

# Priority Queues

# Priority Queues

## Problem.

- Let  $S = \{(s_1, p_1), (s_2, p_2), \dots, (s_n, p_n)\}$  where  $s(i)$  is a key and  $p(i)$  is the priority of  $s(i)$ .

How to design a data structure/algorithm to support the following operations over  $S$ ?

**ExtractMin:** Returns the element of  $S$  with minimum priority

**Insert( $s, p$ ):** Insert a new element  $(s, p)$  in  $S$

**RemoveMin:** Remove the element in  $S$  with minimum  $p$

# Priority Queues

Solution 1. Used a sorted list

ExtractMin :

Insert:

DeleteMin:

Solution 2. Use a list with the pair with minimum  $p$  at the first position

ExtractMin :

Insert:

DeleteMin:

Can we do better? How?

# Priority Queues

Solution 1. Used a sorted list

ExtractMin :  $O(1)$  time

Insert:  $O(n)$  time

DeleteMin:  $O(1)$  time

Solution 2. Use a list with the pair with minimum  $p$  at the first position

ExtractMin :  $O(1)$  time

Insert:  $O(1)$  time

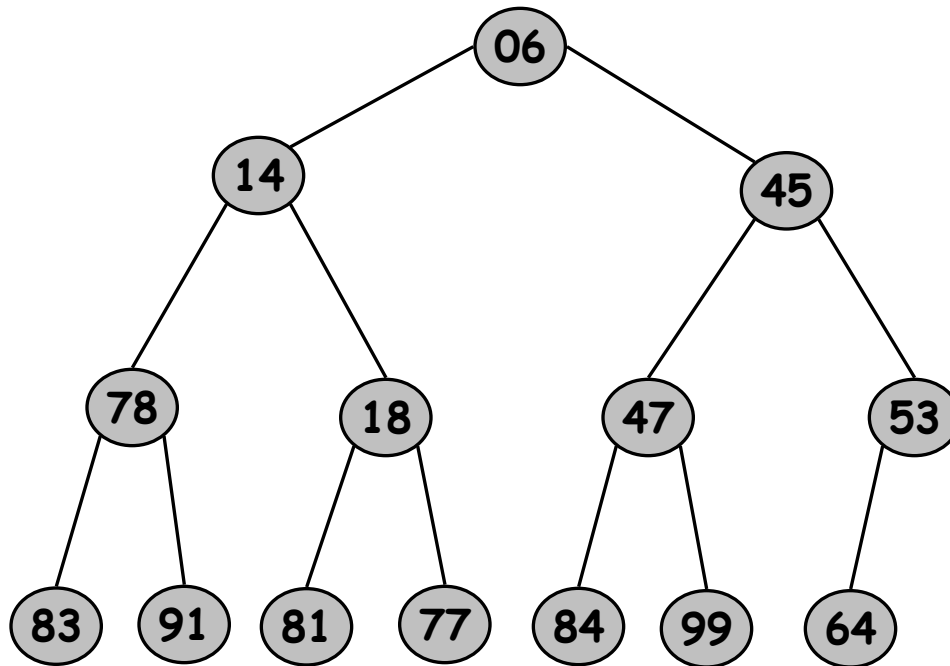
DeleteMin:  $O(n)$  time

Can we do better? How?

# Binary Heap: Definition

## Binary heap.

- Almost complete binary tree
  - filled on all levels, except last, where filled from left to right
- Min-heap ordered
  - every child greater than (or equal to) parent



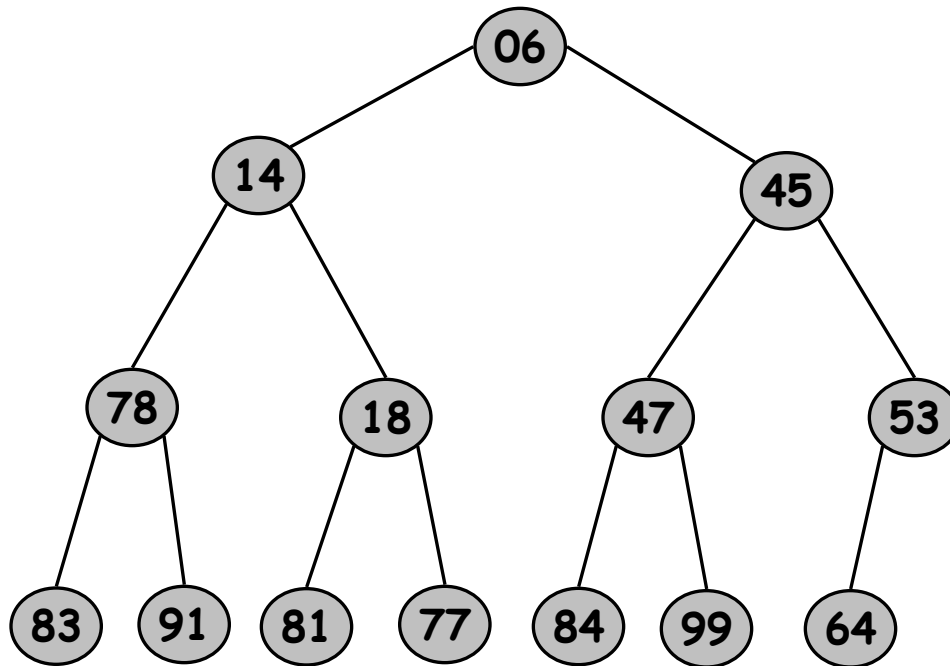
Max-heap is analogous (every child is smaller than its parent)

