

**CP312**

# **Algorithm Design and Analysis I**

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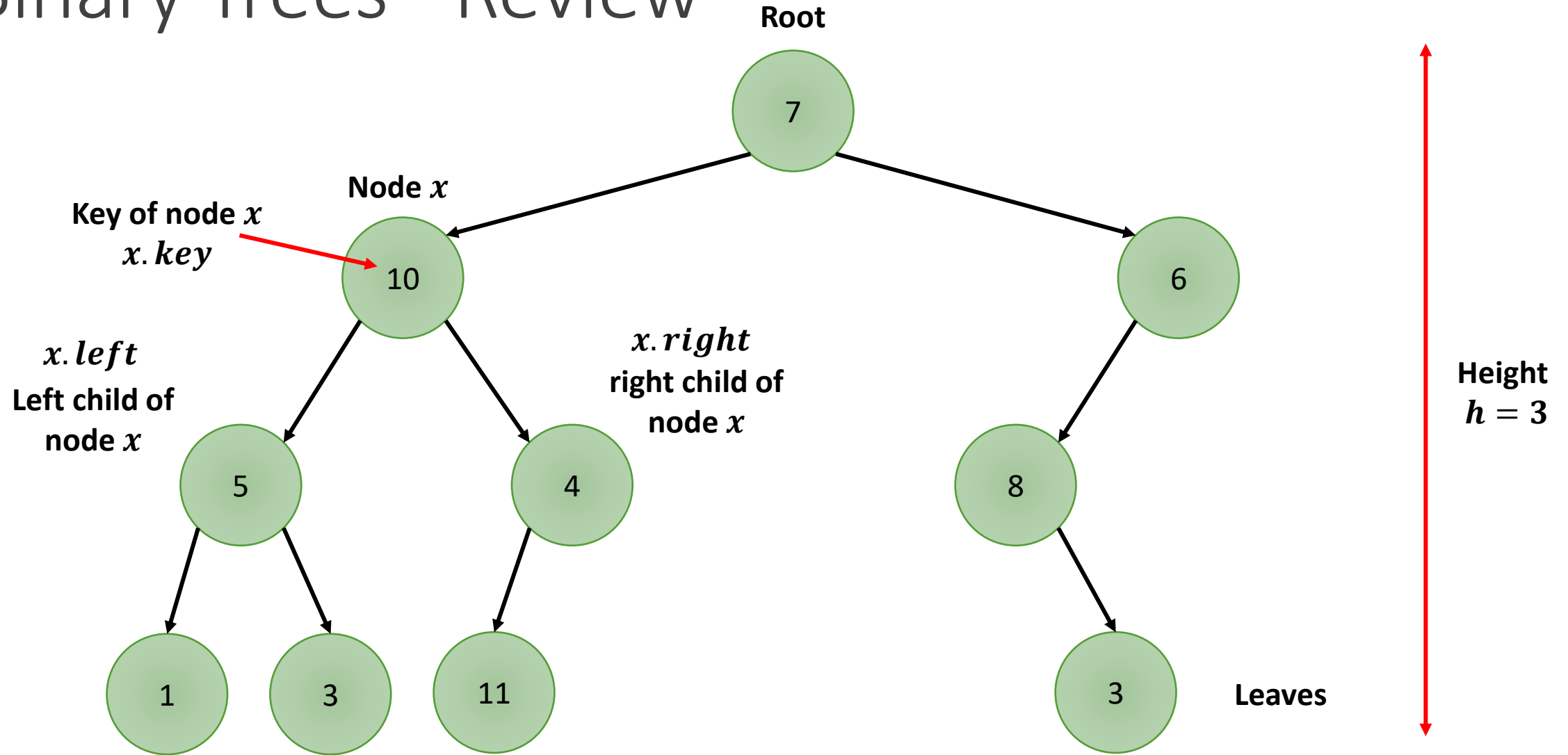
## **LECTURE 7: TREE-BASED SORTING**

