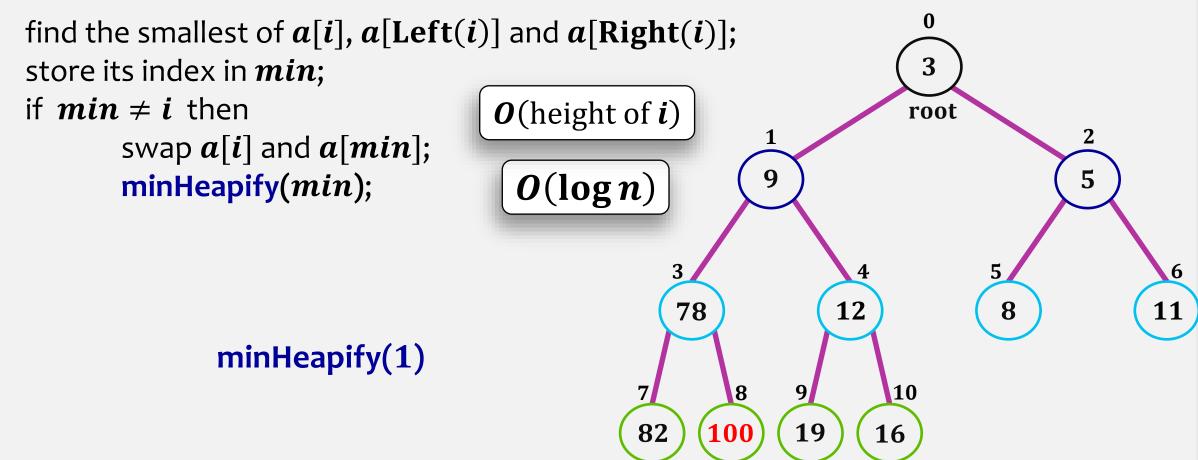
minHeapify(1)

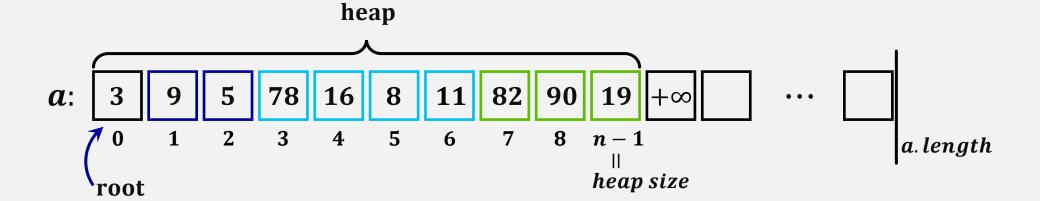
minHeapify(min);