# Binary Heap: Properties

## Properties.

- Min element is in
- Heap with N elements has height

In particular can obtain minimum element (ExtractMin) in  $O(1)$



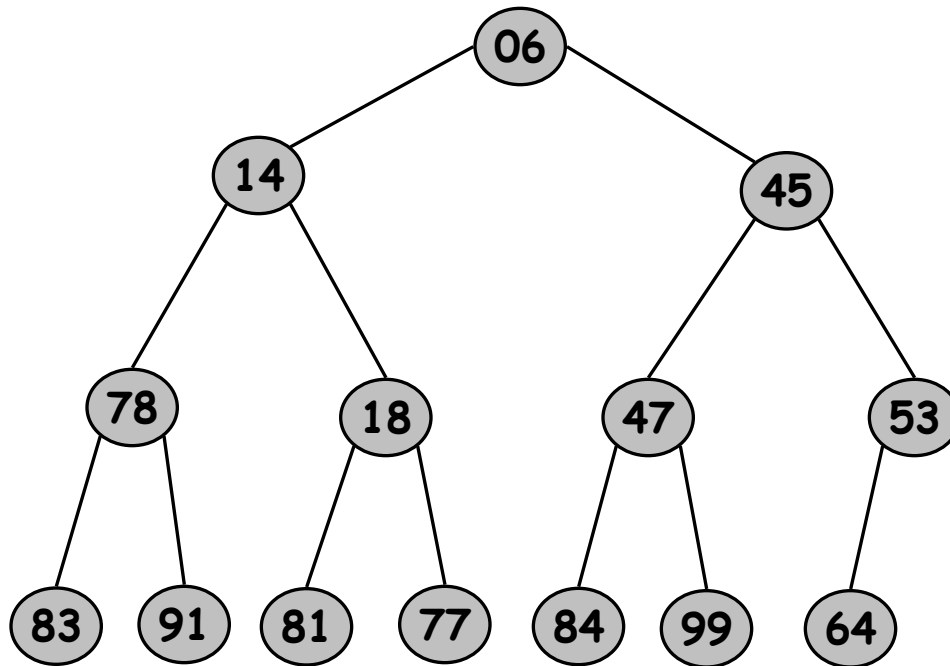
**N = 14**  
**Height = 3**

# Binary Heap: Properties

## Properties.

- Min element is in root.
- Heap with N elements has height =  $\lfloor \log_2 N \rfloor$ .

In particular can obtain minimum element (ExtractMin) in  $O(1)$



**N = 14**  
**Height = 3**

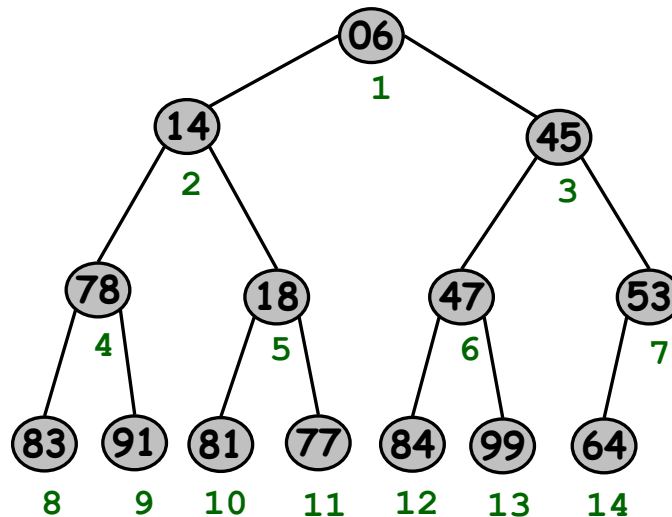
# Binary Heaps: Array Implementation

## Implementing binary heaps

- Use an array: no need for explicit parent or child pointers.
  - $\text{Parent}(i) = \lfloor i/2 \rfloor$
  - $\text{Left}(i) = 2i$
  - $\text{Right}(i) = 2i + 1$

Only important properties for us:

1. Last level of the tree is occupied **from left to right**
2. We can find the first (leftmost) **empty space in constant time** (keep track of first empty space with an integer firstEmpty)

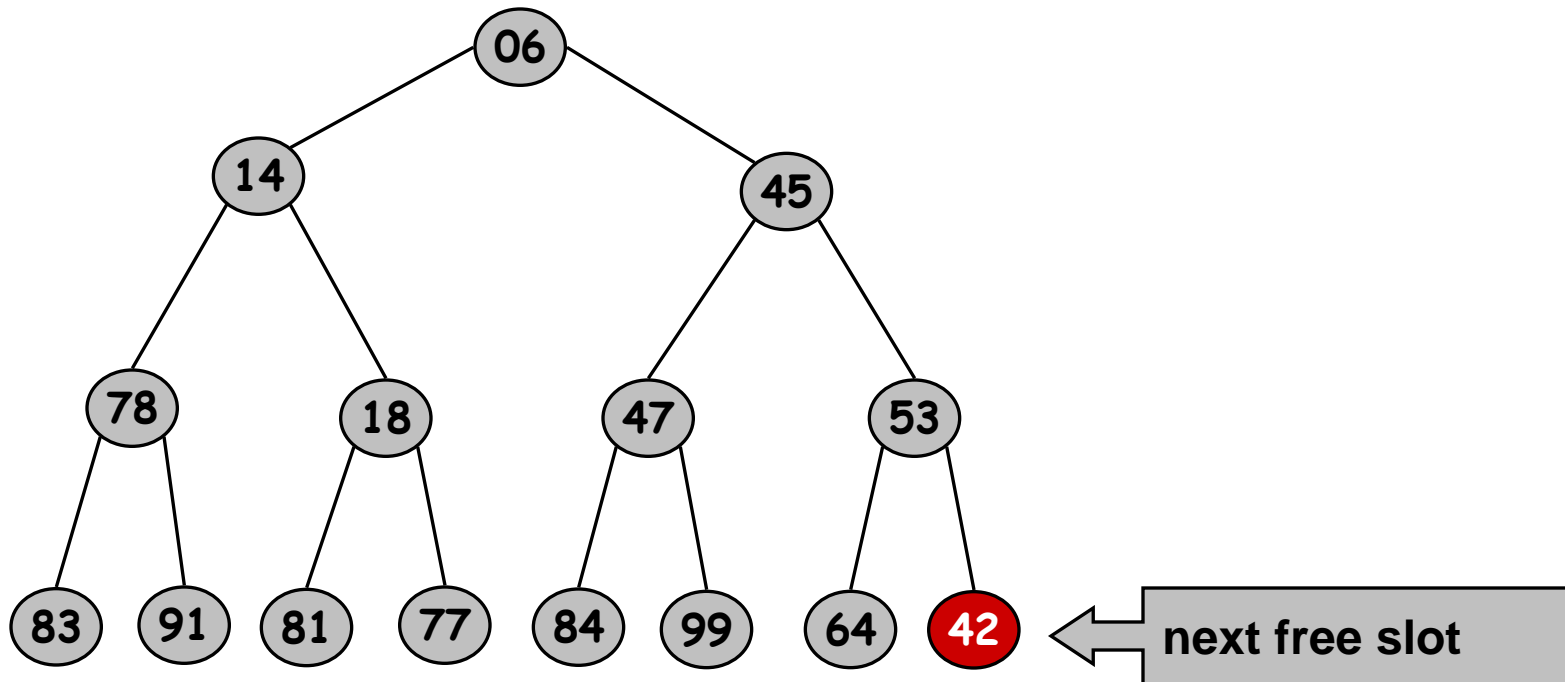




# Binary Heap: Insertion

Insert element  $x$  into heap.

