

# AVL Trees Heaps And Complexity

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CSC212 — Fall 2014

# Complexity Of BST Operations

or

"Why Should  
We Use BST  
Data Structures"

# What We're After:

- What's the **worst** amount of work we can expect
  - when we insert?
  - when we delete?
  - when we search, successfully or unsuccessfully?
- What's the **average** amount of work we can expect?
- We'd like to know about the **best** case, but in general it doesn't occur that often.

# Which Two-Operations In BSTs are Similar?



- Search/Find  $\left\{ \begin{array}{l} \bullet \text{ Successful} \\ \bullet \text{ Unsuccessful} \end{array} \right.$
- Insert
- Delete

# Worst Case for *Inserts* or Unsuccessful *Searches*

- **Definition:** The depth of a node  $x$ ,  $d(x)$ , is the # of nodes from the root to that node.  $d(x) = 0$ -based level of the node plus 1.
- What is the worst possible number of nodes visited for searching or inserting in a BST?
- So, what is the worst-case complexity for insert or unsuccessful operations?

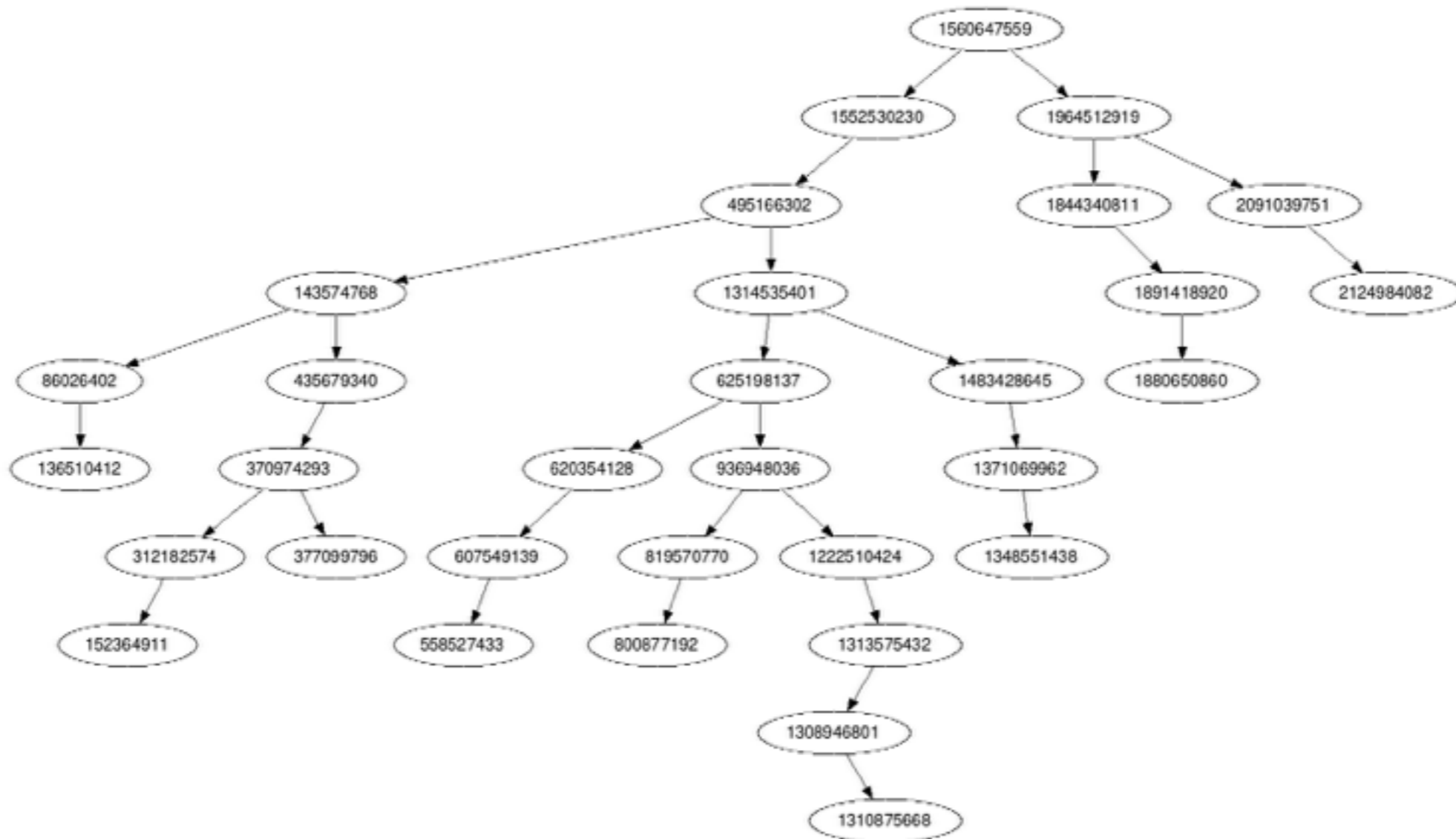
Operation	Outcome	Worst Case Complexity	Average Complexity	Best Complexity
<i>Insertion</i>	<i>X</i>	<b><i>O(N)</i></b>		
<i>Search</i>	<i>Successful</i>			
	<i>Unsuccessful</i>	<b><i>O(N)</i></b>		
<i>Deletion</i>	<i>X</i>			

Best-Case complexity  
for search?

Operation	Outcome	Worst Case Complexity	Average Complexity	Best Complexity
<i>Insertion</i>	<i>X</i>	$O(N)$		
<i>Search</i>	<i>Successful</i>			<b><math>O(1)</math></b>
	<i>Unsuccessful</i>	$O(N)$		<b><math>O(1)</math></b>
<i>Deletion</i>	<i>X</i>			



# Average Case for *Successful Searches*



# Need More Definitions


- BST of  $N$  nodes
- $key_i$  = key residing in node  $x_i$ ,  $i=1, 2, \dots, N$
- $d(x_i)$  = depth of node  $x_i$
- $p_i$  = probability of searching for  $key_i$

What is the average  
number of nodes visited  
when searching a BST  
of  $N$  nodes?



$$D_{avg}(N) = \sum_{i=1}^N d_i p_i$$

So, we need to know  $p_i \dots$

- if all the keys are **equally likely**, then all the  $p_i$  are identical.
- if the  $p_i$  are **not identical**, then some keys are more likely than others, and we can take advantage of this  *self-adjusting trees* (Section 6.8)

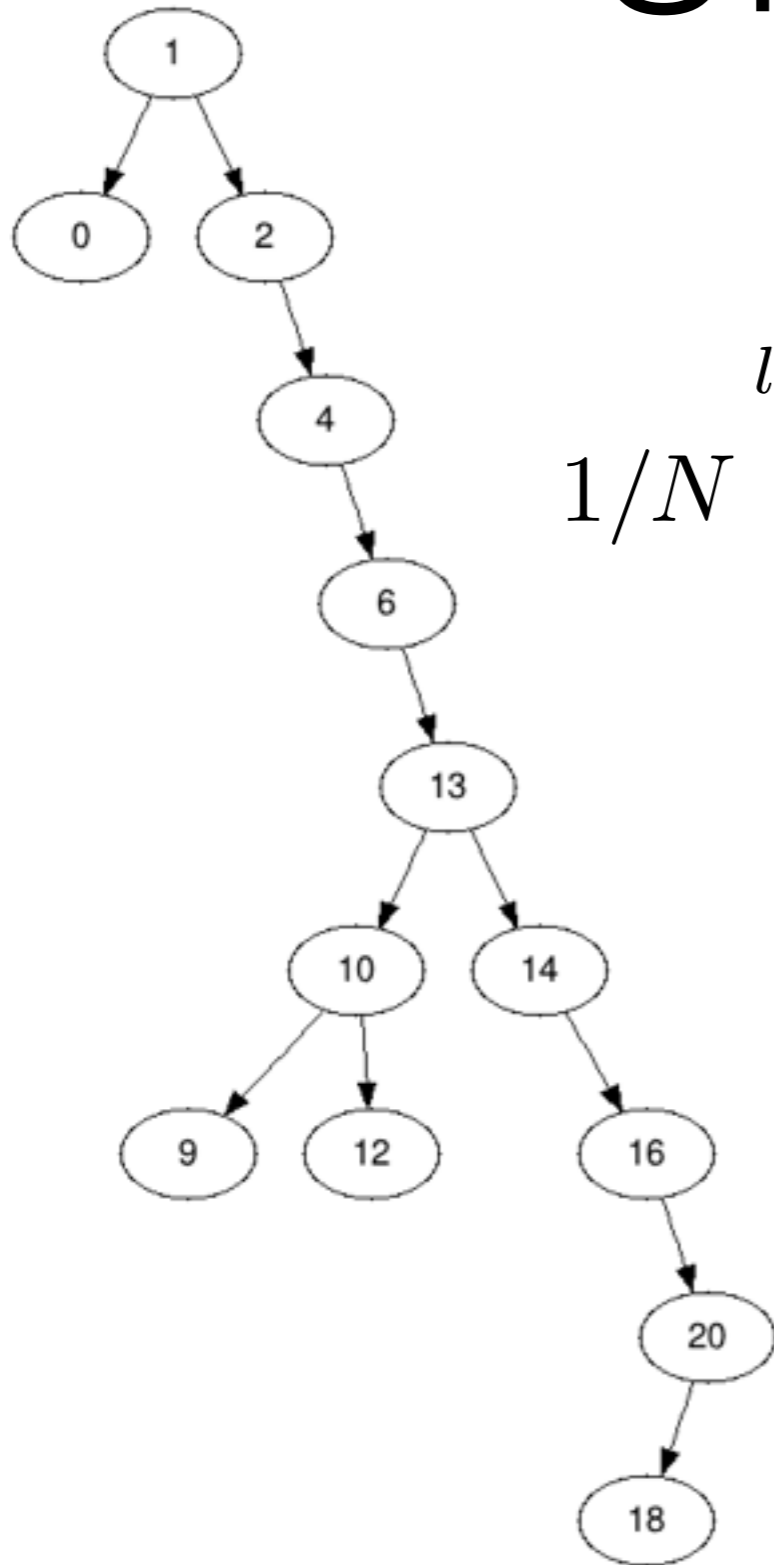
# *Equally Likely* Keys

- equally likely keys:  $p_i = 1/N$

$$D_{avg}(N) = 1/N \sum_{i=1}^N d_i = \text{total node depth}$$

- $D_{avg}(N)$  depends on tree shape!

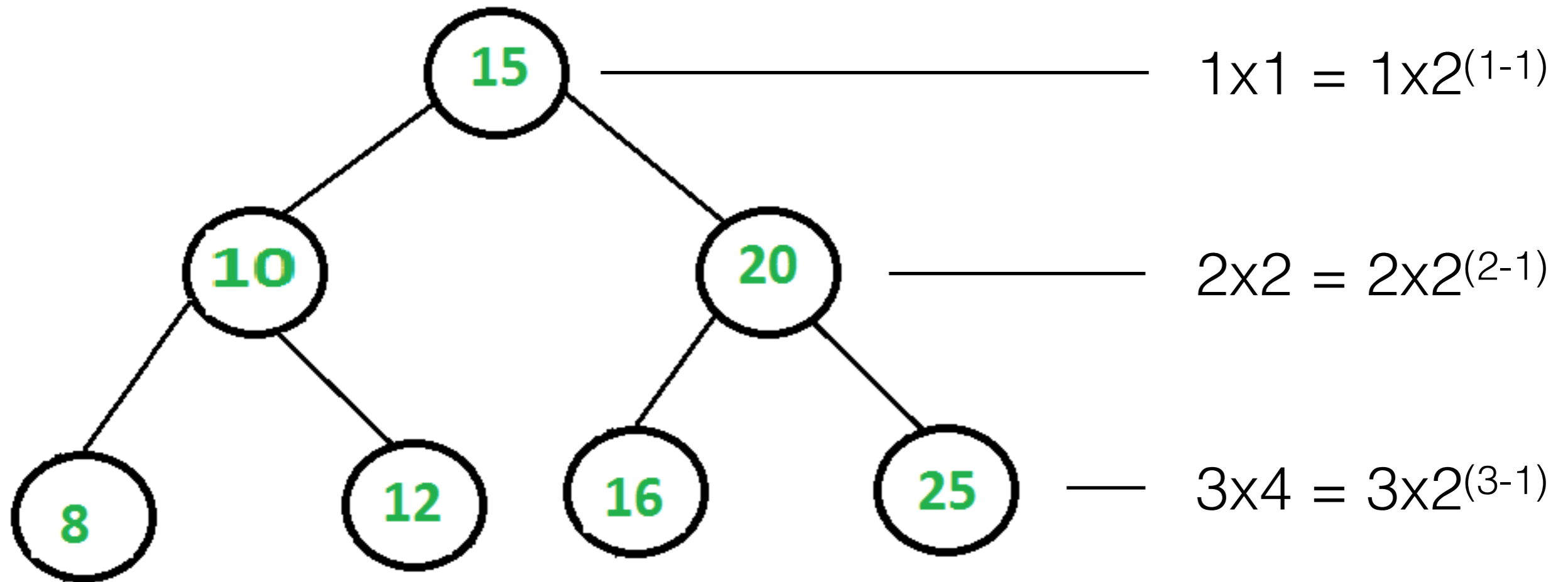
# Unbalanced Tree



$$\frac{1}{N} \sum_{i=1}^{\log(N+1)} d_i p_i = \frac{1}{N} (1 + 2 + 3 + \dots + N)$$

= ?

# Fully Balanced Tree



.....  $i \times 2^{(i-1)}$

$$\frac{1}{N} \sum_{i=1}^{\log(N+1)} i 2^{(i-1)} < \log(N+1) = O(\log N)$$

Successful Search:  $\left\{ \begin{array}{ll} \times O(N) & \text{Unbalanced Trees} \\ \times O(\log N)? & \text{Balanced Trees} \end{array} \right.$





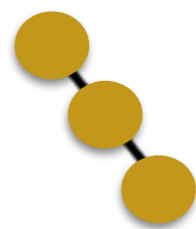
But...

Are average  
BSTs tall or  
fat?

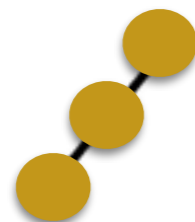
What about Successful Search  
in a Random BST?

# What about Successful Search in a Random BST?

- We need to look at all possible shapes the BST of  $N$  nodes can have and compute the average number of probes required for each successful search, and average over all possible shapes of the BST...
- Huge amount of combinations!