# Sorting with Trees

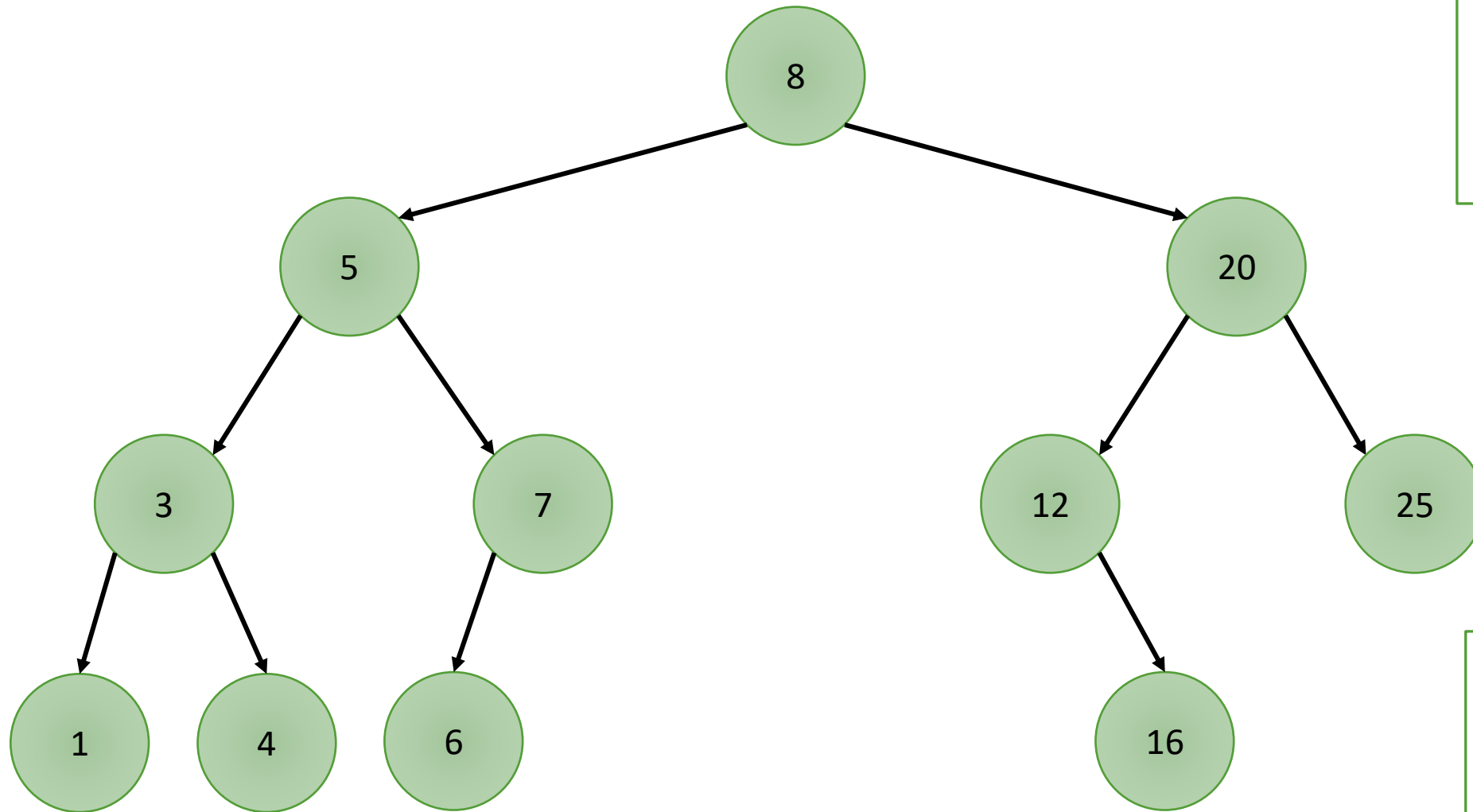
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- In previous lectures we saw how to sort using arrays.
- In this lecture, we will see the effect of changing data structures on the performance of an algorithm.
- In particular, we will use **binary trees** instead of arrays.