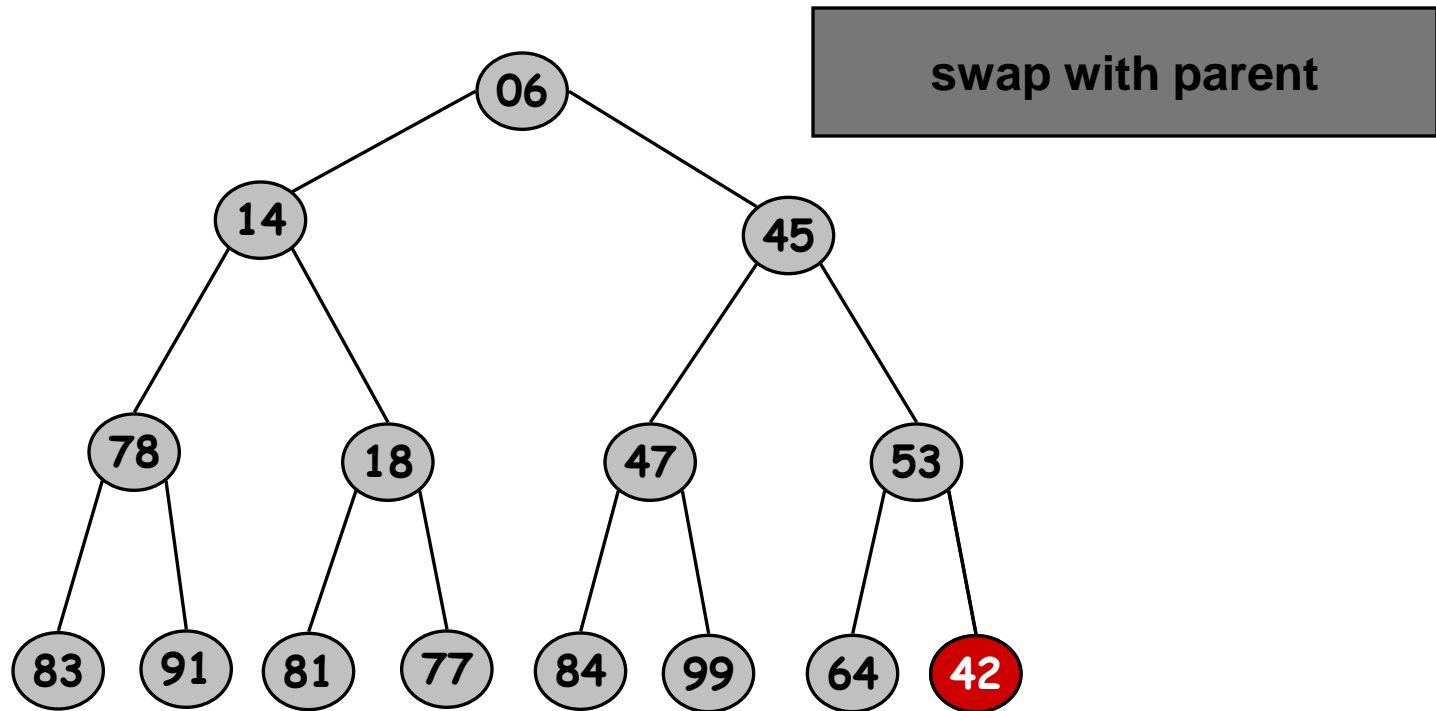
- Insert into first available slot.
- Bubble up until it's heap ordered.