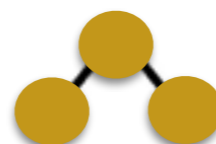
$$1+2+3=6$$



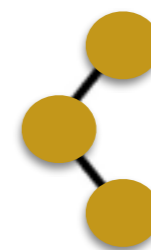
$$1+2+3=6$$



$$1+2+3=6$$



$$1+2+2=5$$



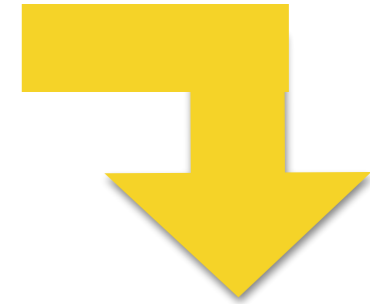
$$1+2+3=6$$

Different BSTs of  
3 nodes  
and lengths of  
different paths for each

Average # of Probes  
for Successful Searches in a "random" BST  
of  $N$  nodes:

$$D_{avg}(N) \approx 1.386 \log_2 N$$

We don't really care



Operation	Outcome	Worst Case Complexity	Average Complexity	Best Complexity
<i>Insertion</i>	<i>X</i>	$O(N)$	$O(\log N)$	$O(1)$
<i>Search</i>	<i>Successful</i>	$O(N)$	$O(\log N)$	$O(1)$
	<i>Unsuccessful</i>	$O(N)$	$O(\log N)$	$O(1)$
<i>Deletion</i>	<i>X</i>	$O(N)$	$O(\log N)$	$O(1)$

Unlikely  
to happen



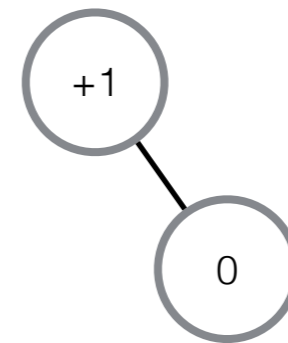
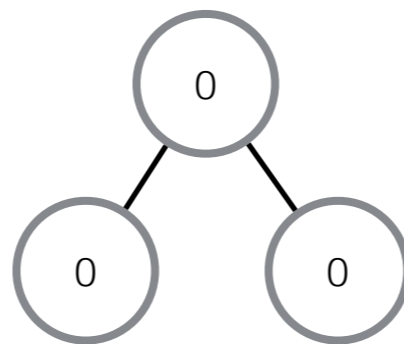
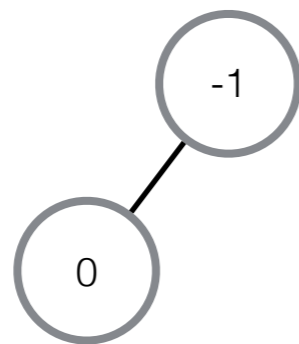
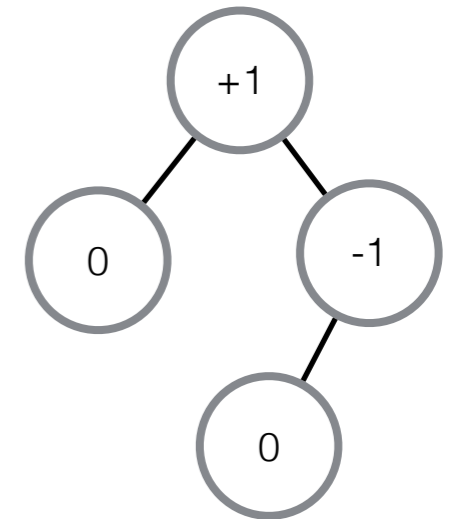
That's what  
we expect

# AVL-Trees

- Named for Adelson-Velskii and Landis

- 1962

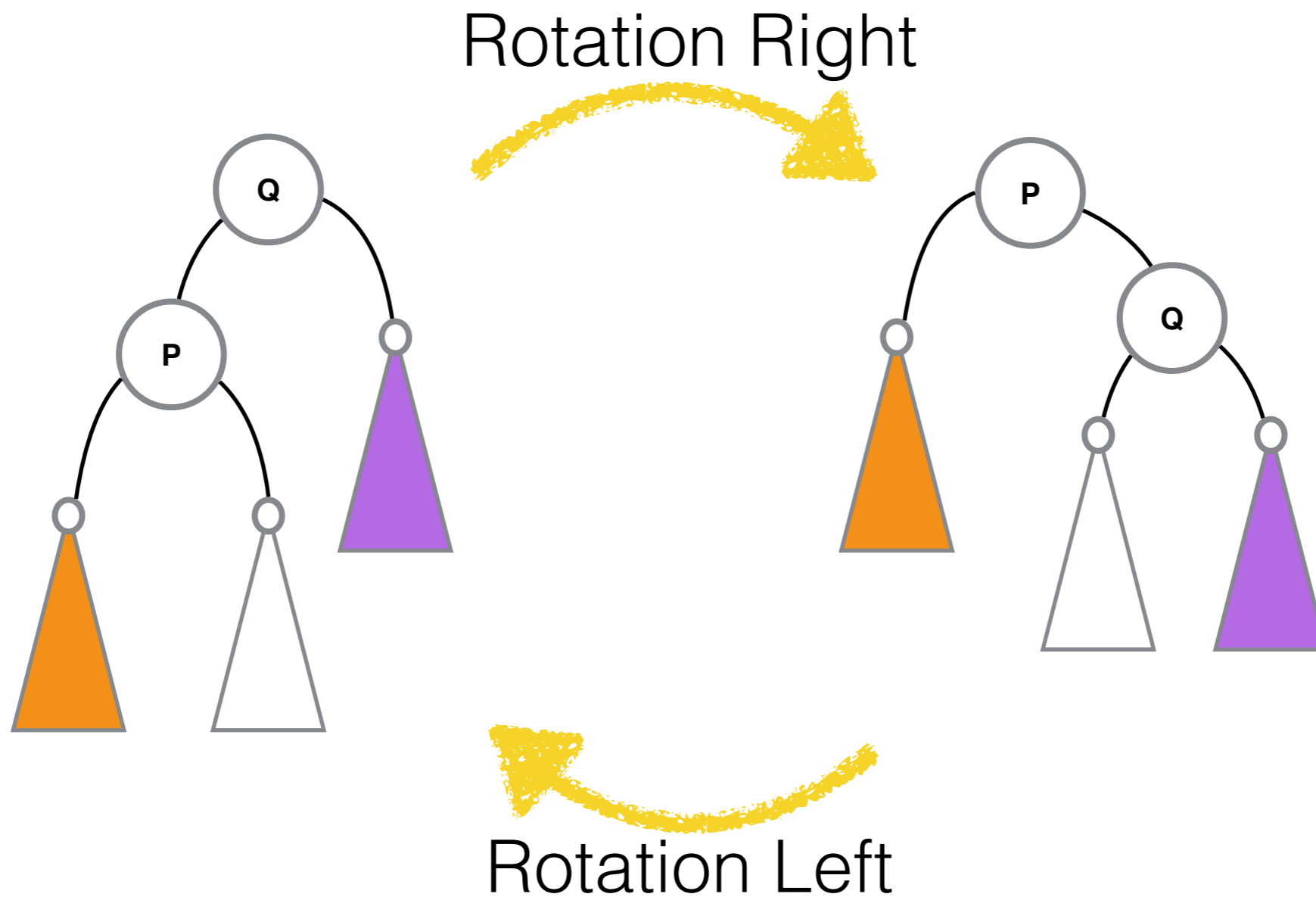
- Important property: For any node  $X$  in the tree, the heights of the left and right subtrees of  $X$  differ by at most 1





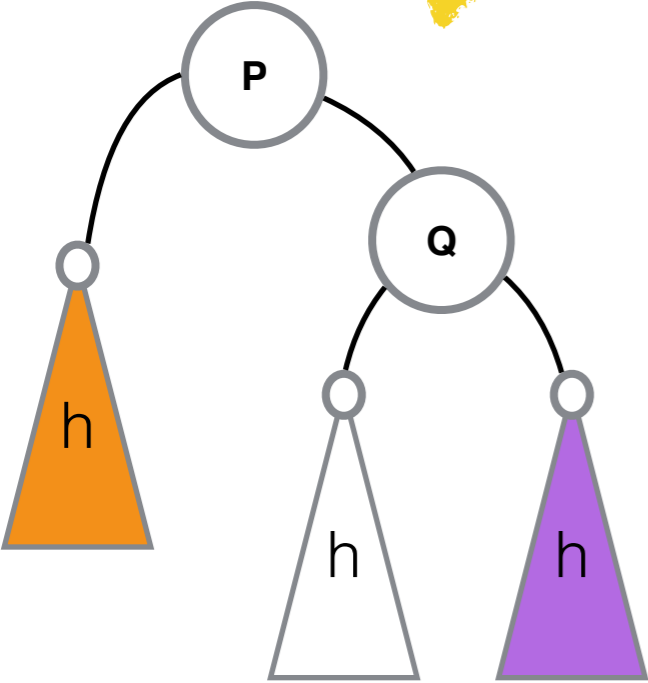
# AVL Trees

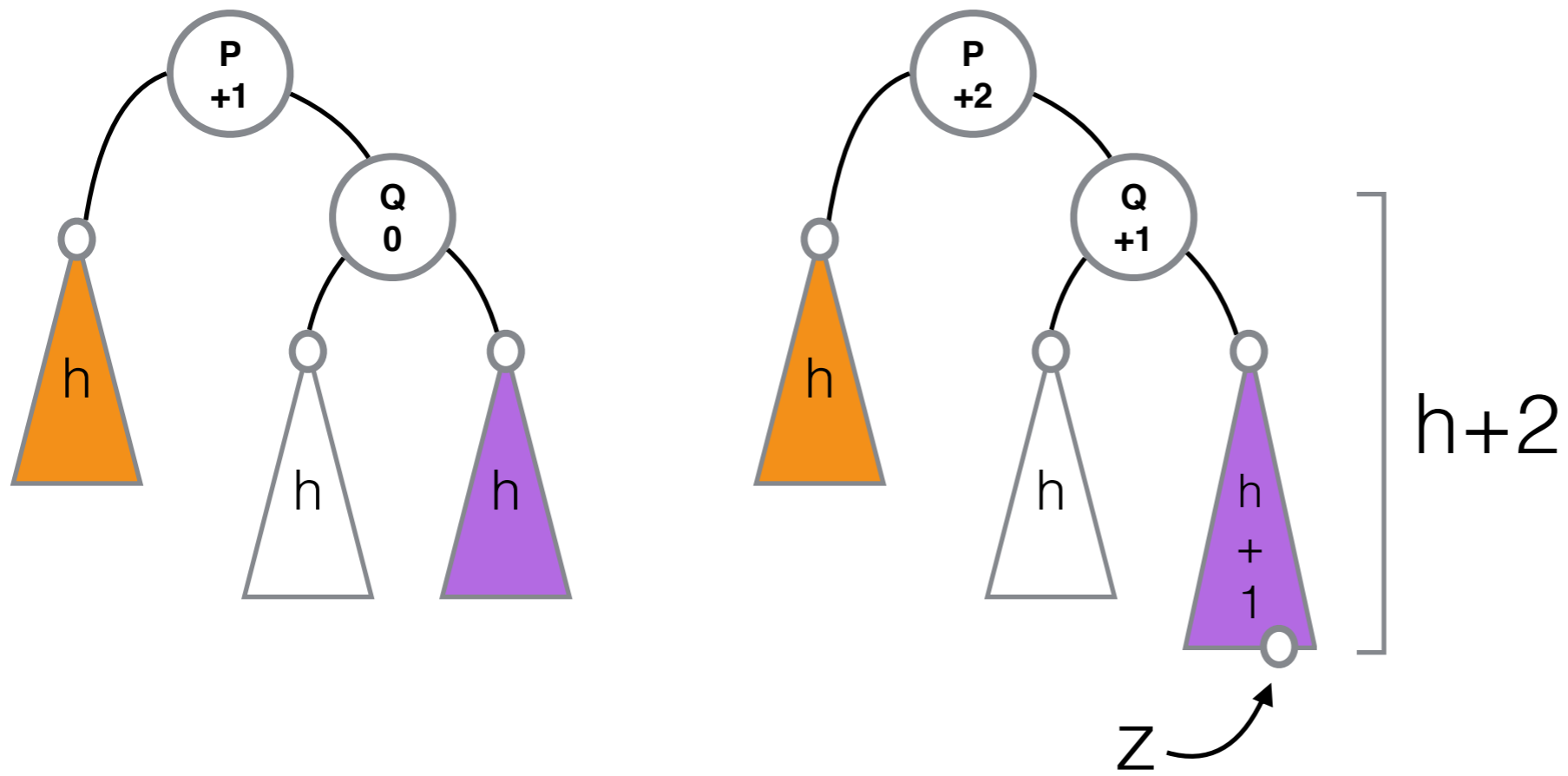
## Rely on *Rotations*



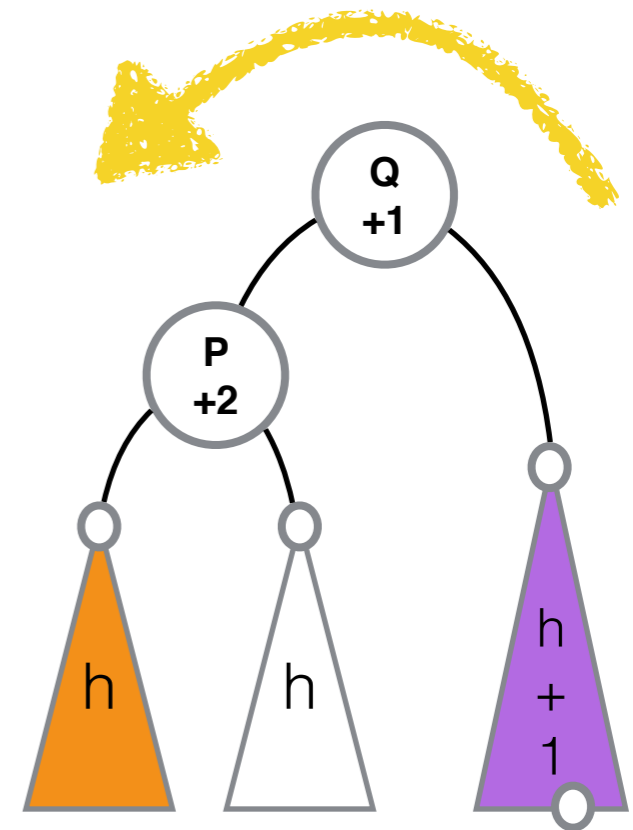
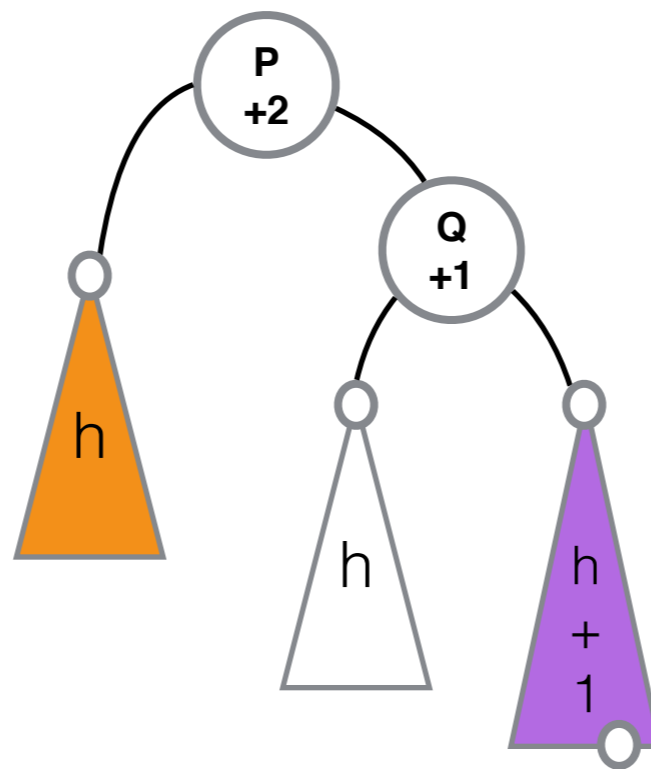
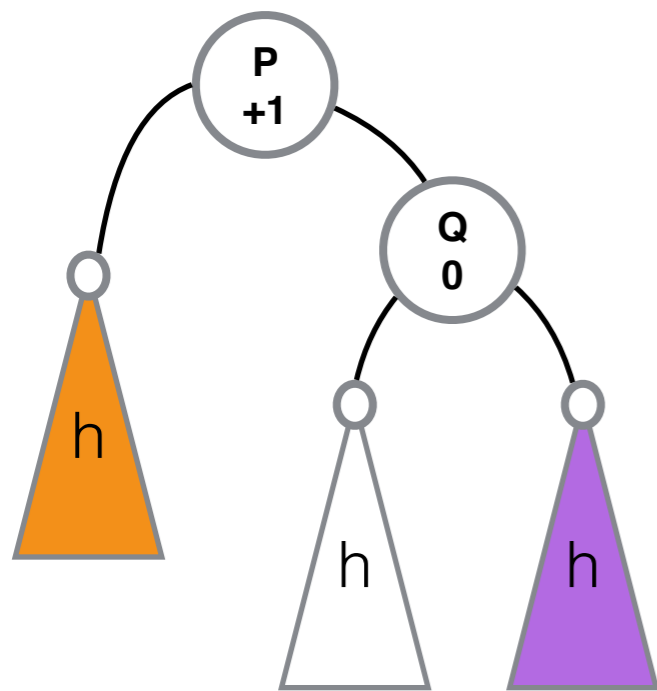


add 'Z'





# Rotation Left



# AVL Time Complexity

Operation	Worst Case Complexity	Average Complexity
<i>Insertion</i>	$O(\log N)$	$O(\log N)$
<i>Search</i>	$O(\log N)$	$O(\log N)$
<i>Deletion</i>	$O(\log N)$	$O(\log N)$