# Binary Trees - Review



# Binary Search Trees (BST)



## Key Property:

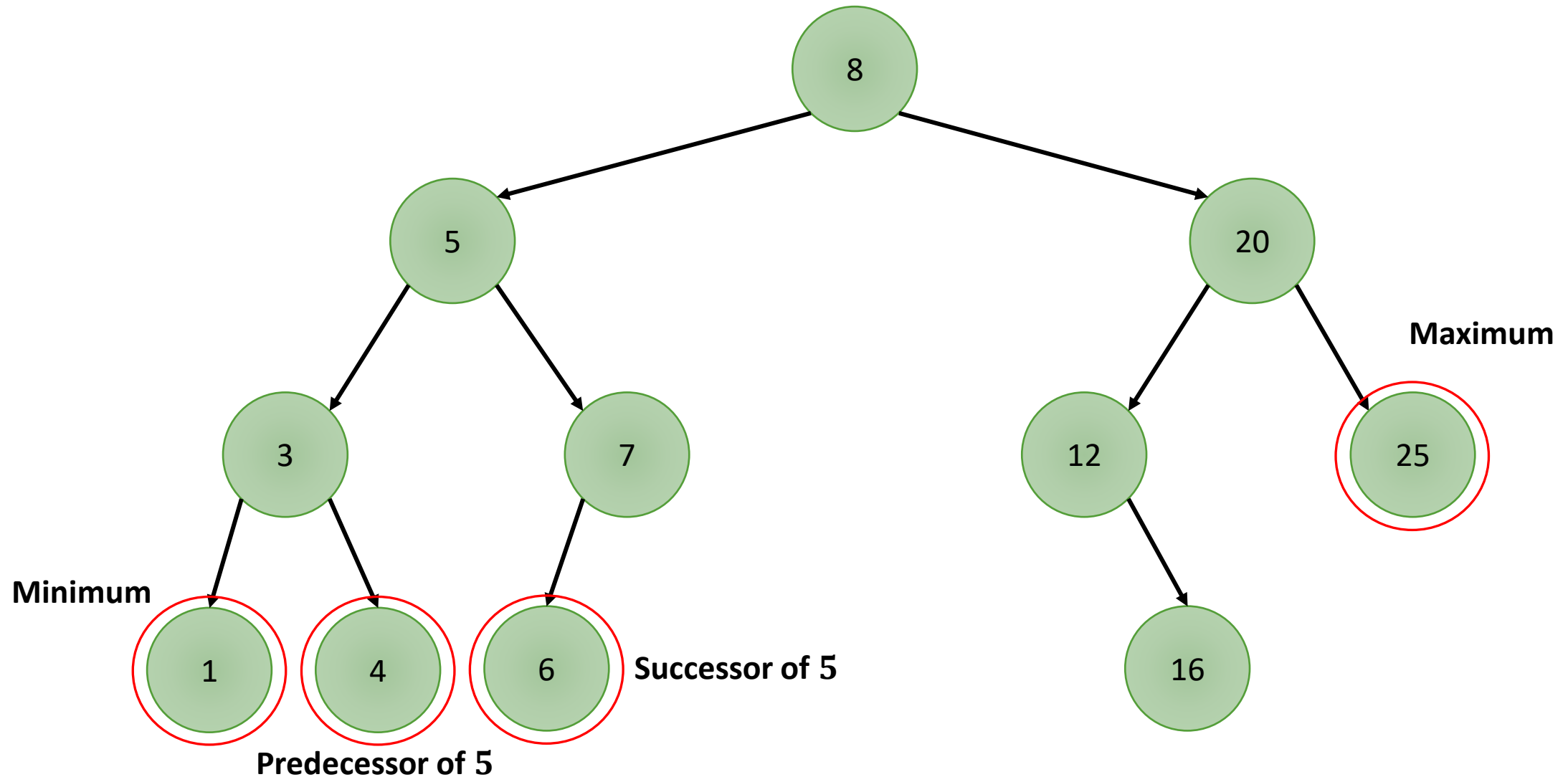
BST is a binary tree data structure where for every node  $x$  with left child  $y$  and right child  $z$ :

$$y.key < x.key < z.key$$

## BST operations:

- 1.) BST-search( $k$ )
- 2.) BST-insert( $k$ )
- 3.) BST-delete( $k$ )

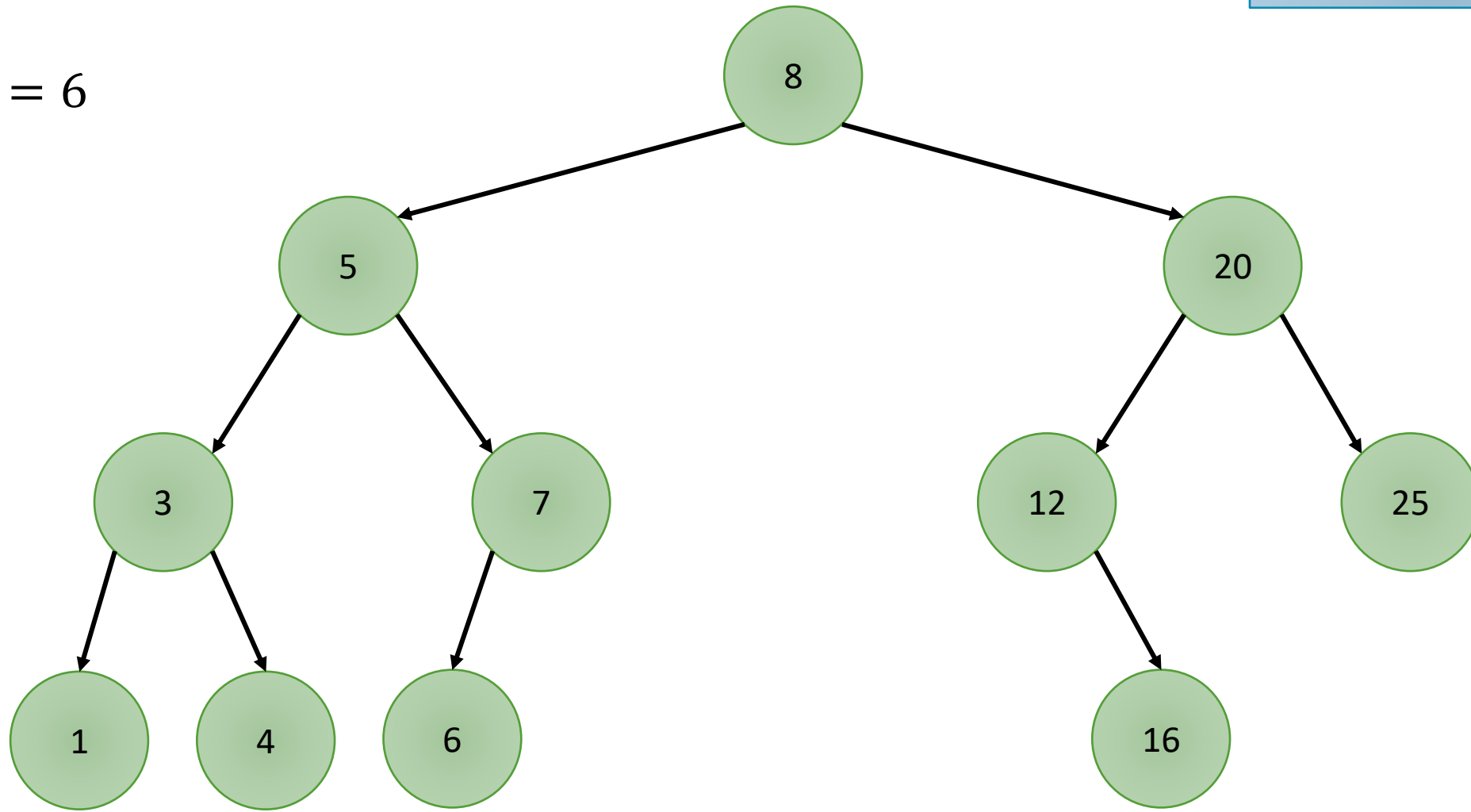
# Binary Search Trees (BST)