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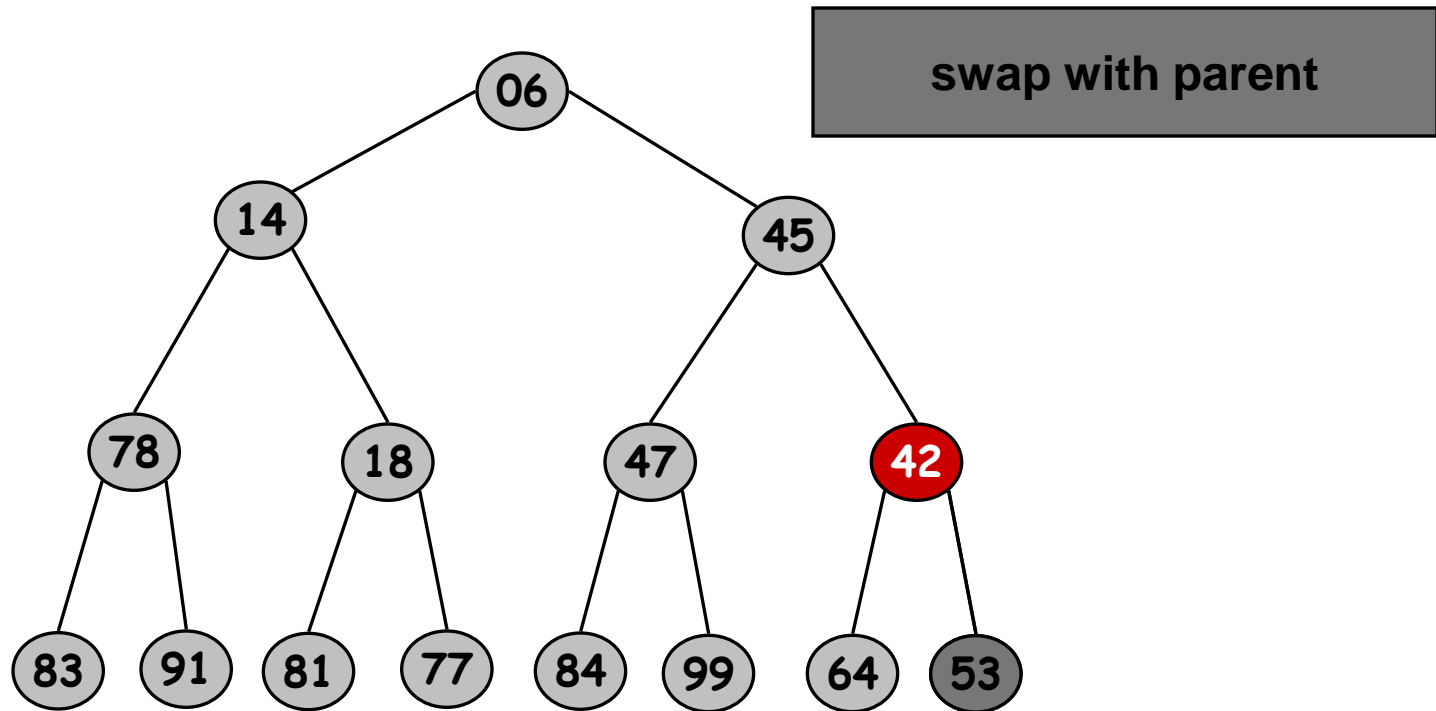
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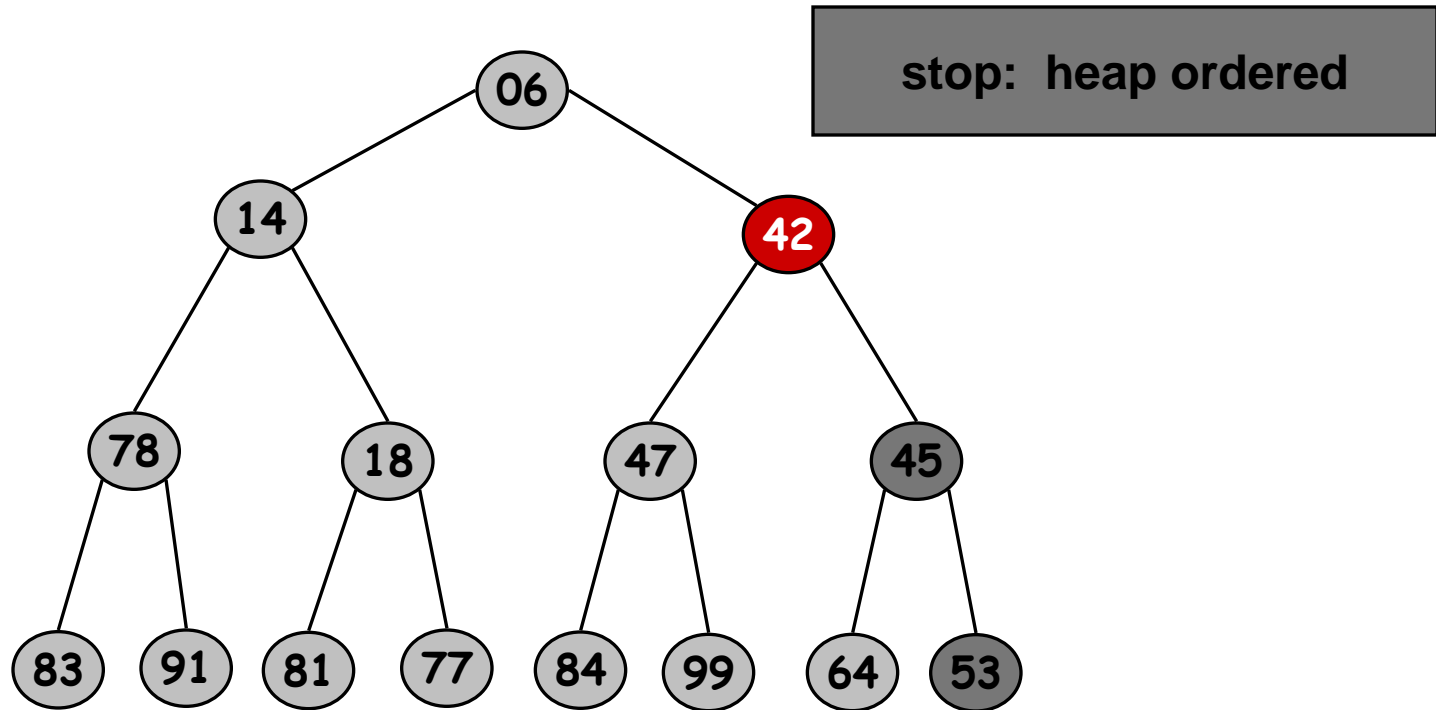
# Binary Heap: Insertion

Insert element  $x$  into heap.

- Insert into first available slot.
- Bubble up until it's heap ordered.

Q: What is the time complexity of Insert?

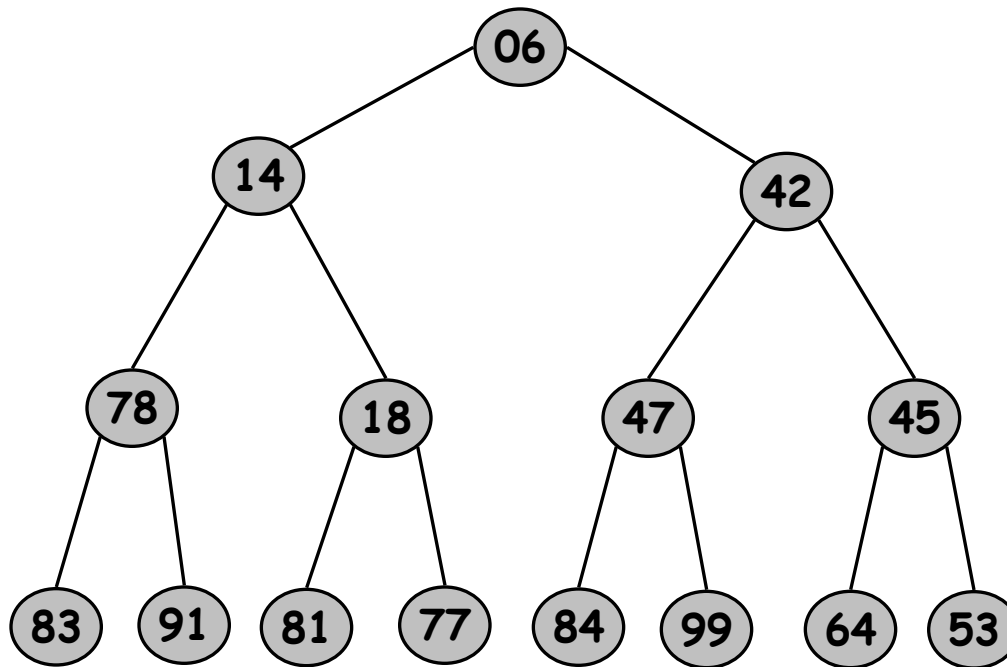
A:  $O(\log N)$  operations



## Binary Heap: Decrease Key

Decrease key of element  $x$  to  $k$ .