# Java

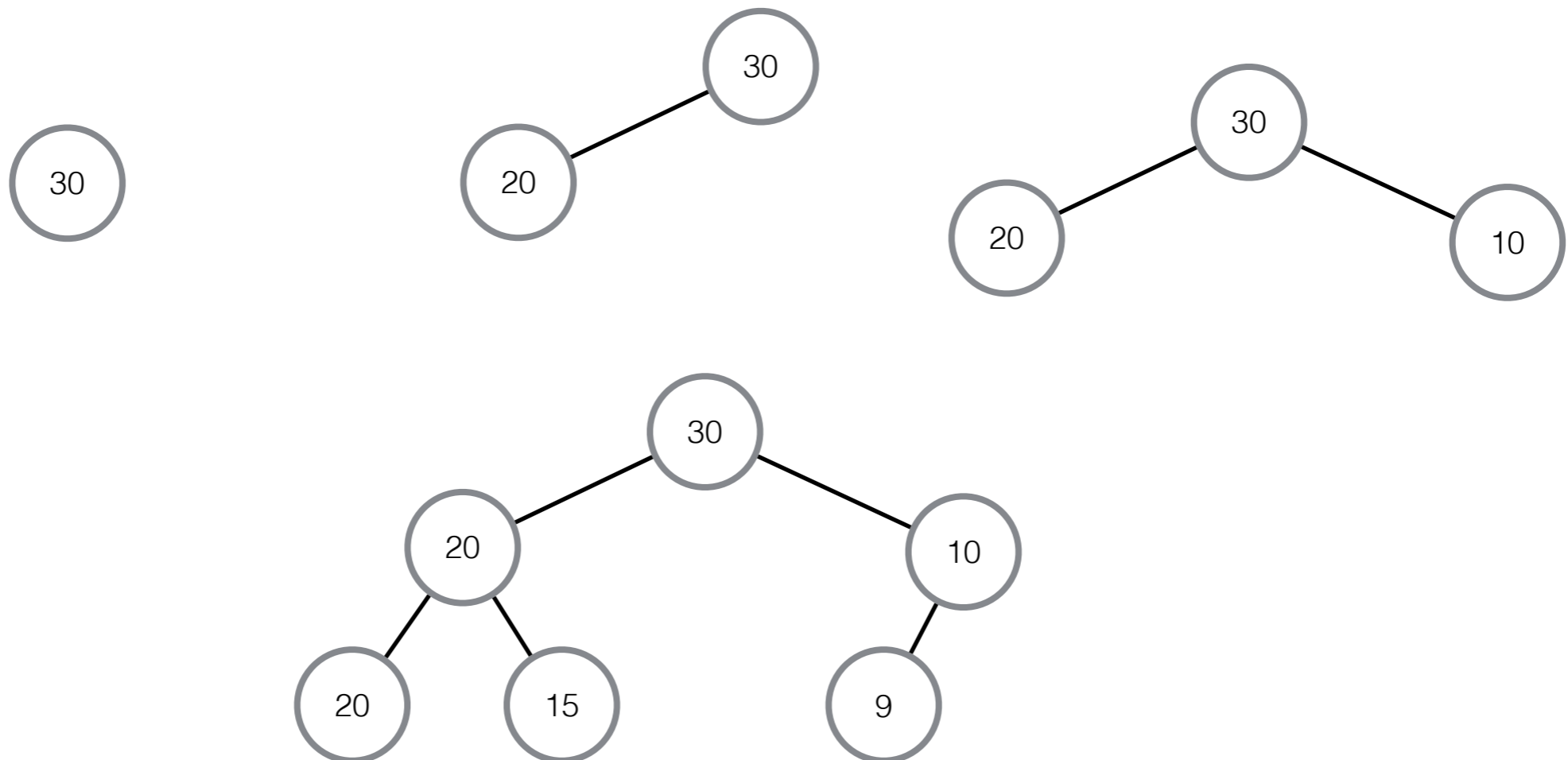
- Not that many "pure" trees
- **javax.swing.tree**
- **javax.swing.tree** TreeModel
- **javax.swing.tree** TreeNode

A large, conical pile of brown powder, possibly a mineral or chemical, is shown against a white background. The powder is finely textured and has a slightly uneven surface. The word "Heaps" is written in a bold, dark red font across the middle of the pile.

**Heaps**

A heap is:

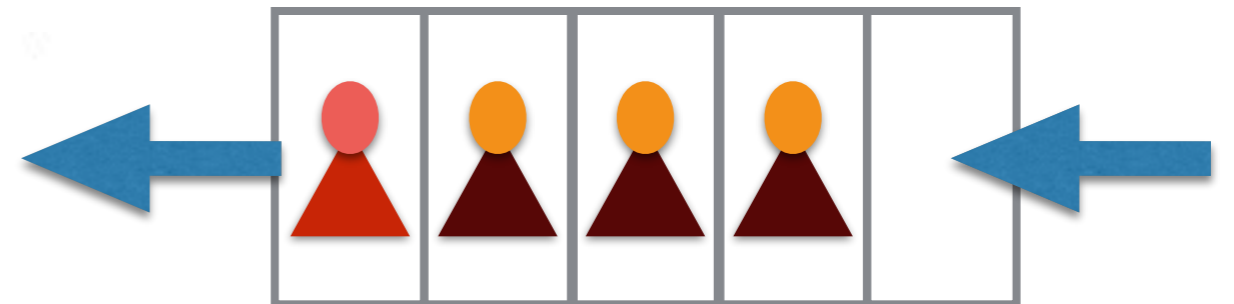
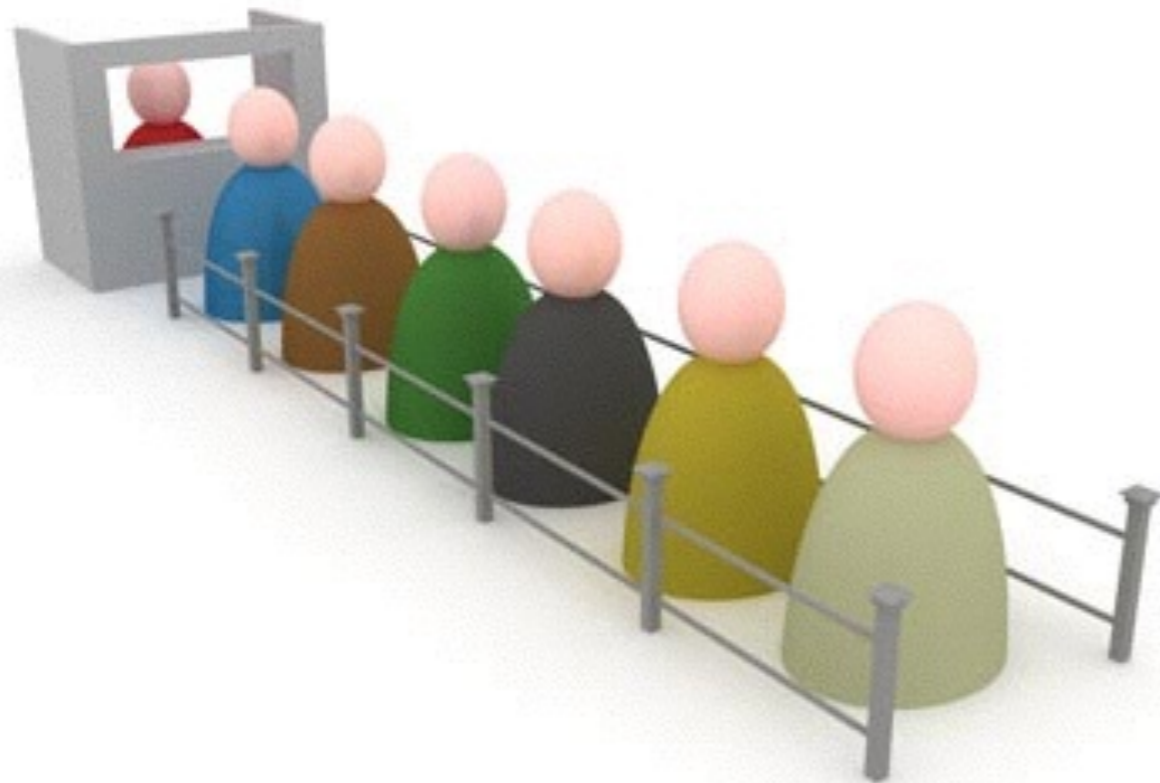
- A **fully balanced binary tree**, will all leaves on the **left-most** inner nodes
- The key of a *parent* is **larger** than or equal to the key of its *children*



# Typically used for...

- **Priority queues**: get the next element with the highest priority

07 07207

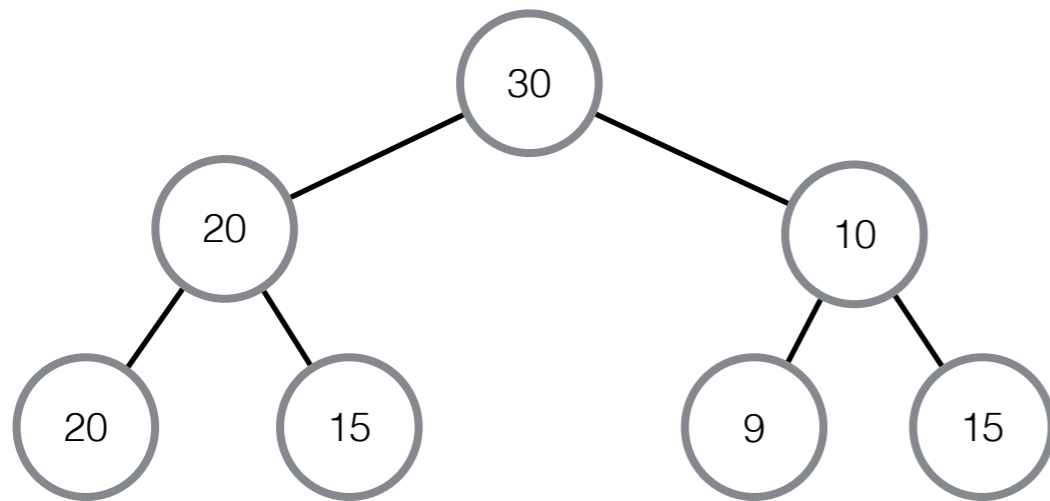


First-In...  
Highest-Priority Out



# Properties

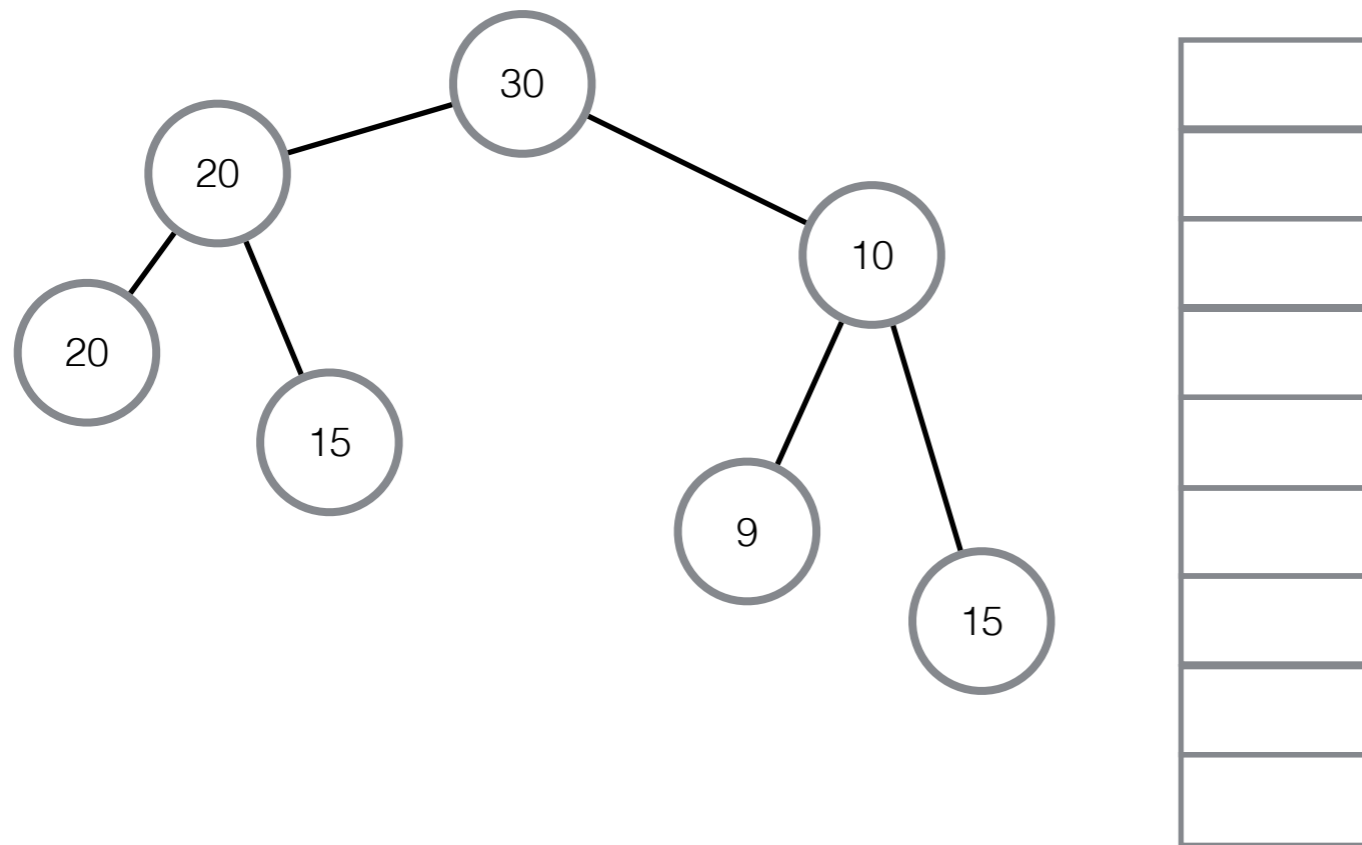
- The largest element is in the root, always
- The heap folds nicely...



What can you say of the keys other than the root?