# Binary Search Trees (BST)

BST-Search( $k$ )

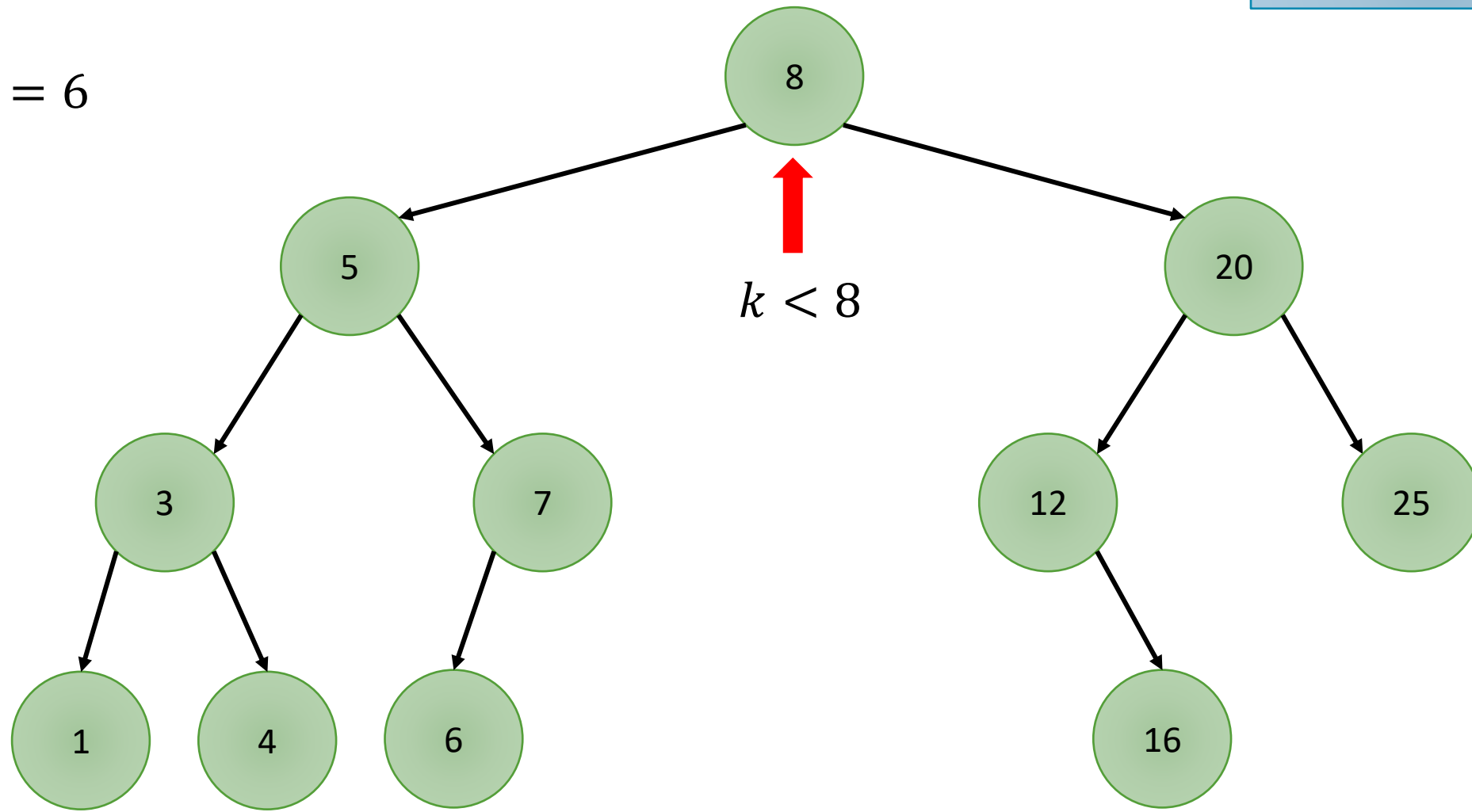
$k = 6$



# Binary Search Trees (BST)

BST-Search( $k$ )