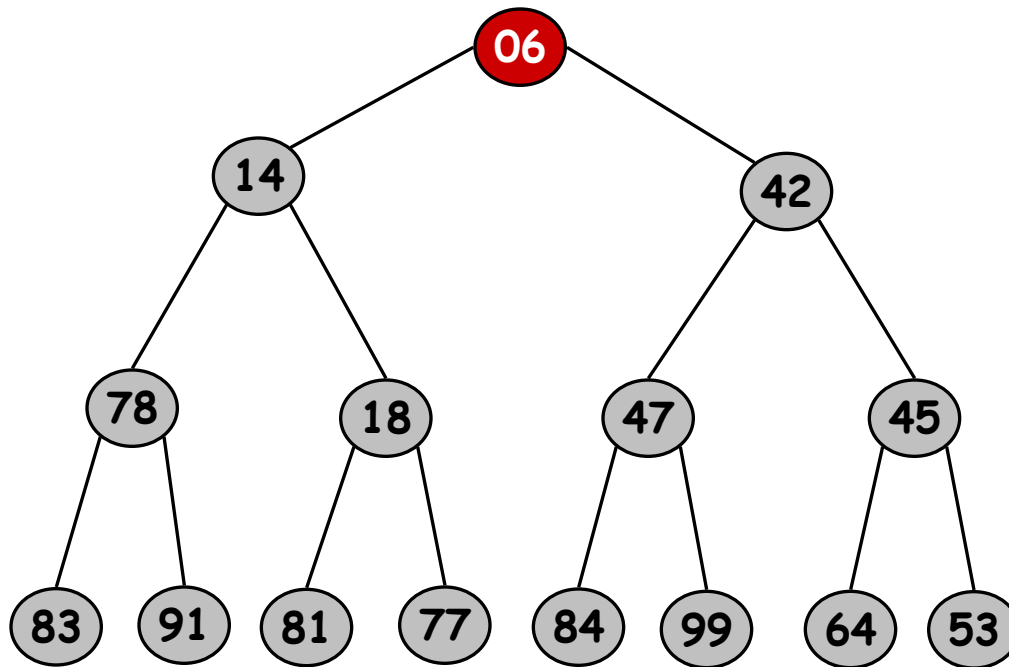
- Bubble up until it's heap ordered.
- $O(\log N)$  operations.



# Binary Heap: Delete Min

Delete minimum element from heap.

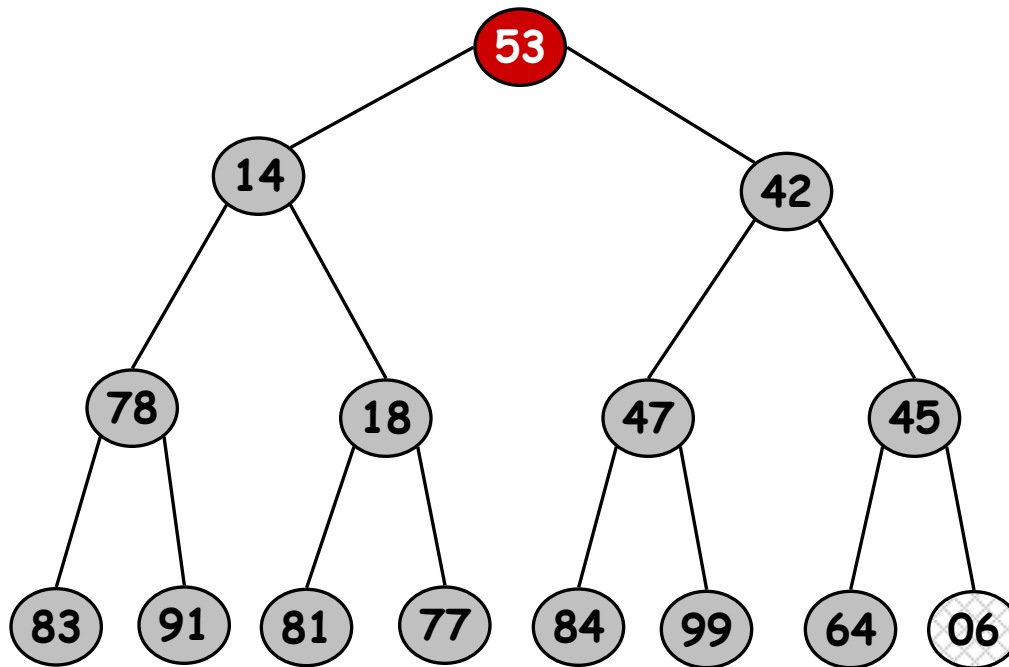
- Exchange root with last (rightmost) leaf
- Delete this leaf
- Bubble root down until it's heap ordered, exchanging it with **smallest** child