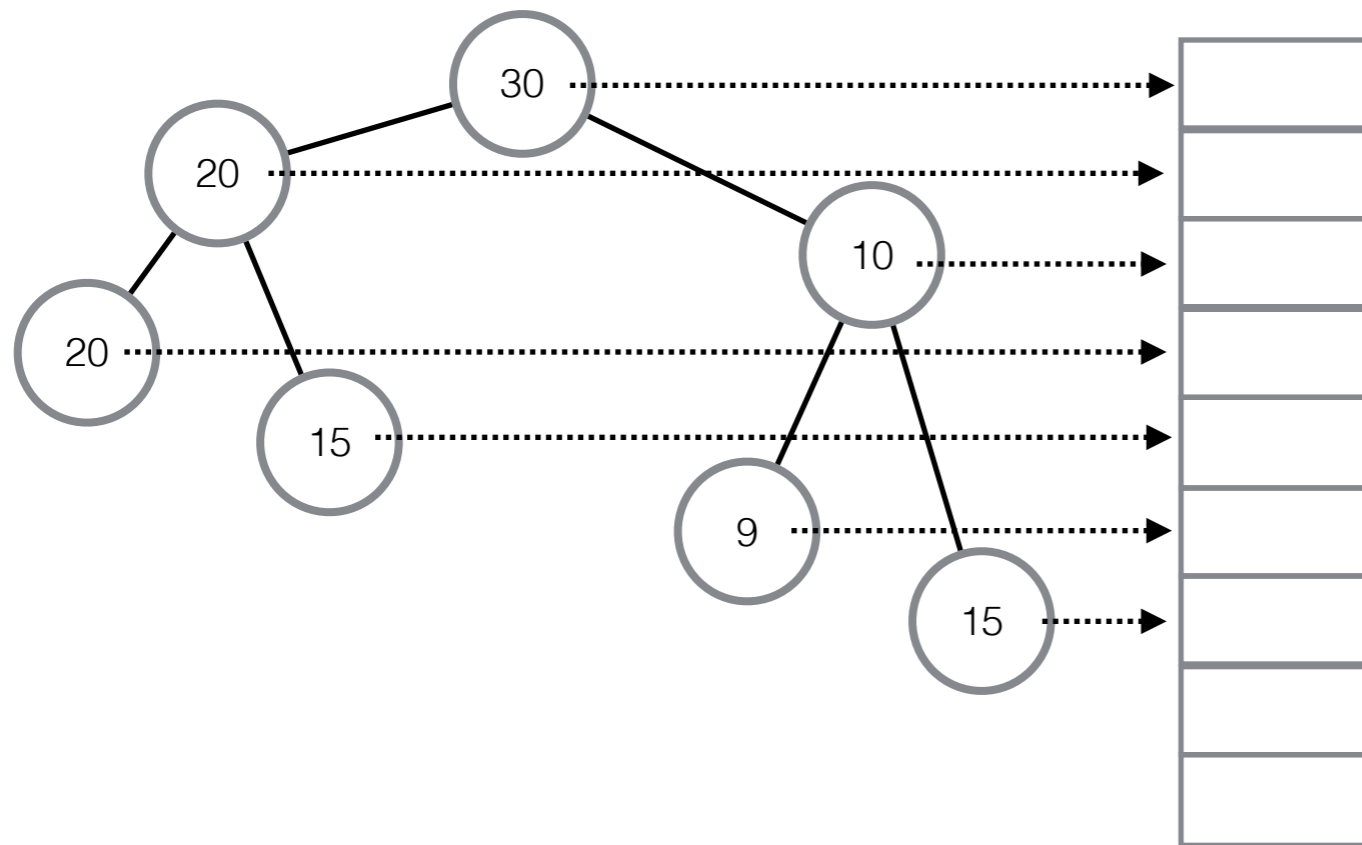
# Properties

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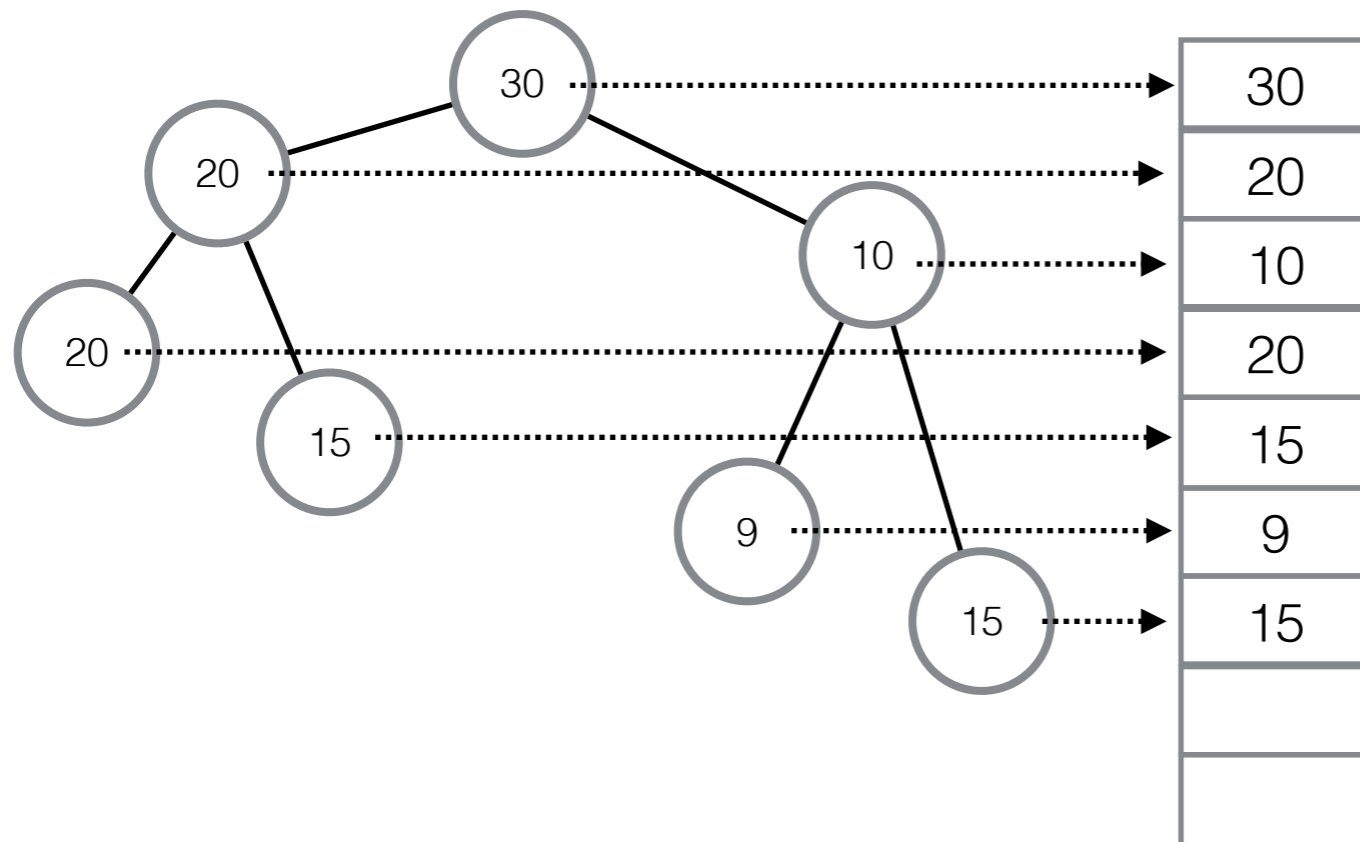
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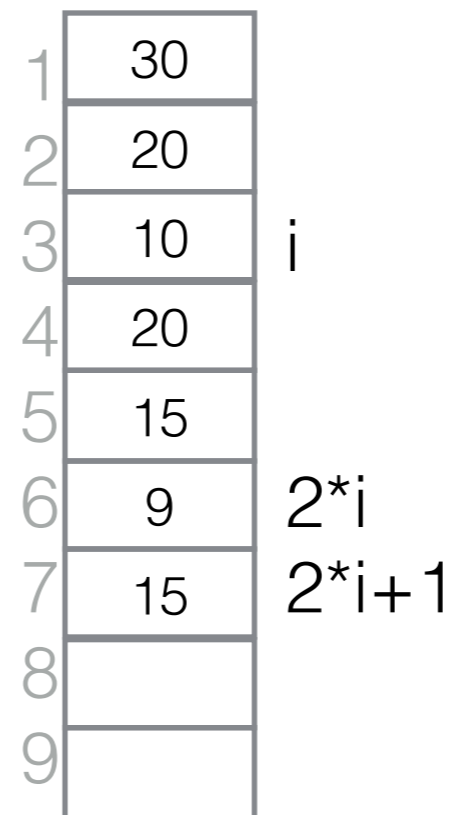
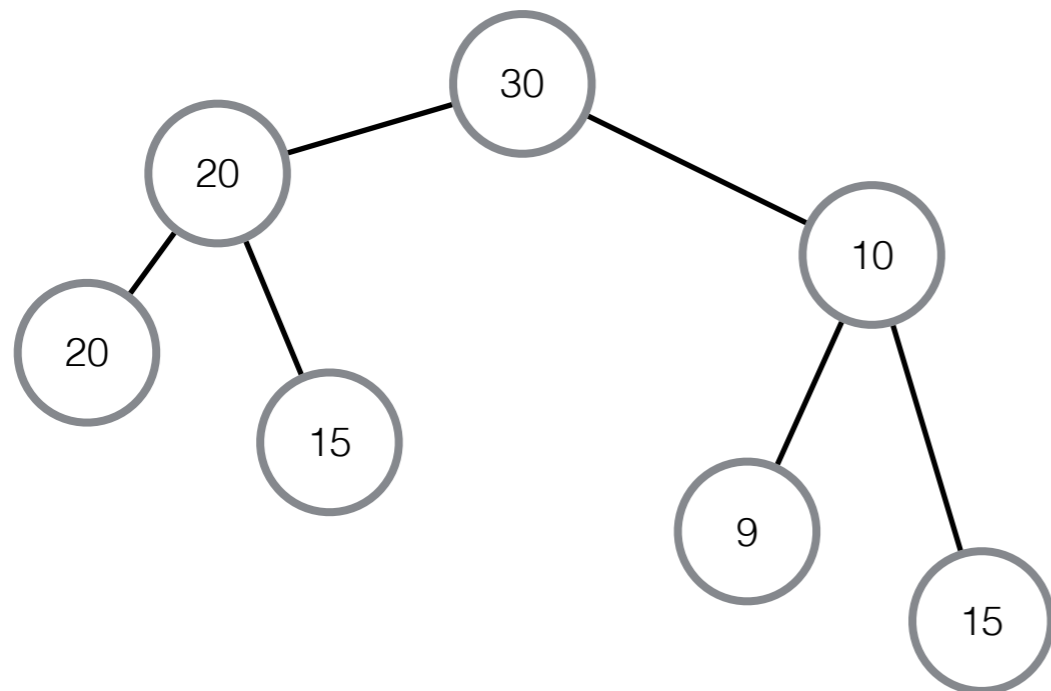
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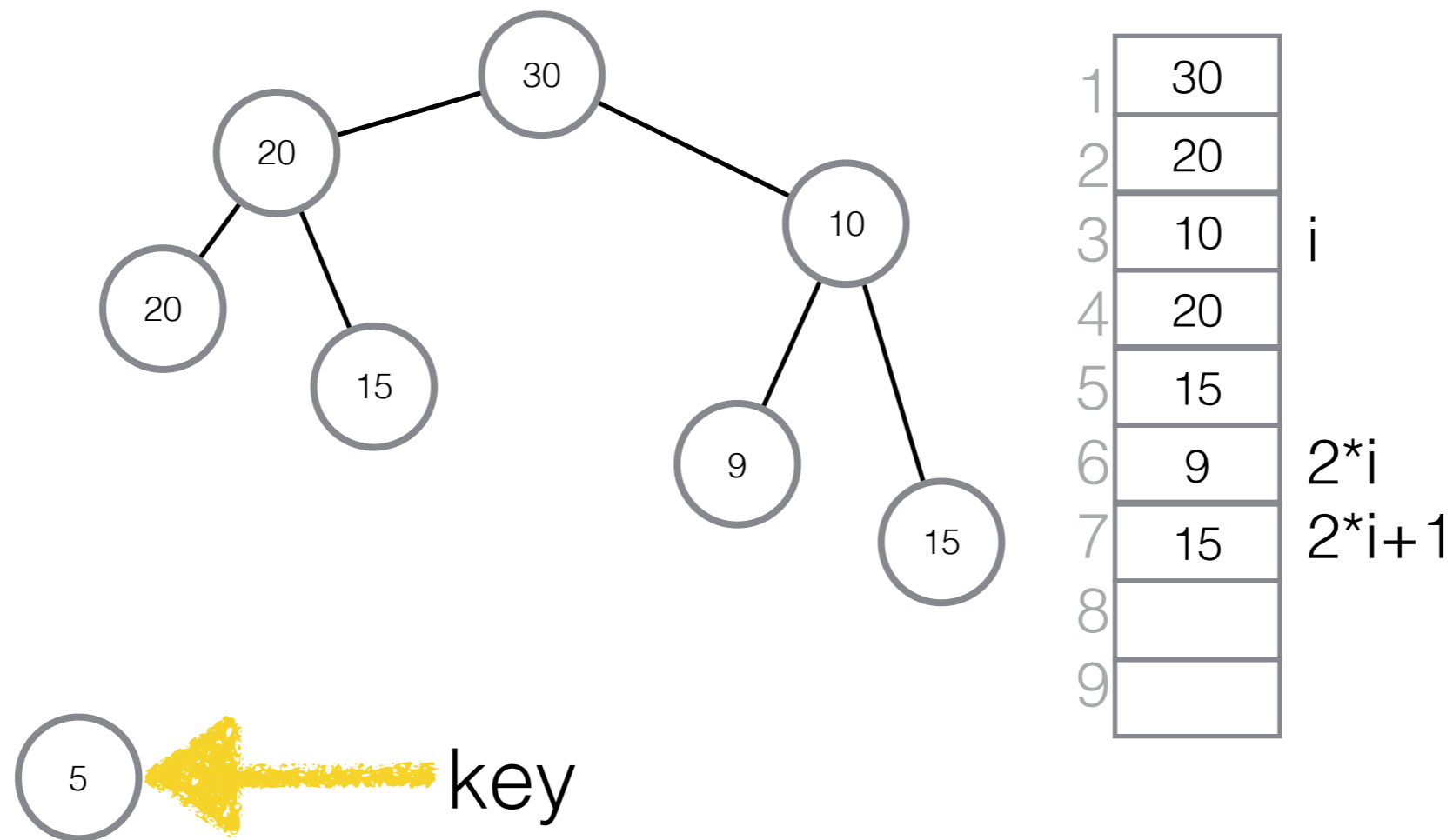


# Properties

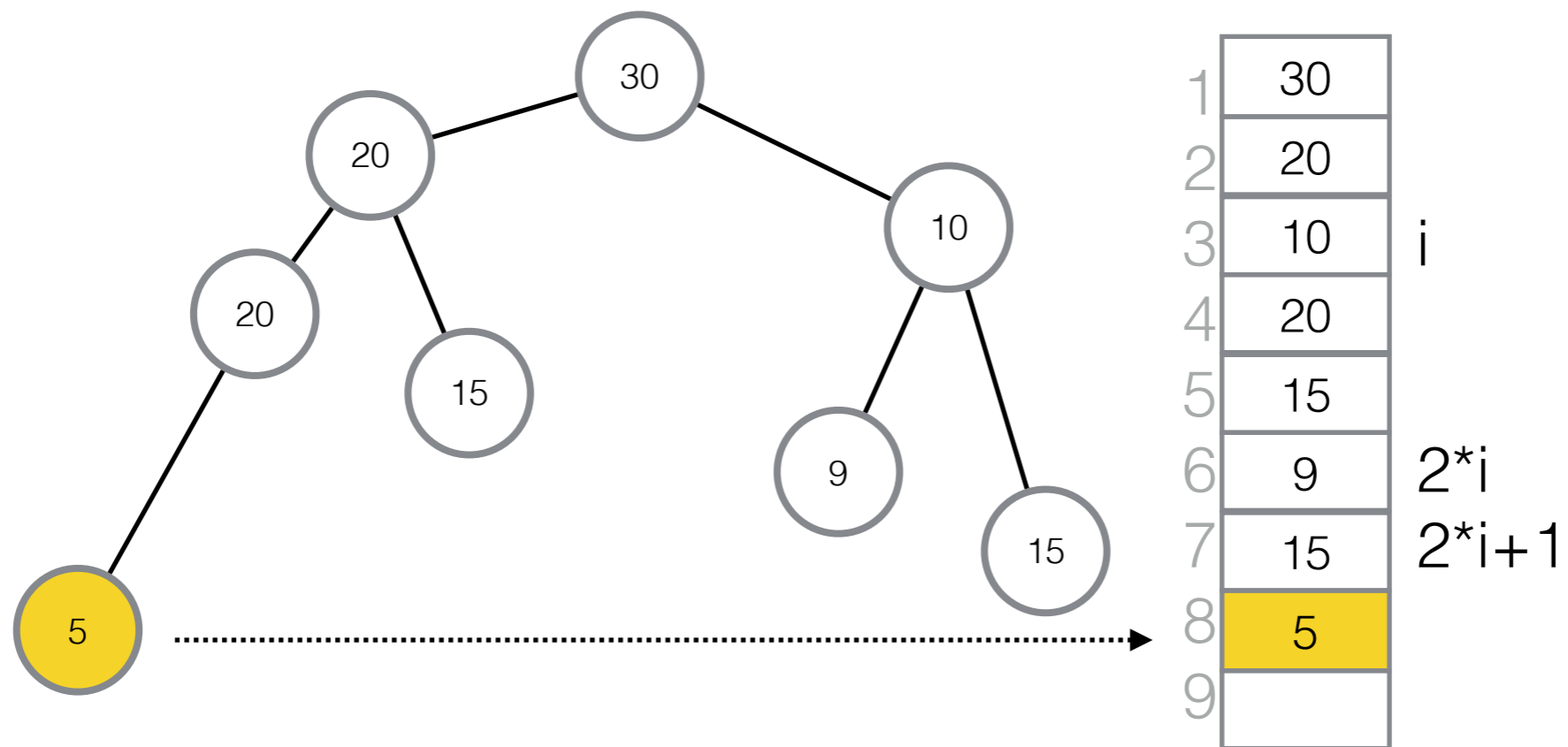
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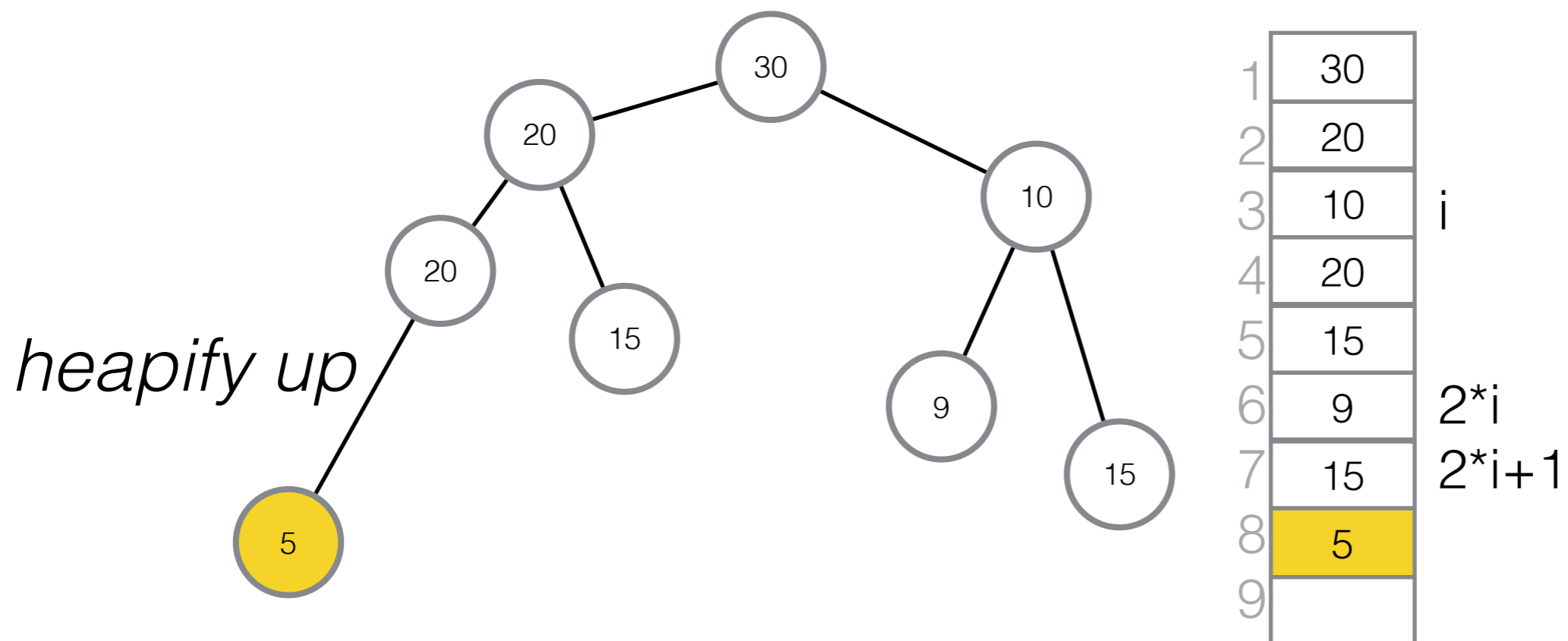
# insertion of a new key