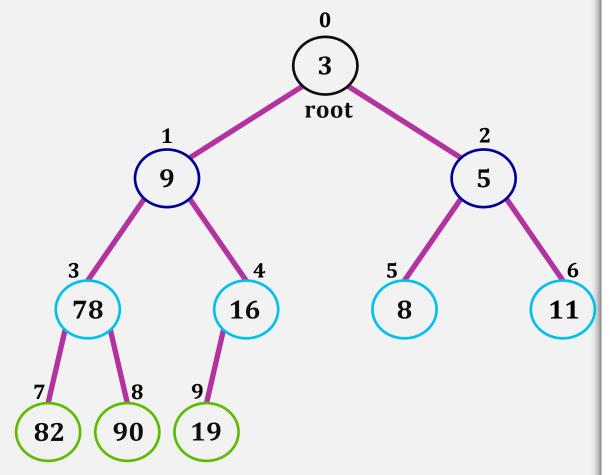


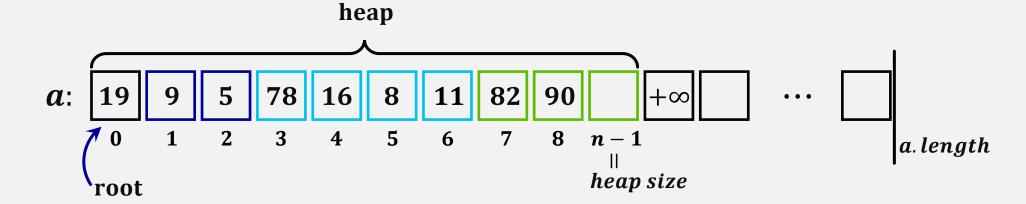


#### minHeapify(i):

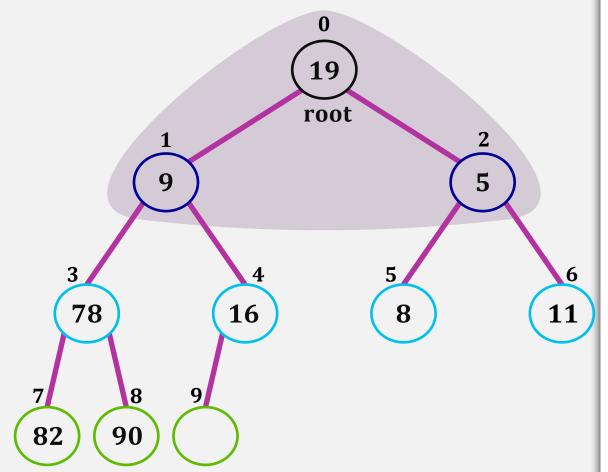


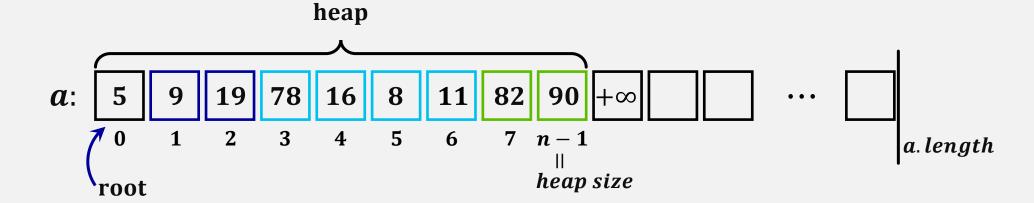




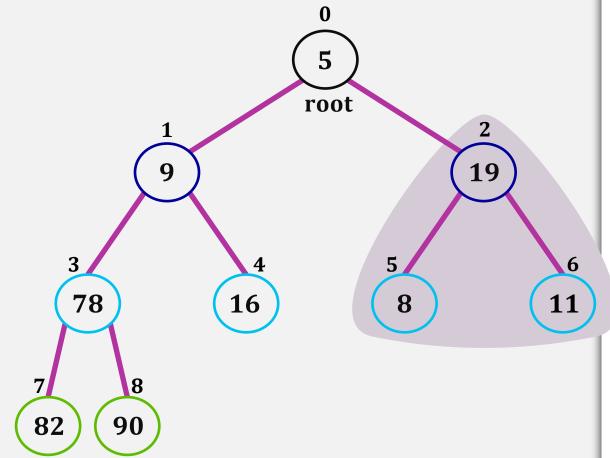


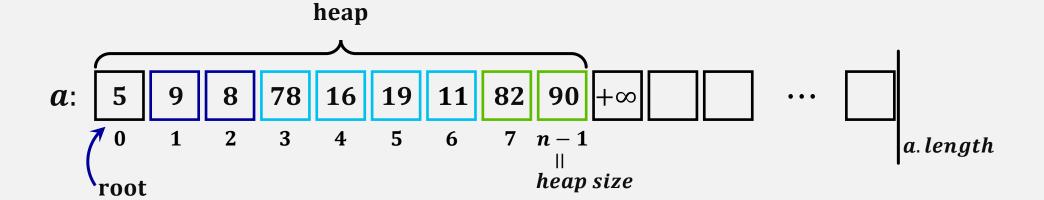
run minHeapify(0);



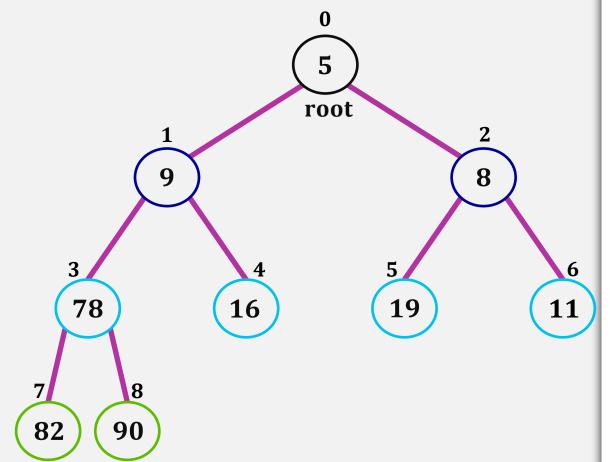


run minHeapify(0);





run minHeapify(0);



**removeMin()** – return and delete the smallest element in heap  $a[0..n-1], n \ge 0$  and restore the heap property.

```
min = a[0];
a[0] = a[n-1];
n--;
minHeapify(0);
if (3n < a.length) then resize();
return min;
```

$$O(1) + O(\log n) = O(\log n)$$
minHeapify(0)

### Theorem 10.1

A **BinaryHeap** implements the (priority) **Queue** interface. Ignoring the cost of calls to resize(), a **BinaryHeap** supports the operations add(x) and removeMin() in  $O(\log n)$  time per operation, and findMin() in O(1) time per operation.

Furthermore, beginning with an empty **BinaryHeap**, any sequence of m add(x) and removeMin() operations results in a total of O(m) time spent during all calls to resize().