## Binary Heap: Delete Min

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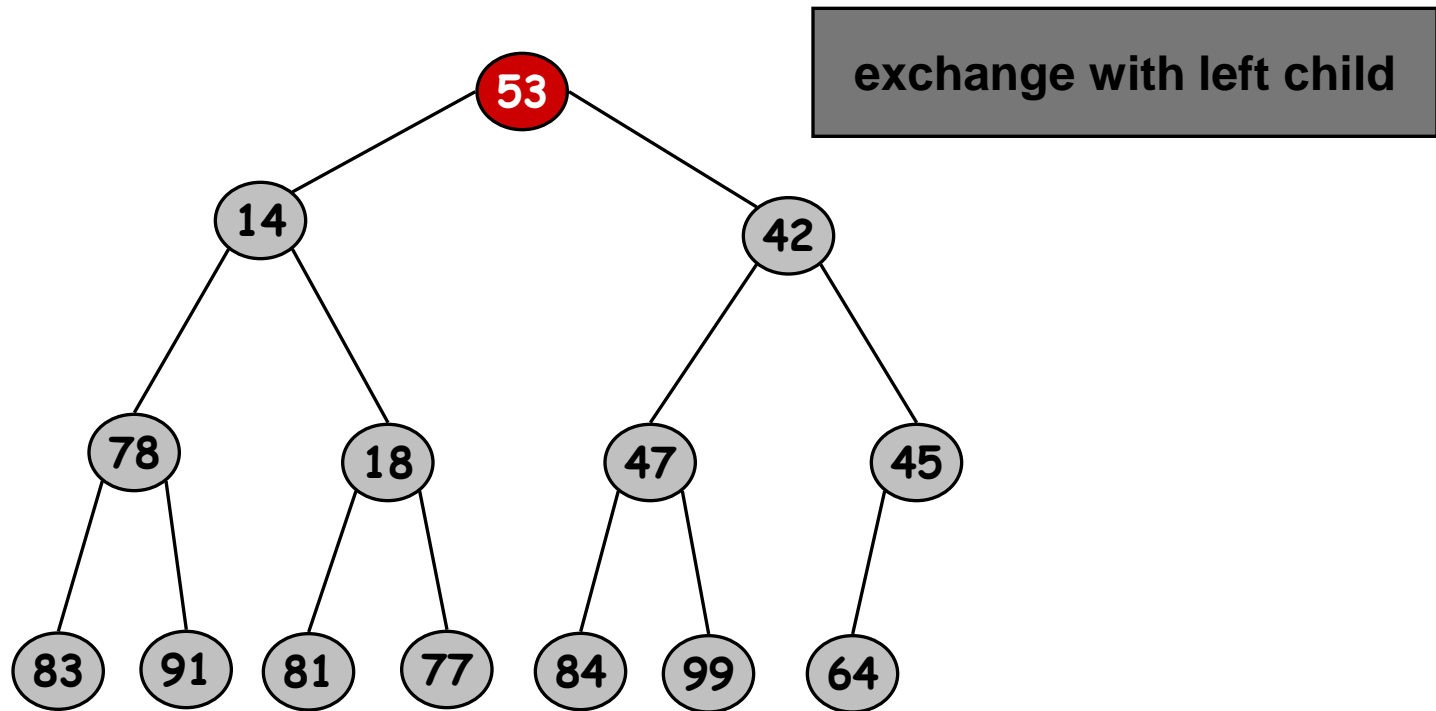
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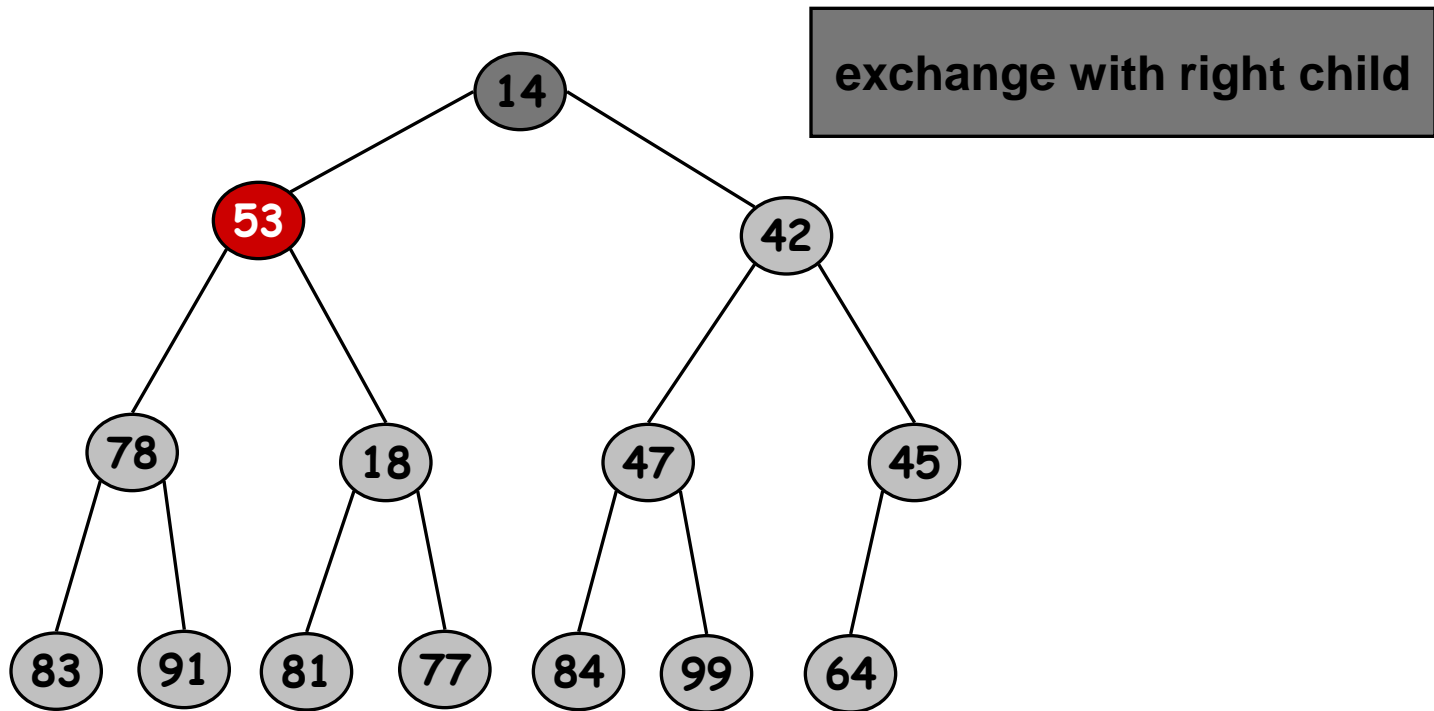