# insertion of a new key

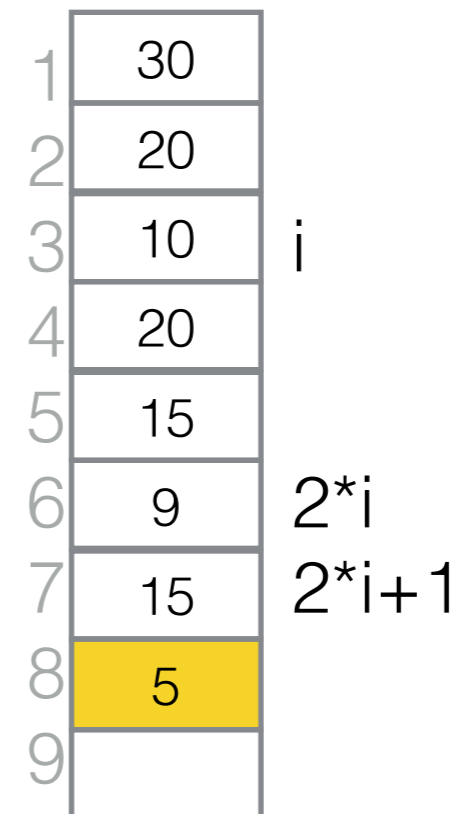
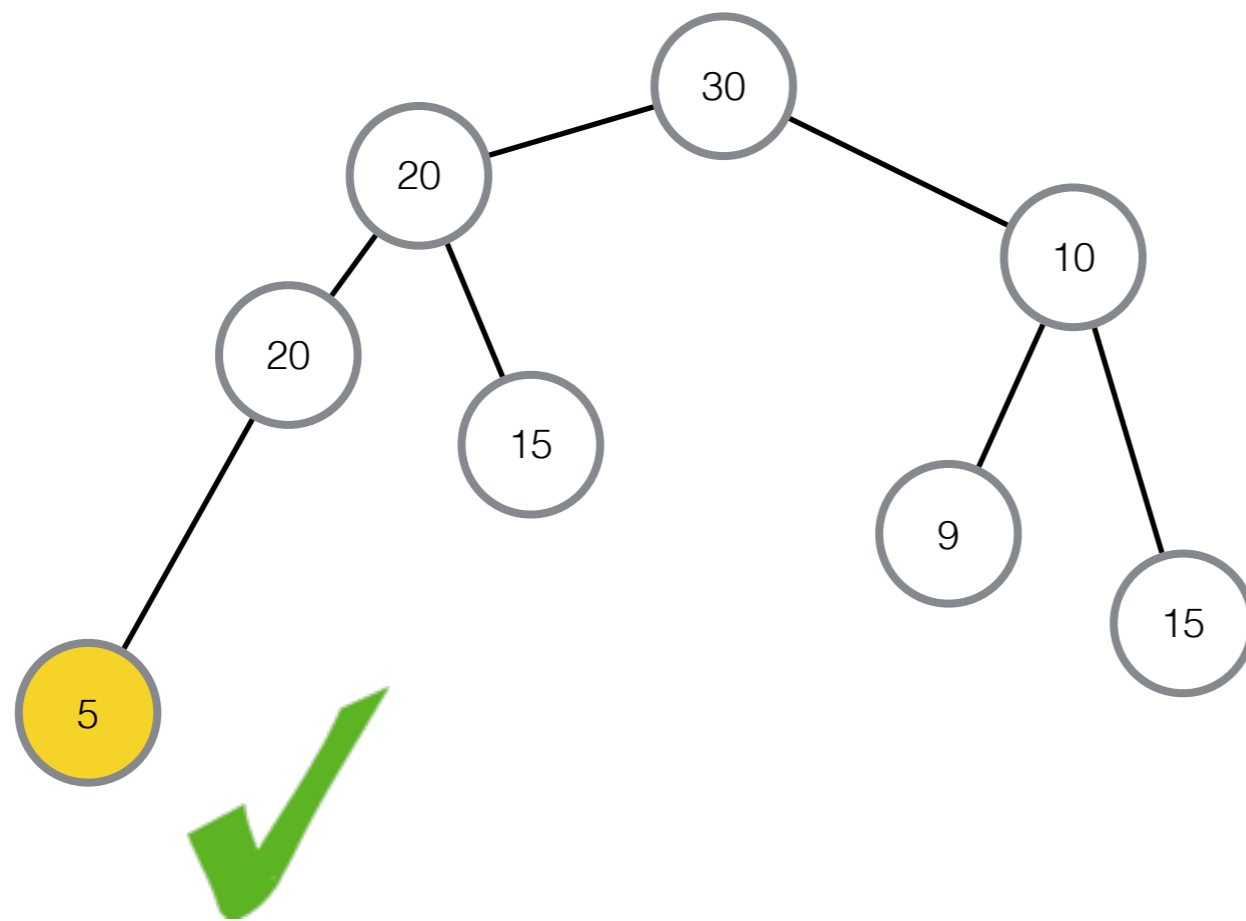


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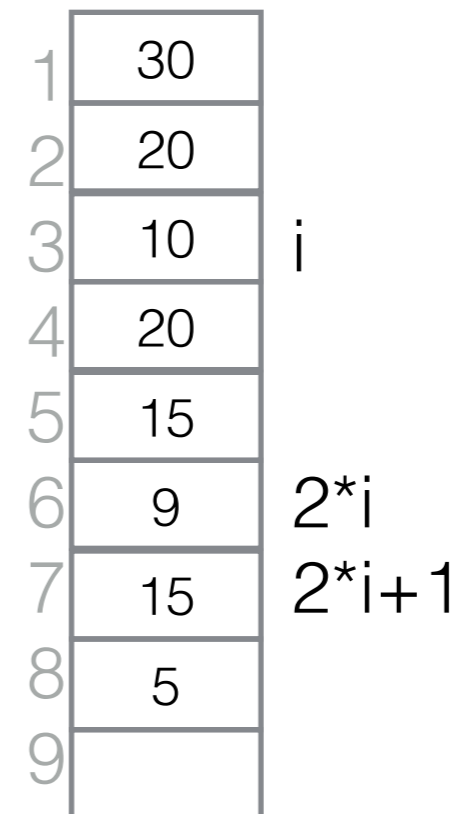
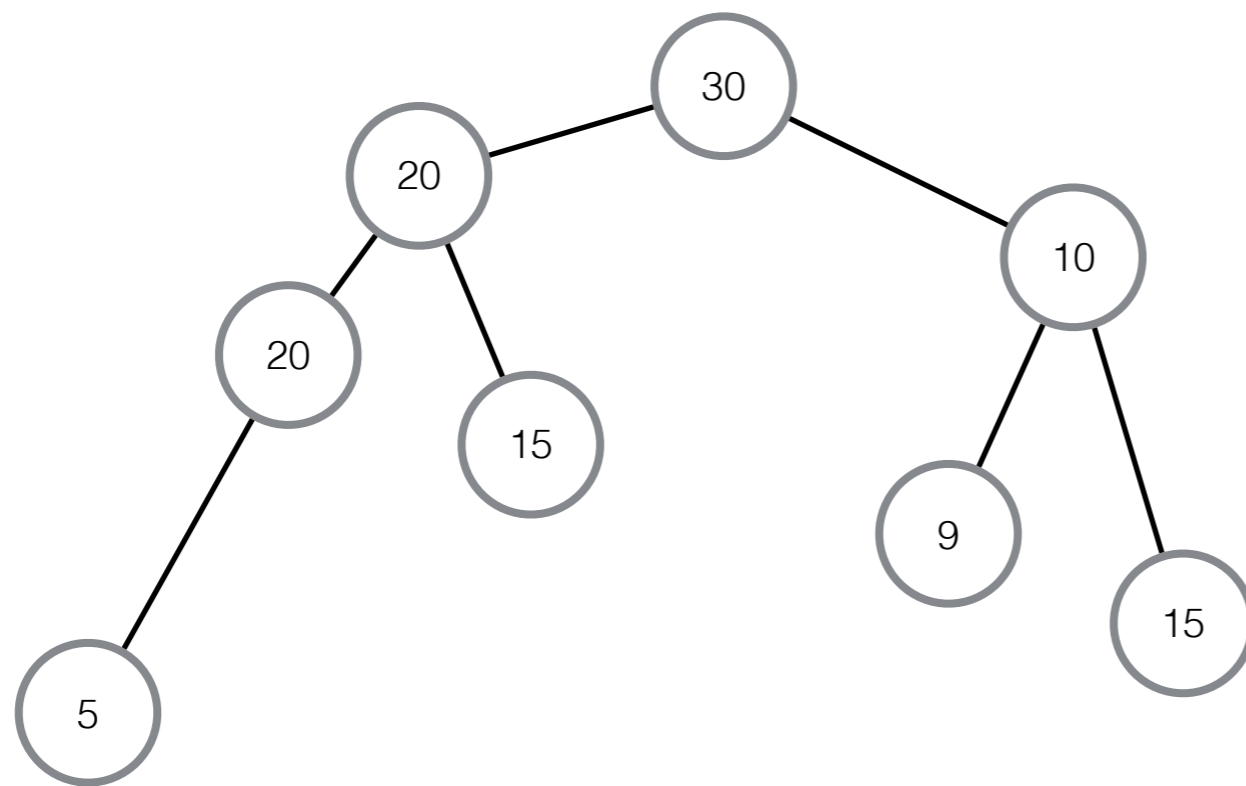




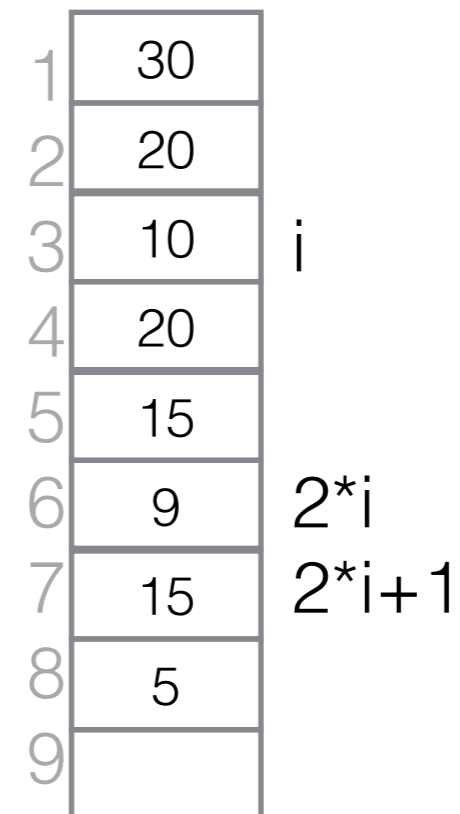
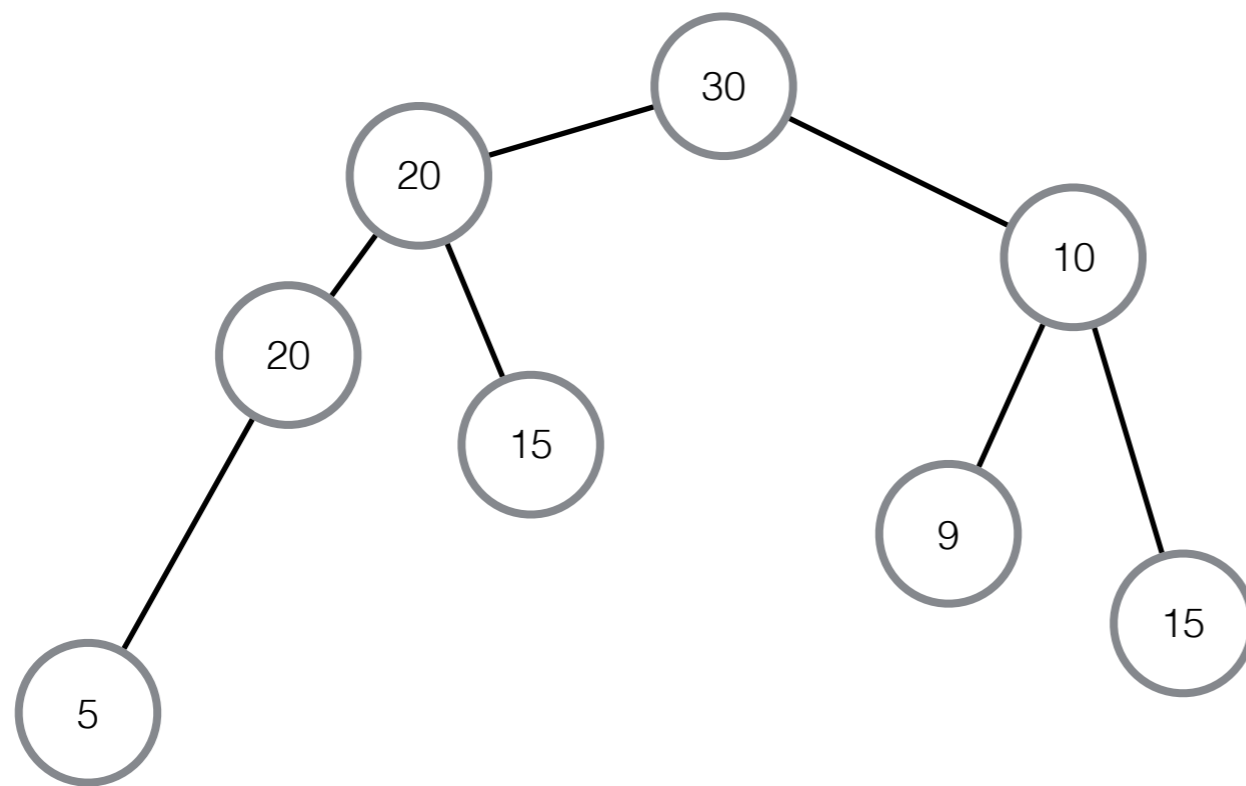
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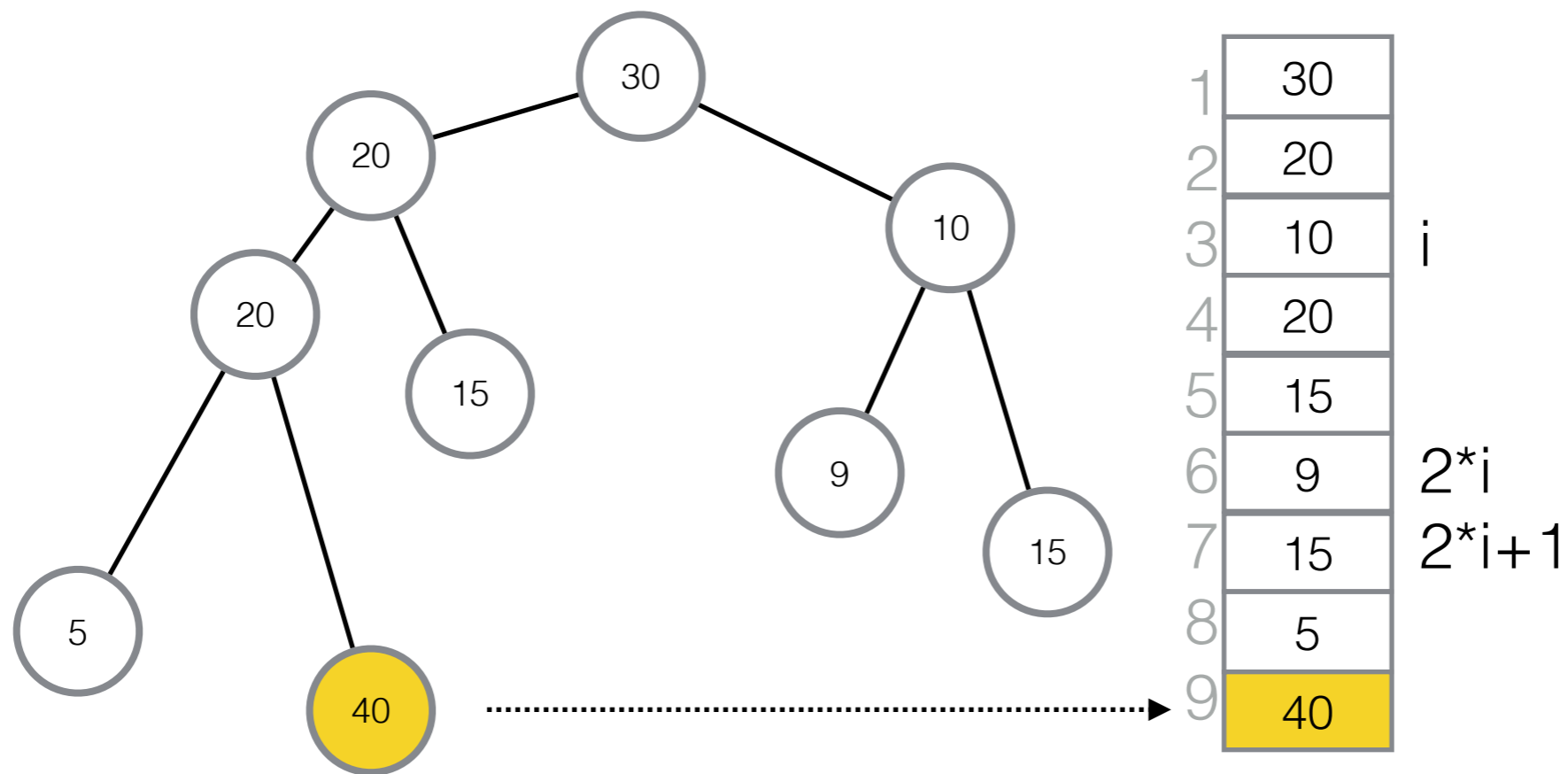
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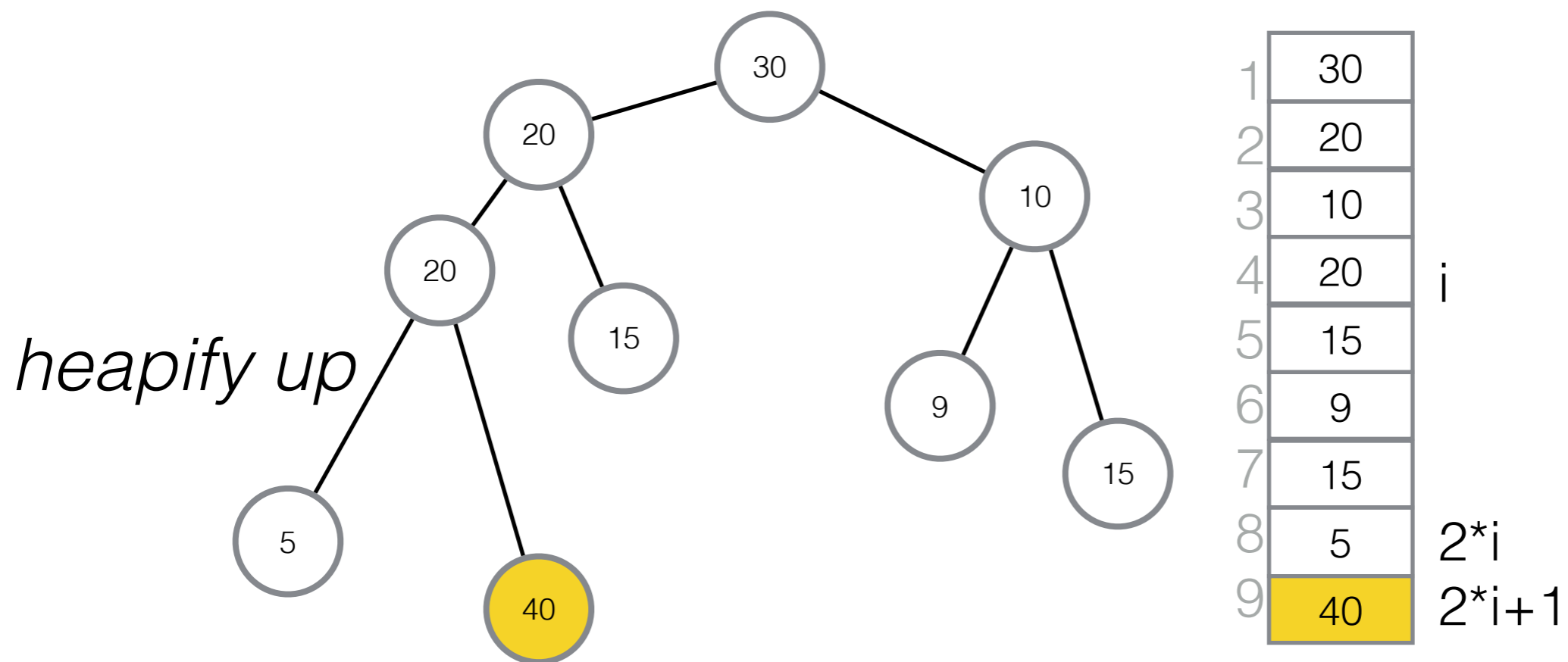
# insertion of a new key