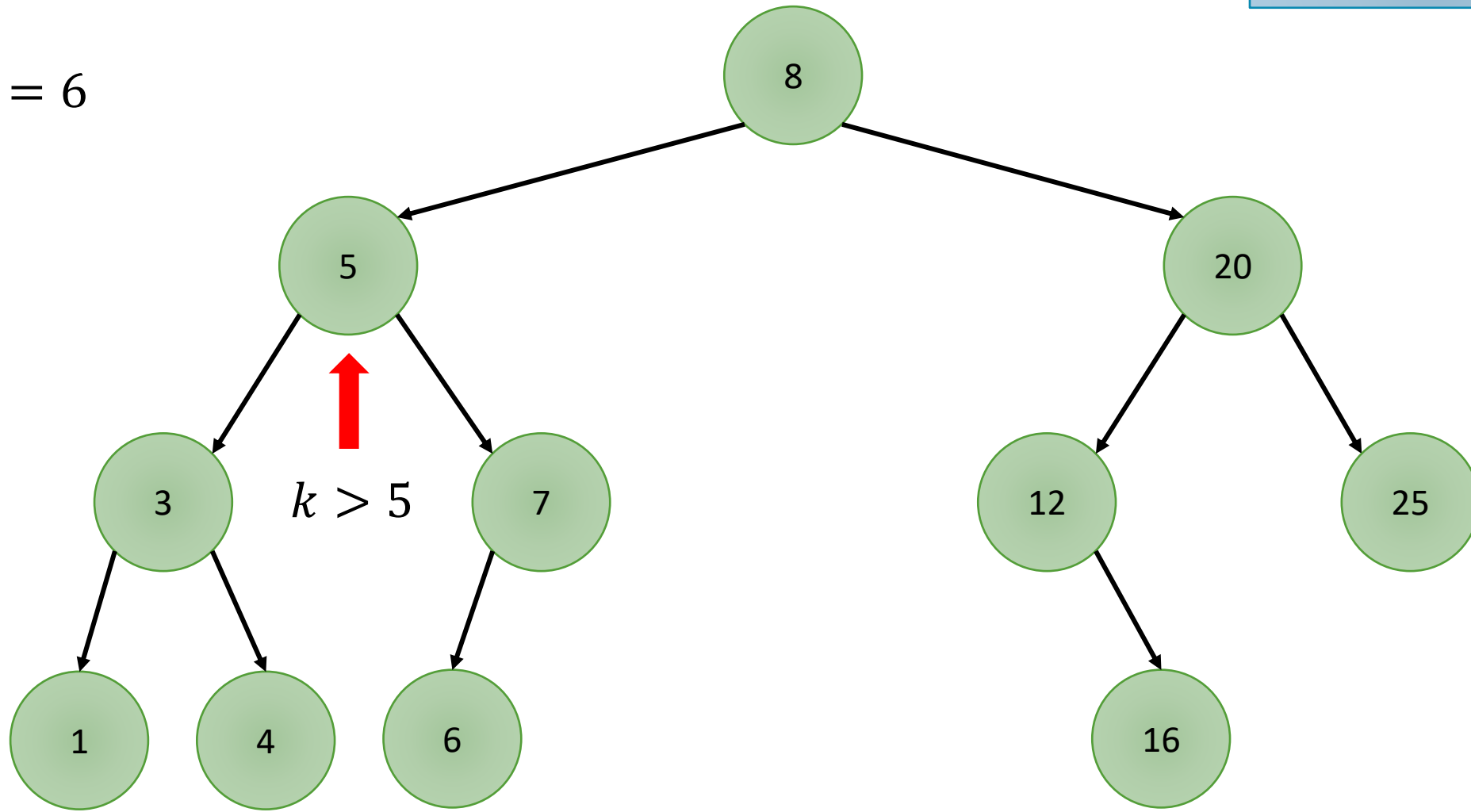
$k = 6$



# Binary Search Trees (BST)

BST-Search( $k$ )