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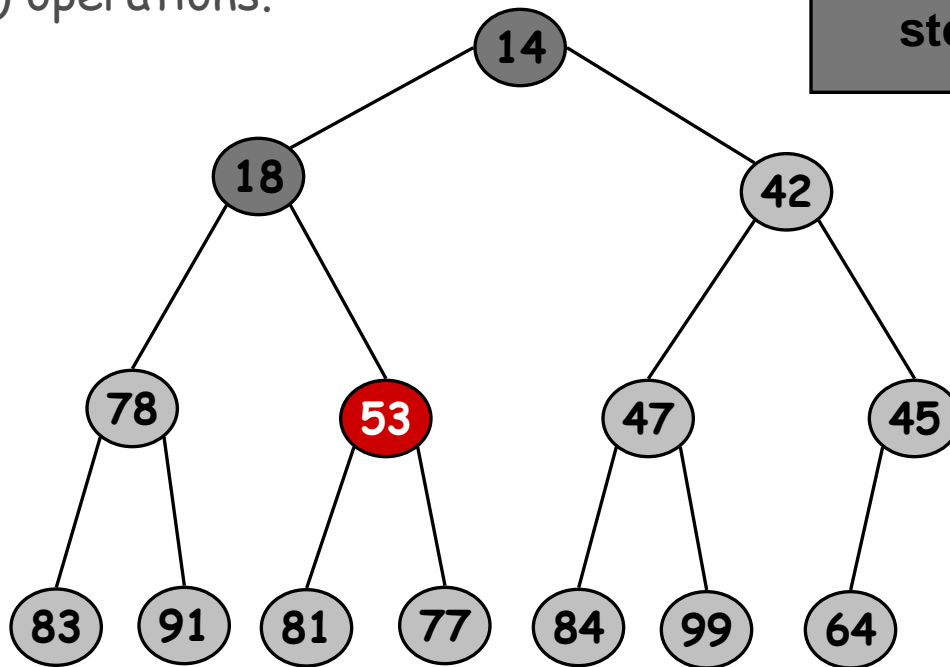
# Binary Heap: Delete Min

Delete minimum element from heap.

- Exchange root with last (rightmost) leaf
- Delete this leaf
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Q: What is the time complexity of DeleteMin?

A:  $O(\log N)$  operations.



# Priority Queues

Solution 1. Use a sorted list

ExtractMin:  $O(1)$  time

Insert:  $O(n)$  time

DeleteMin:  $O(1)$  time

Solution 2. Use a list with the pair with minimum  $p$  at the first position

ExtractMin:  $O(1)$  time

Insert:  $O(1)$  time

DeleteMin:  $O(n)$  time

Solution 3. Binary Heap

ExtractMin:  $O(1)$  time

Insert:  $O(\log n)$  time

DeleteMin:  $O(\log n)$  time

## Application: Heapsort

Q: Can we sort  $N$  numbers using a Binary Heap?

- Insert  $N$  items into a binary min-heap
  - $O(N \log N)$
- Perform  $N$  delete-min operations.
  - $O(N \log N)$
- Overall
  - $O(N \log N)$  sort.

Obs: If you use a max-heap can do with no extra storage (in-place)

## Exercise

- Run the execution of Heapsort over list of numbers  
18 25 41 34 10 52 50 48

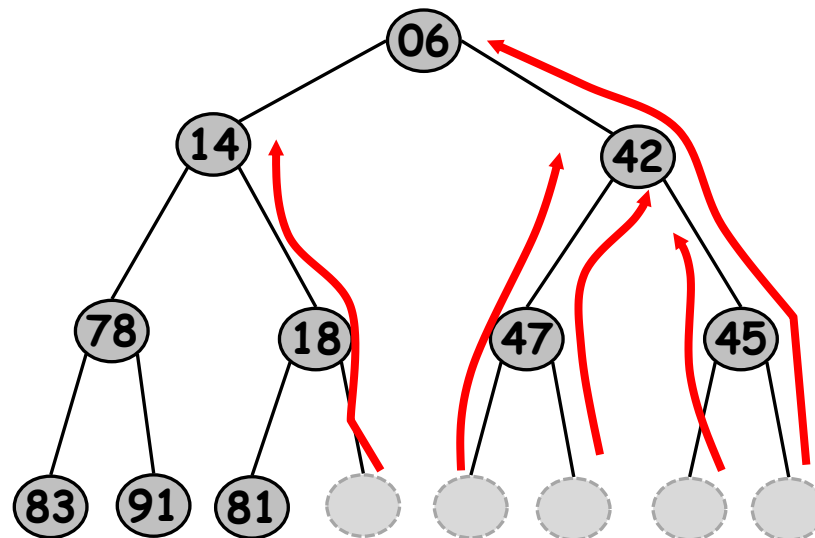
# Building a Heap

In Heapsort, we just built a heap for  $n$  numbers by inserting them one at a time

Time complexity of this procedure:  $O(n \log n)$

Q: Can we build a heap faster?

**Bottleneck:** When adding the last few nodes, each may move a lot up in the heap, costs too much



# Building a Heap in $O(n)$

Idea: Move nodes **down** instead of up

Operation `down(node)`: keeps exchanging node with smallest child until it is in the right position, that is, has smaller priorities than its children

Building a Heap of numbers  $a_1, a_2, \dots, a_n$  in  $O(n)$ :

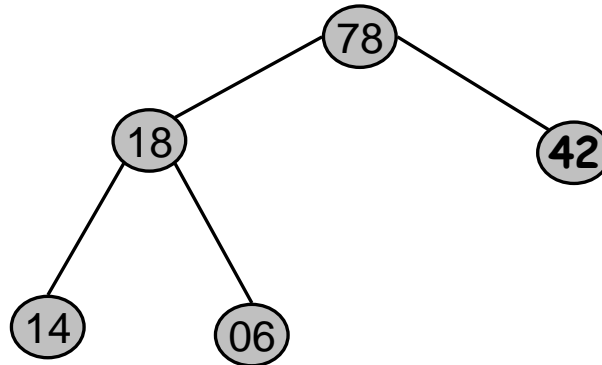
HeapBuild1 (Cormen)

- Make an almost complete binary tree with numbers in any position
- Traverse tree from bottom up applying `down(node)`