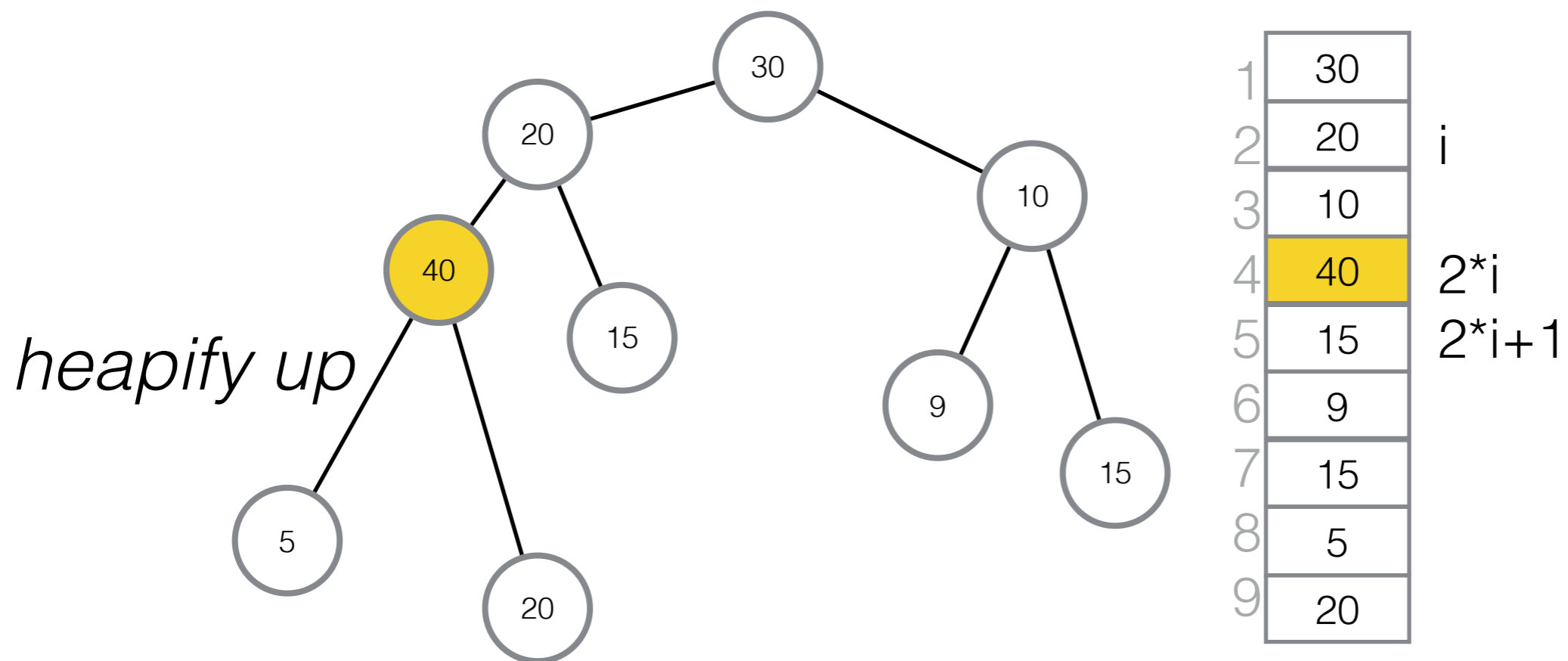
# insertion of a new key



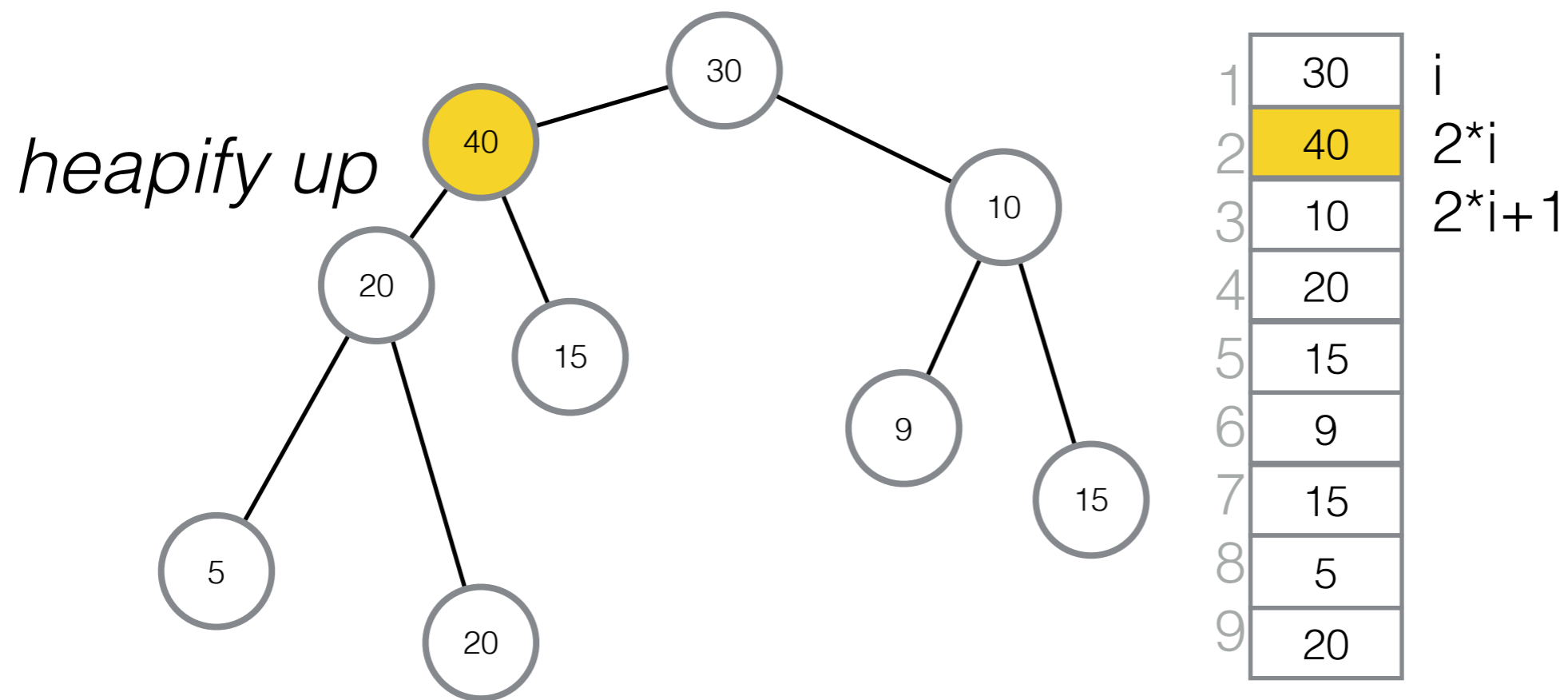
# insertion of a new key



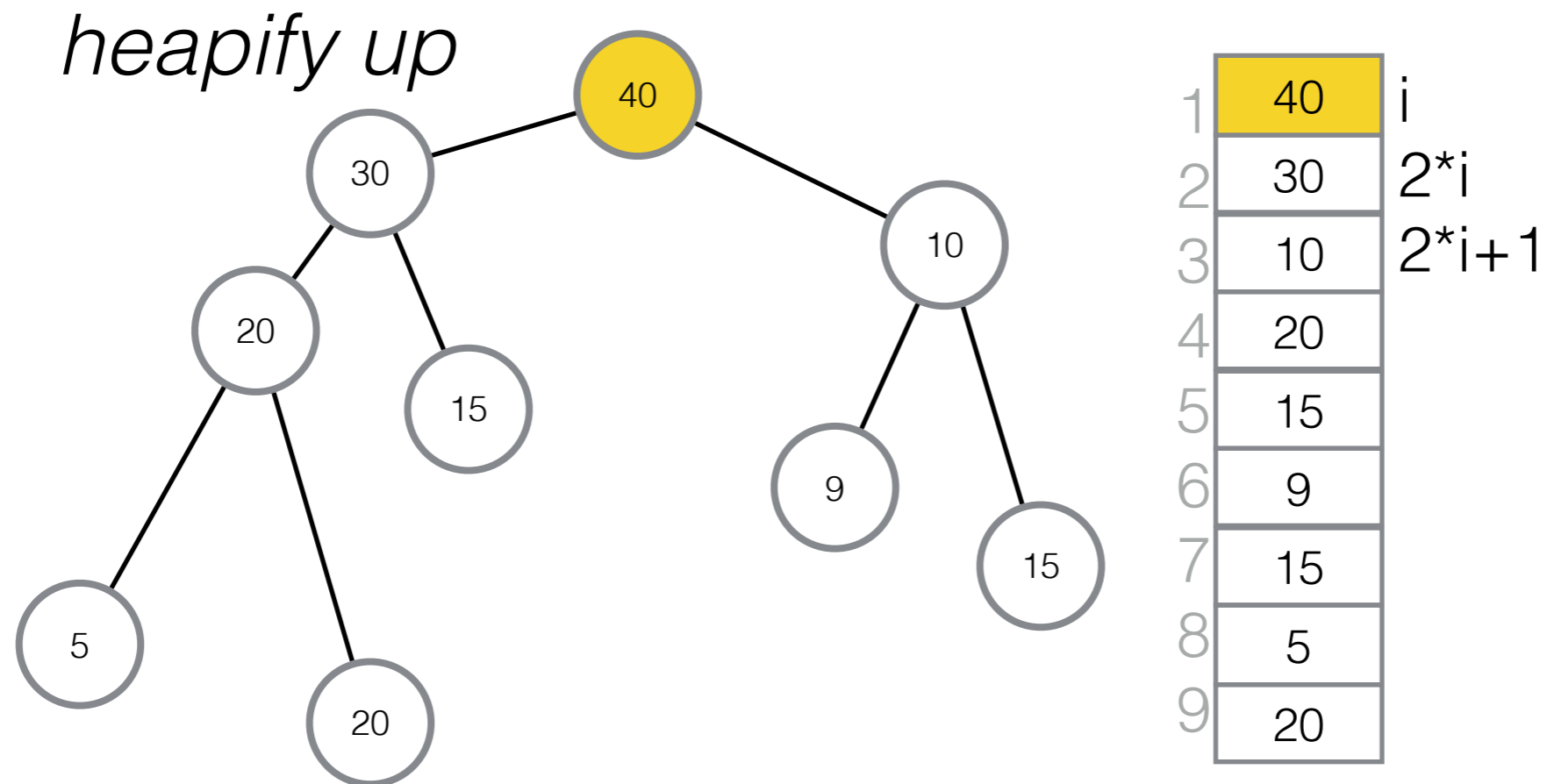
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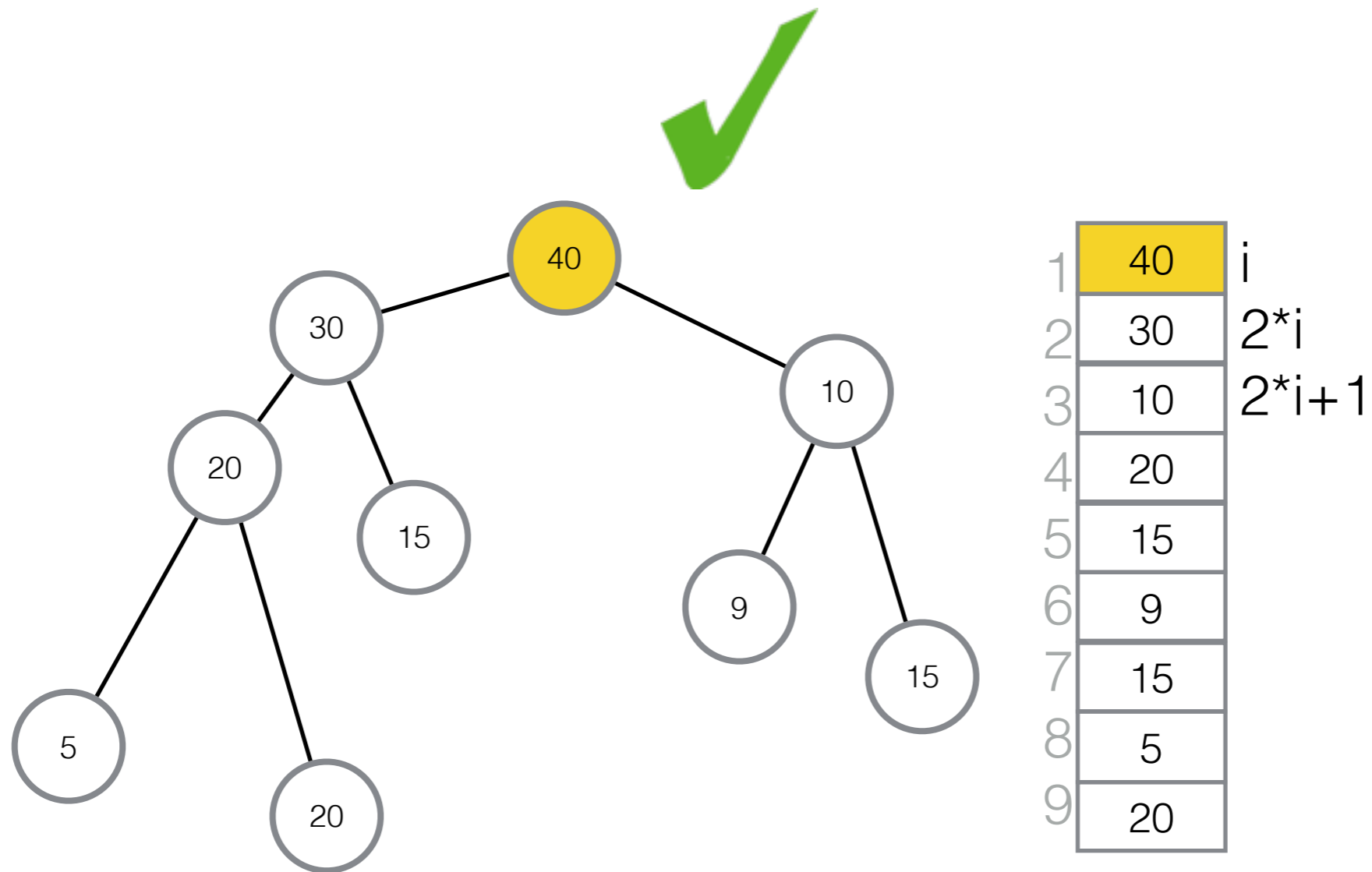