$k = 6$

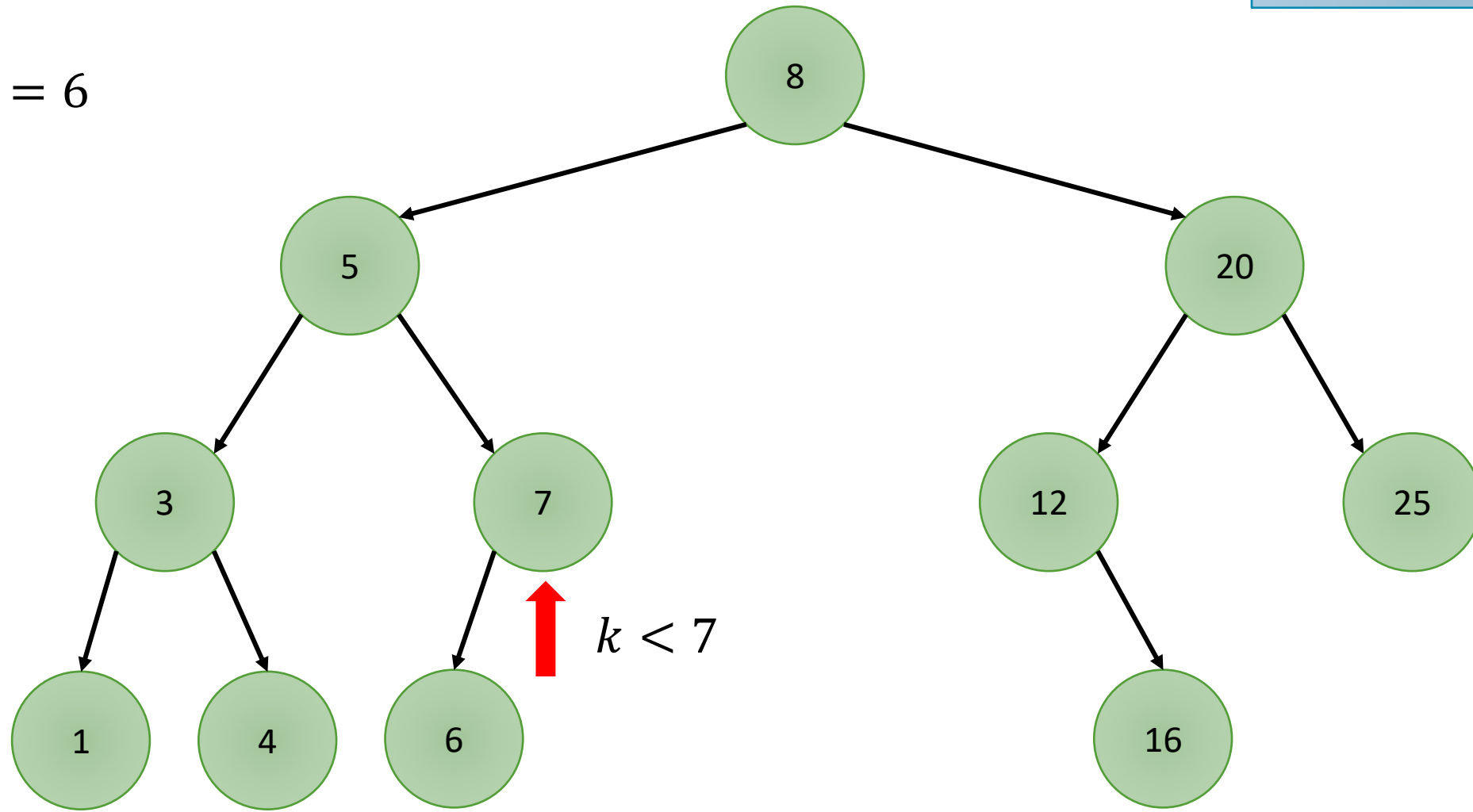




# Binary Search Trees (BST)

BST-Search( $k$ )