# Building a Heap in $O(n)$

Example: Numbers 14, 06, 42, 78, 18

Apply *down()* to  
node with 06

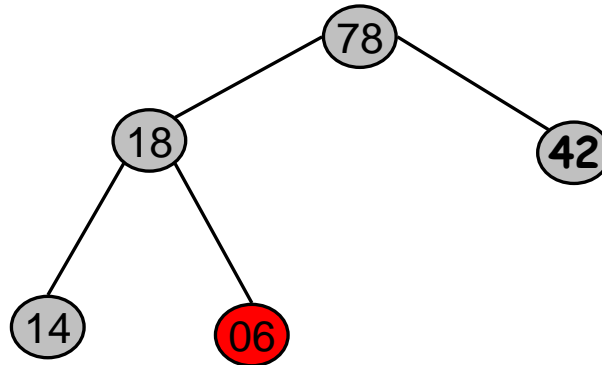




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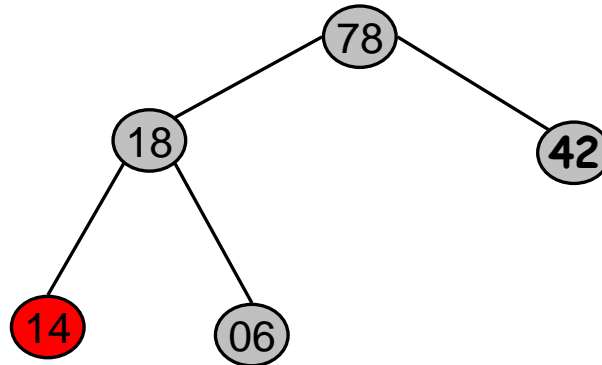
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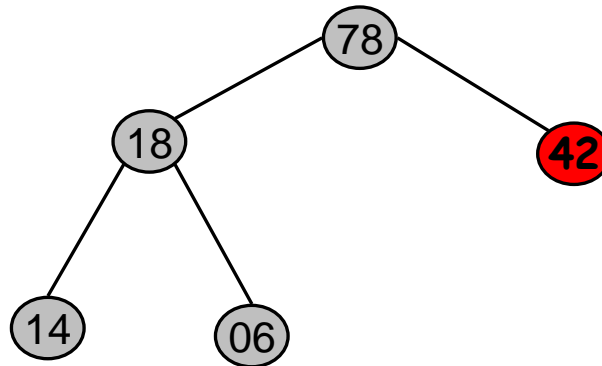
Apply *down()* to node with 14



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Example: Numbers 14, 06, 42, 78, 18

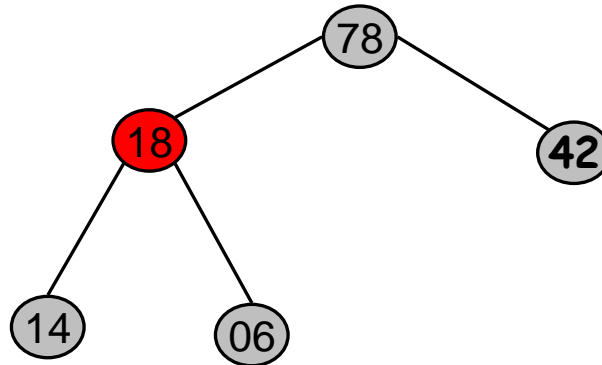
Apply *down()* to  
node with 42



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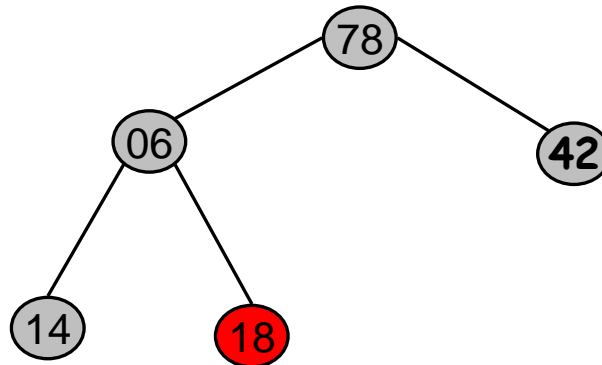
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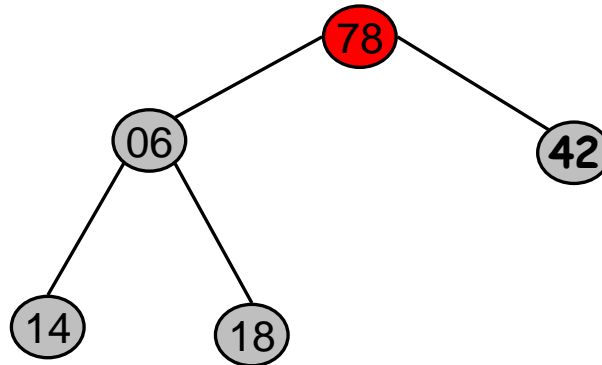
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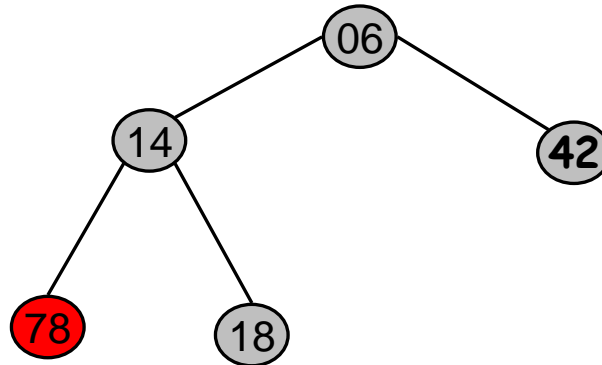
Apply *down()* to  
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# Building a Heap in $O(n)$

Example: Numbers 14, 06, 42, 78, 18

Apply *down()* to node with 78 twice



# Building a Heap in $O(n)$ : Analysis

Q: Quantos nós temos no nível  $i$  de baixo pra cima?

- **Observação:** Para  $i > 0$ , no  $i$ -ésimo nível do heap de baixo para cima temos no máximo  $n / 2^{i-1}$  nós
  - O heap tem  $\lfloor \log_2 n \rfloor$  níveis. No nível 0, de baixo para cima, temos no mínimo 1 e no máximo  $N$  nós.
  - Todos os níveis exceto os dois últimos tem metade dos nós do nível imediatamente abaixo



# Building a Heap in $O(n)$ : Analysis

*Custo do down(node)*: se node está no nível  $i$  de baixo para cima, esse down custa  $O(i)$  operações

$$\text{Custo total} = \sum_{i=1}^{\log n} i \cdot \frac{n}{2^i} = \text{[ ]}$$

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