# insertion of a new key





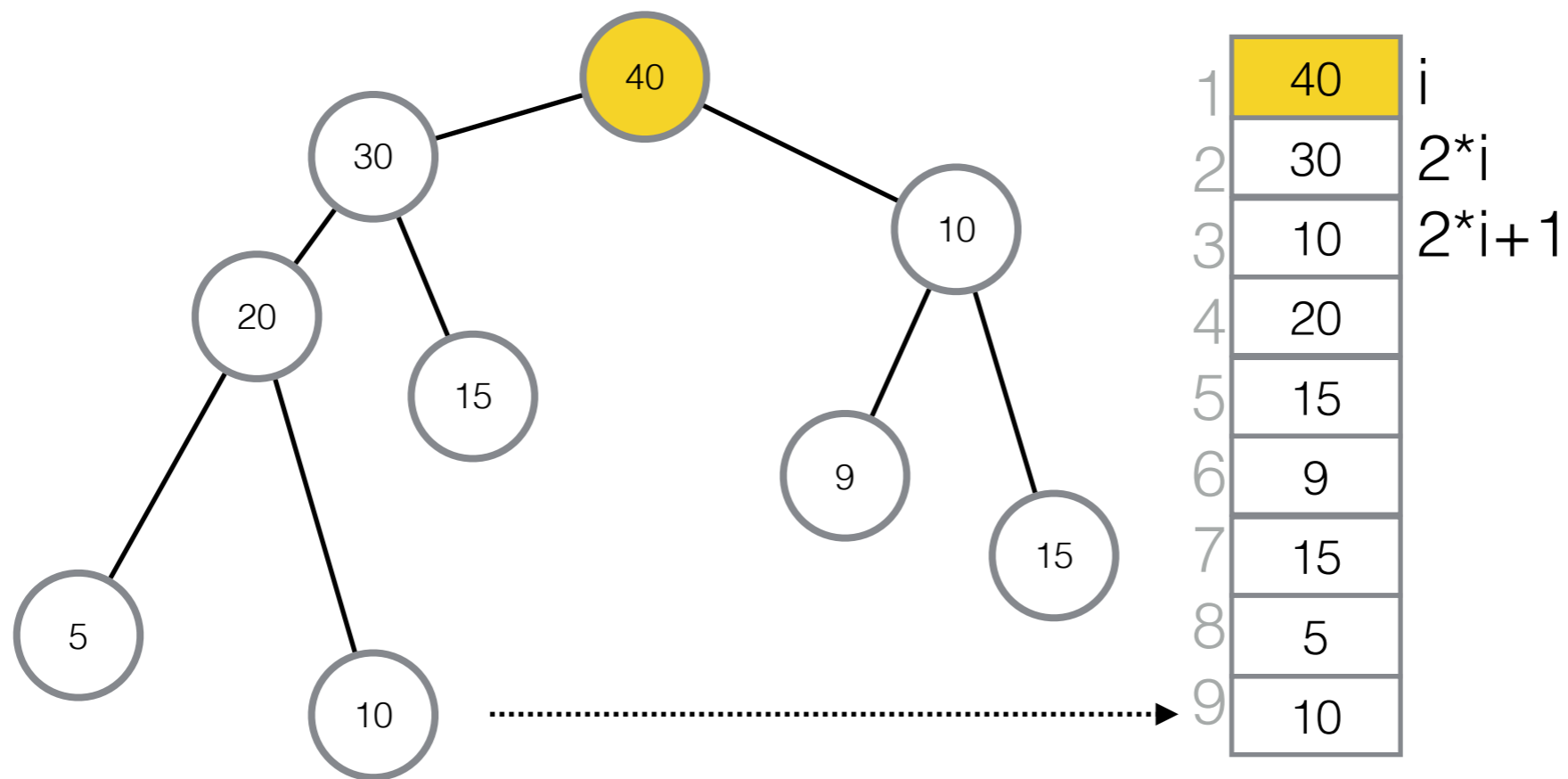
# insertion of a new key



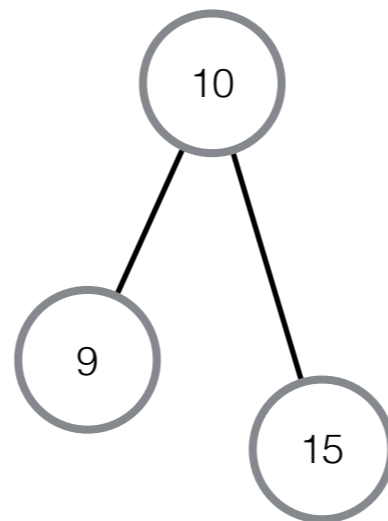
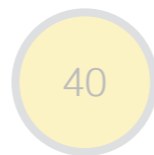
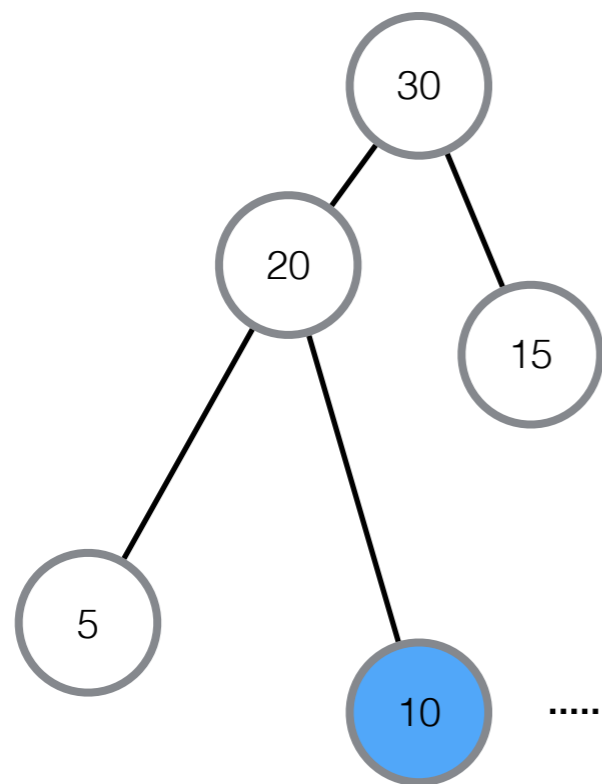
# Algorithm

```
heapifyUp( node )  
begin  
while ( node has a parent ) and ( node.key > parent.key )  
  begin  
    swap( node and its parent )  
    // node is now in its parent's original position  
  end  
end
```