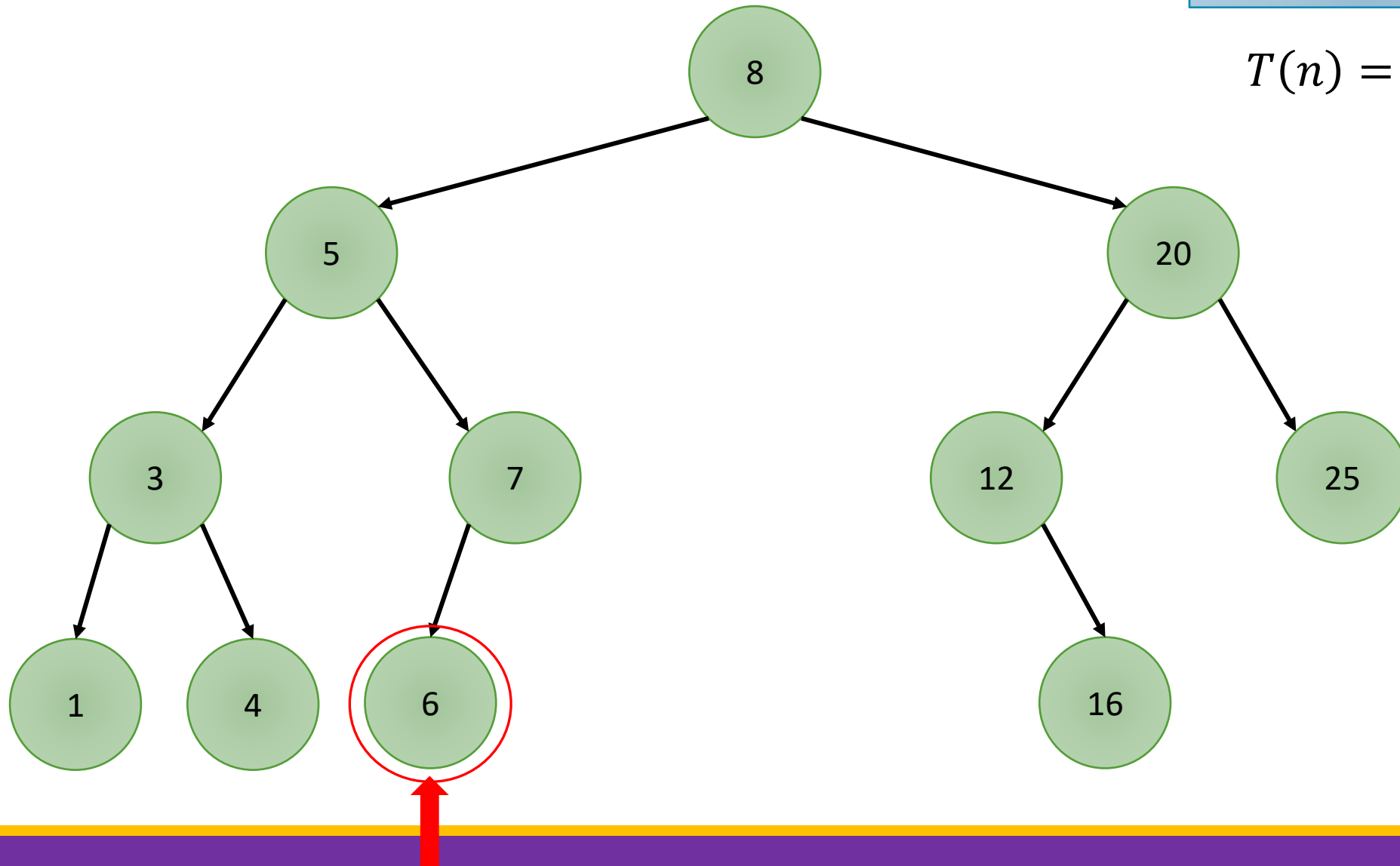
$k = 6$



# Binary Search Trees (BST)

BST-Search( $k$ )