# Deleting the Root



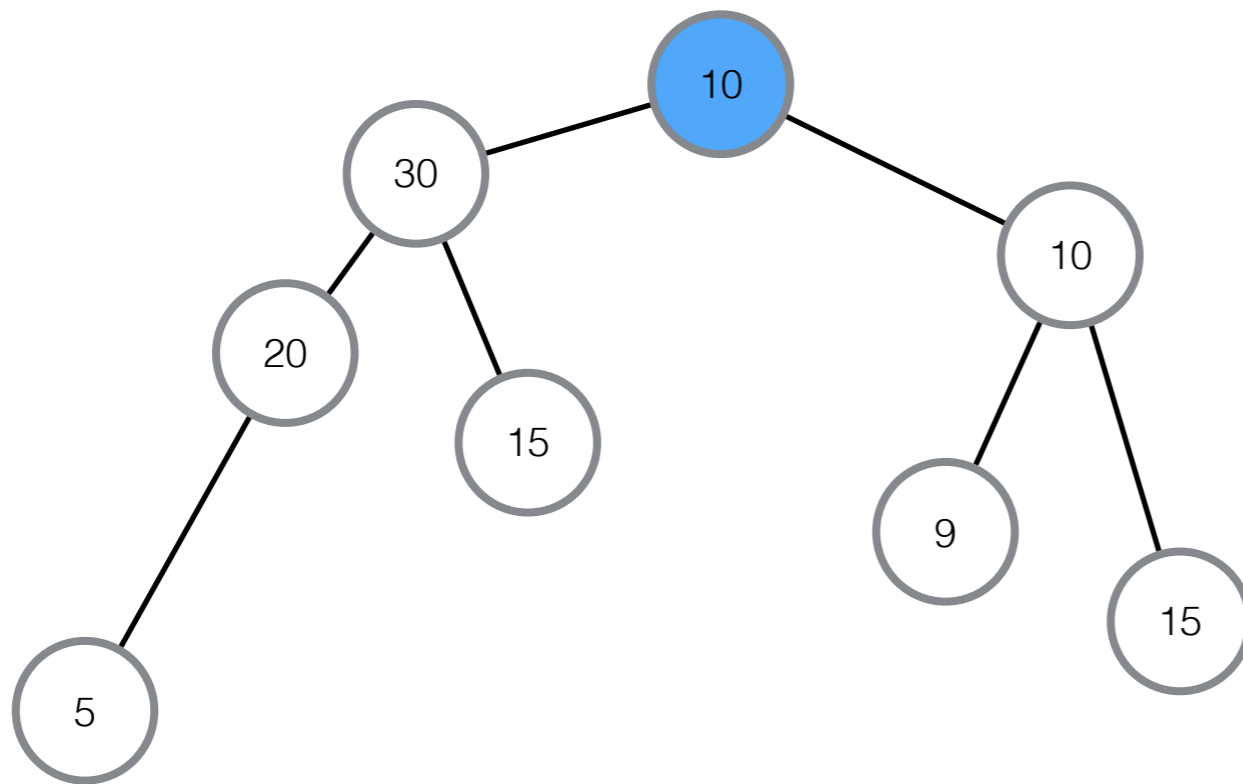
# Deleting the Root



1	40	$i$
2	30	$2*i$
3	10	$2*i+1$
4	20	
5	15	
6	9	
7	15	
8	5	
9	10	



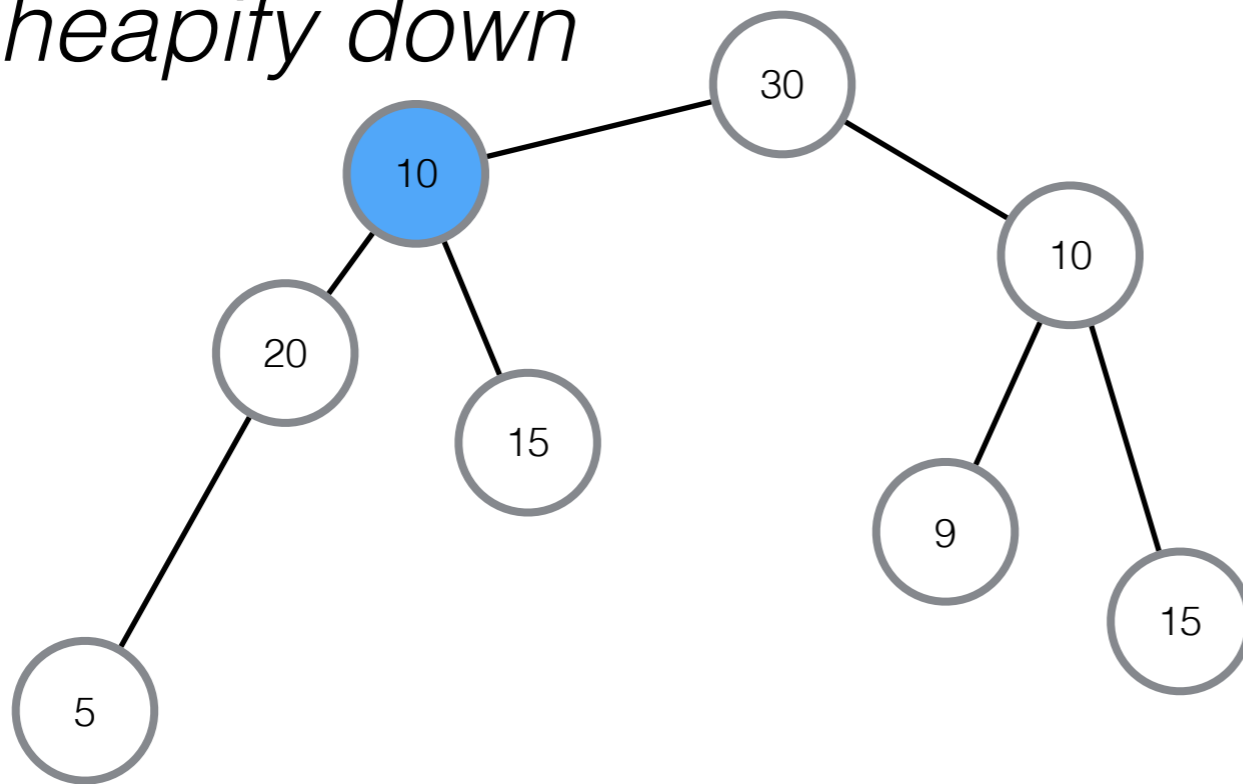
# Deleting the Root



1	10	$i$
2	30	$2*i$
3	10	$2*i+1$
4	20	
5	15	
6	9	
7	15	
8	5	
9	XX	

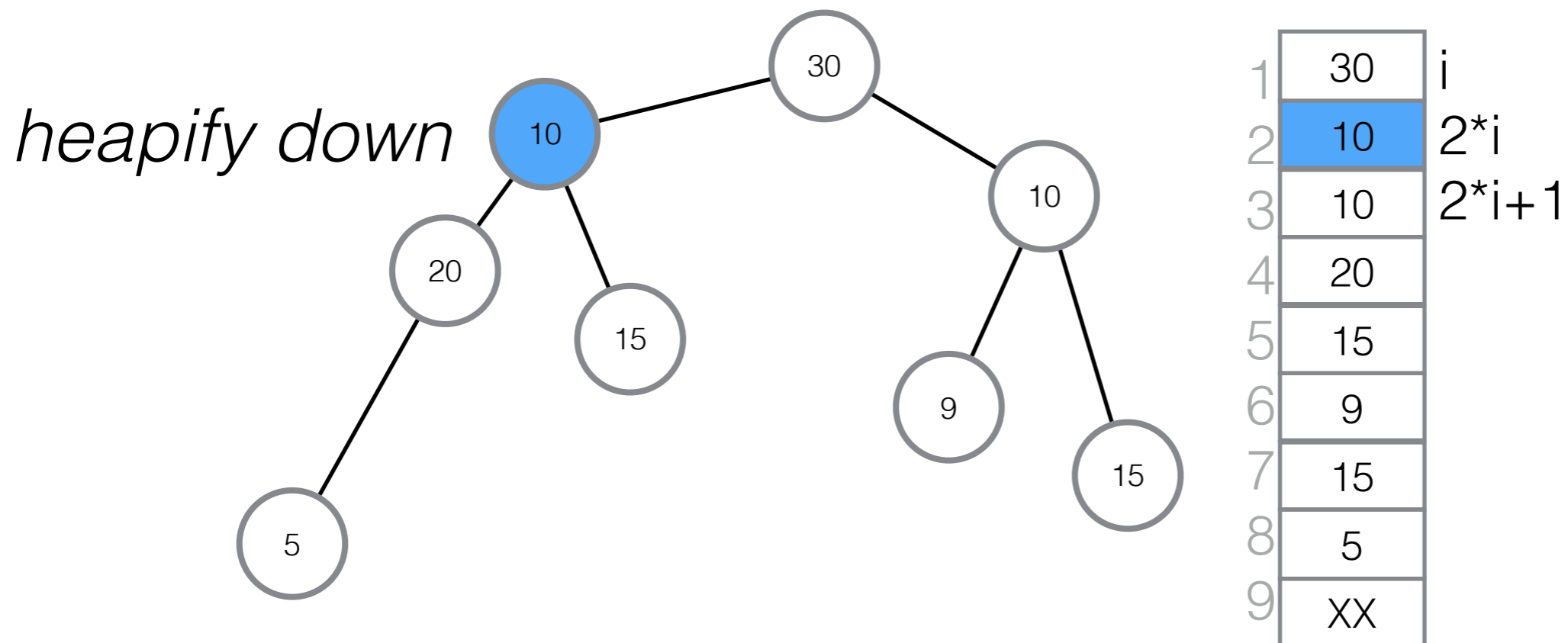
# Deleting the Root

*heapify down*

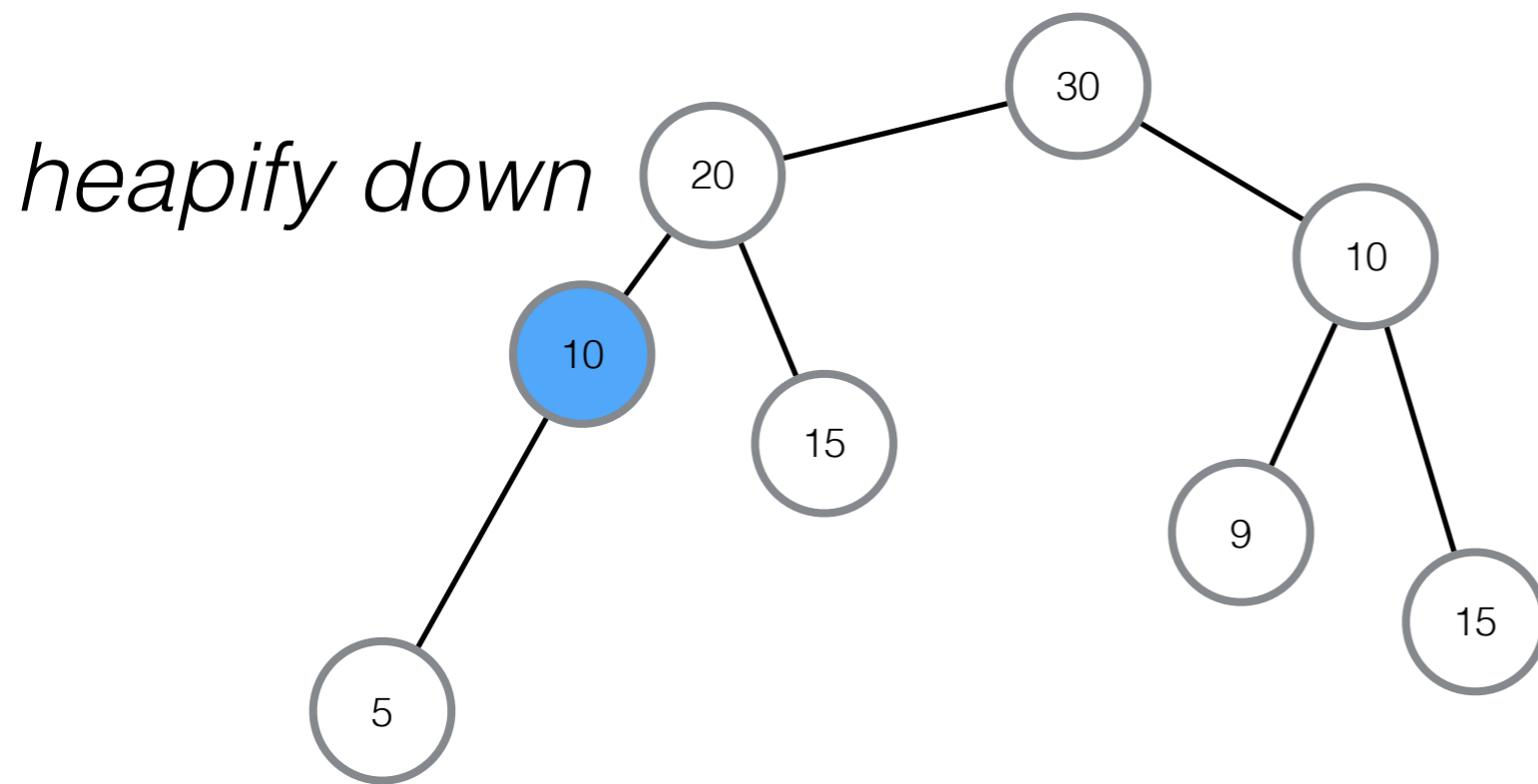


1	30	$i$
2	10	$2*i$
3	10	$2*i+1$
4	20	
5	15	
6	9	
7	15	
8	5	
9	XX	

# Deleting the Root



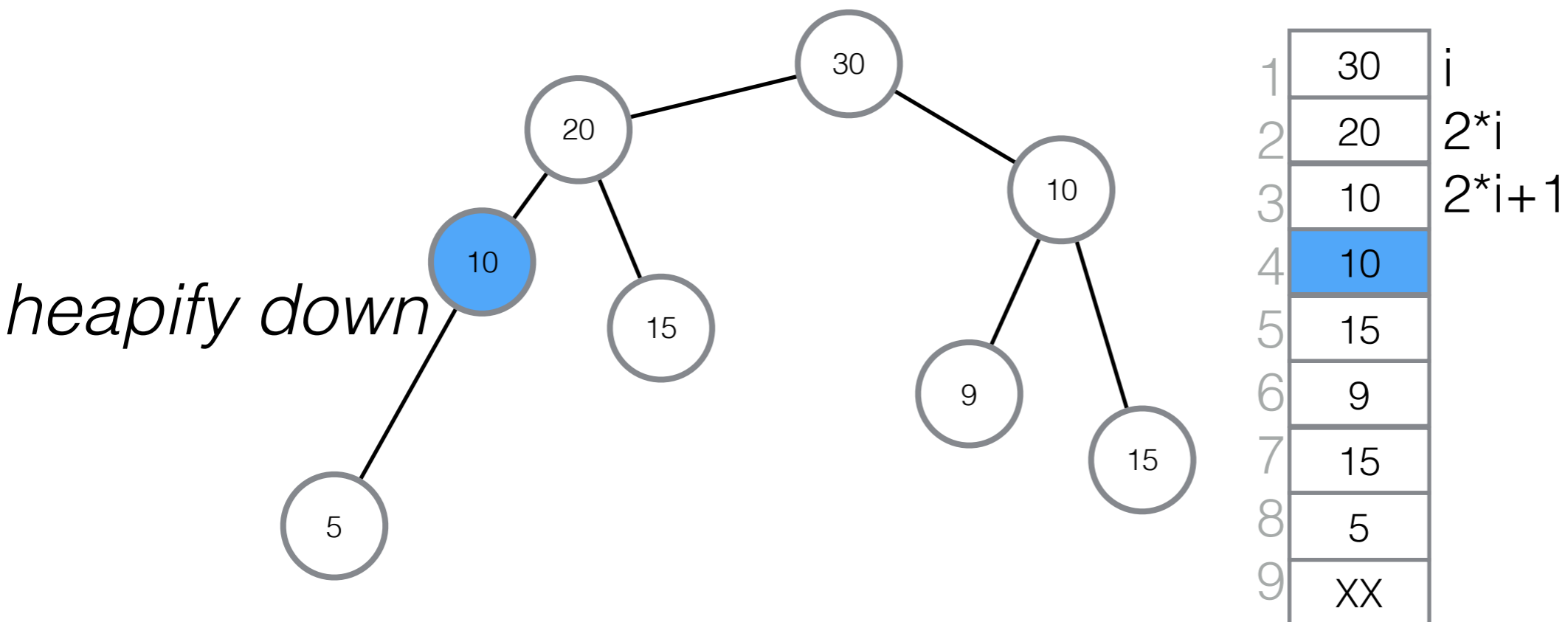
# Deleting the Root



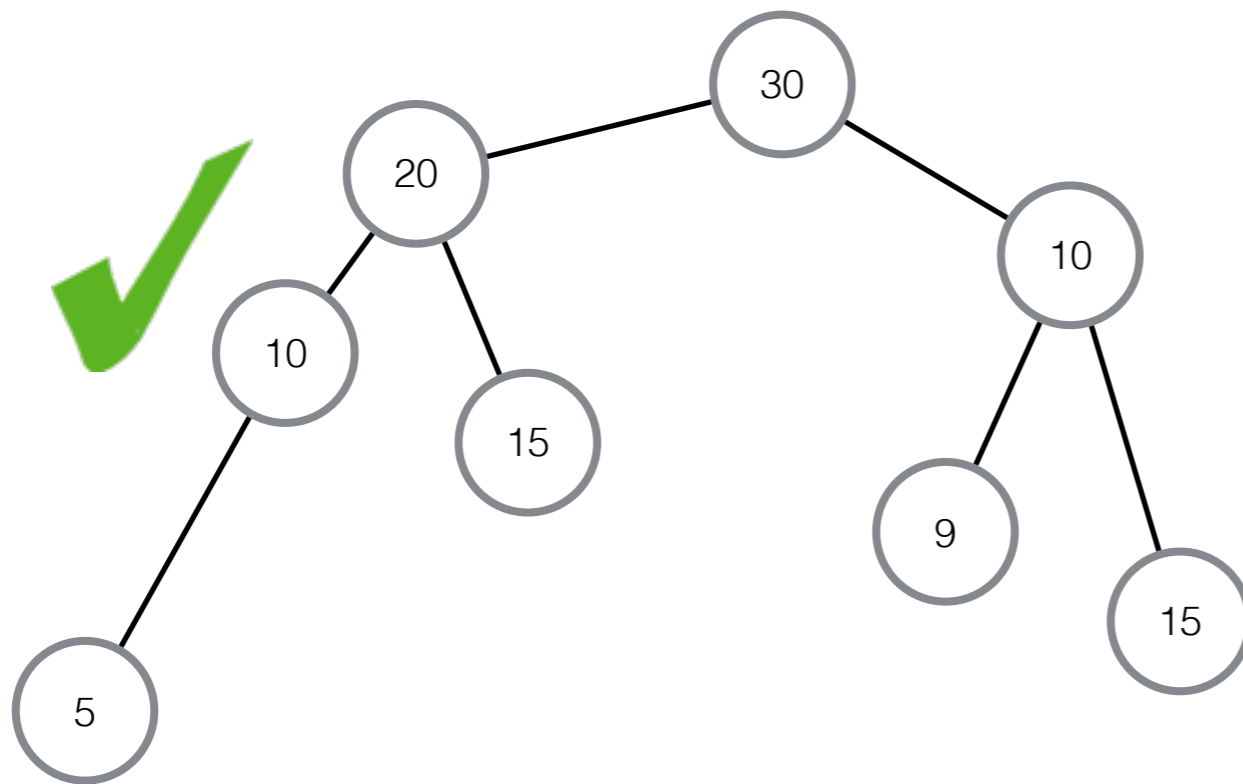
1	30	$i$
2	20	$2*i$
3	10	$2*i+1$
4	10	
5	15	
6	9	
7	15	
8	5	
9	XX	



# Deleting the Root

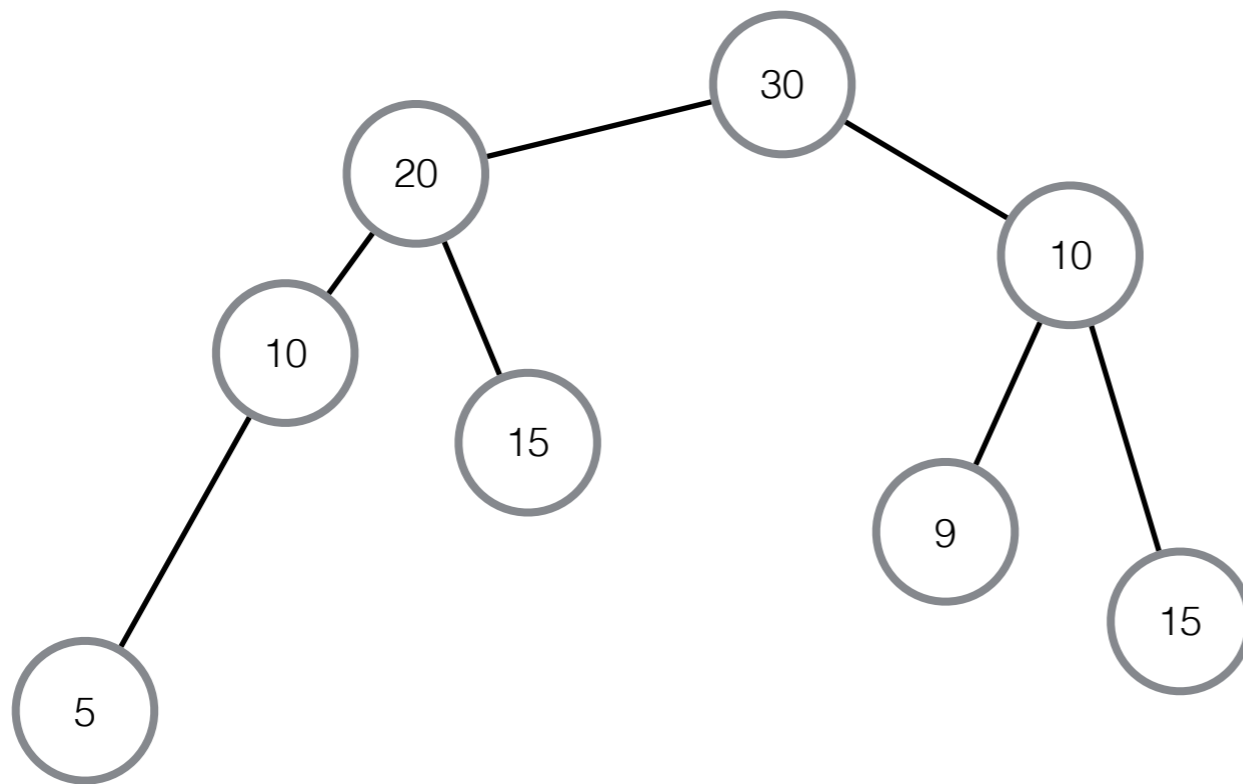


# Deleting the Root



1	30
2	20
3	10
4	10
5	15
6	9
7	15
8	5
9	XX

# Deleting the Root



1	30
2	20
3	10
4	10
5	15
6	9
7	15
8	5
9	XX

# Algorithm

```
heapifyDown( node )  
  begin  
  if node is a leaf then  
    return  
  maxNode = node's child with highest key  
  if maxNode.key > node.key then  
    begin  
    swap( node and maxNode )  
    // now node is in place of maxNode's original position  
    heapifyDown( node ) // recurse down  
    end  
  end
```

Operation	Worst Time Complexity	Average Time Complexity	Best Time Complexity
Insert key	$O(\log N)$	$O(\log N)$	$O(\log N)$
delete root	$O(\log N)$	$O(\log N)$	$O(\log N)$