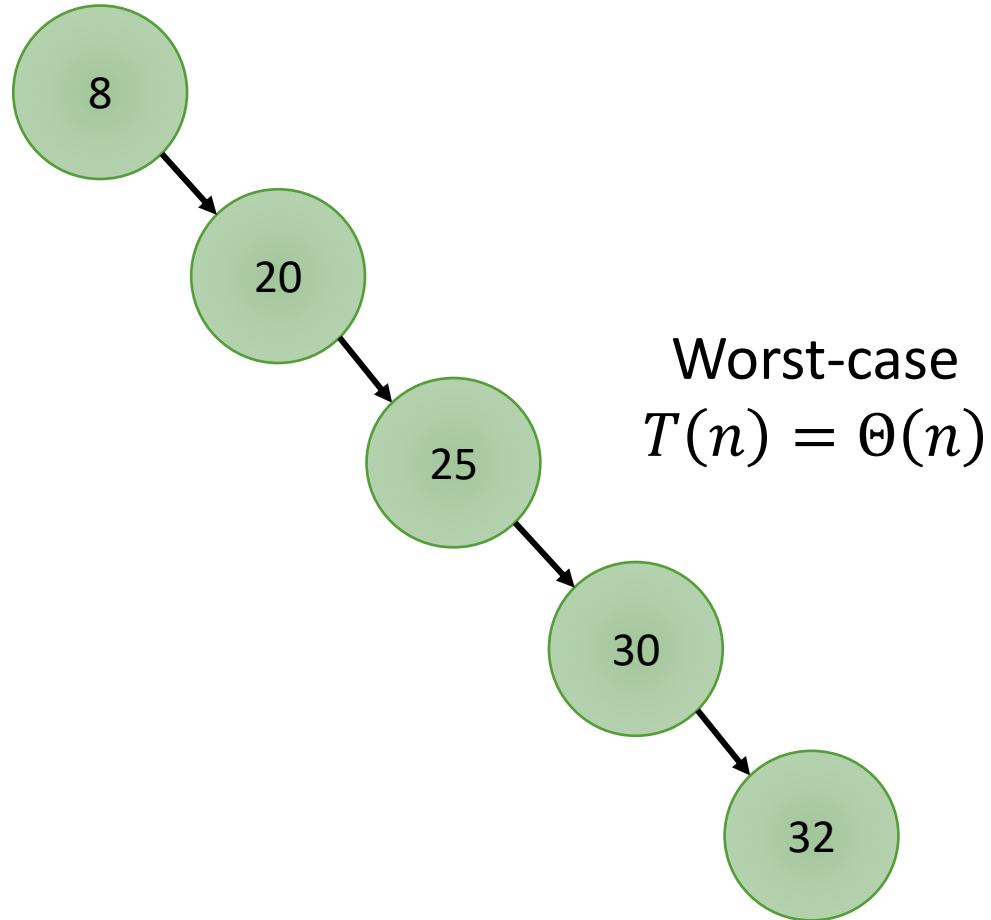
$$T(n) = O(h)$$



# Binary Search Trees (BST)

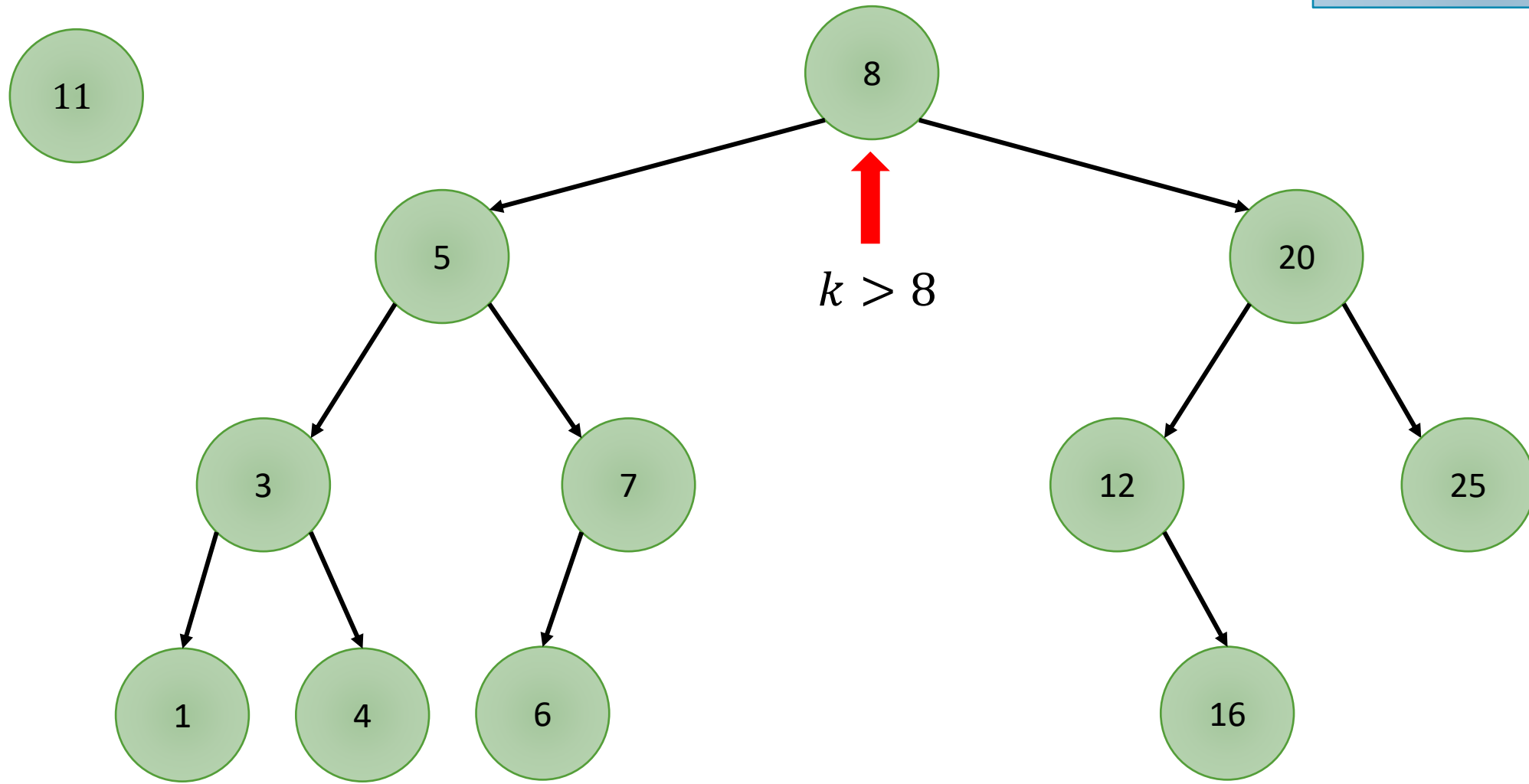
BST-Search( $k$ )

$$T(n) = O(h)$$



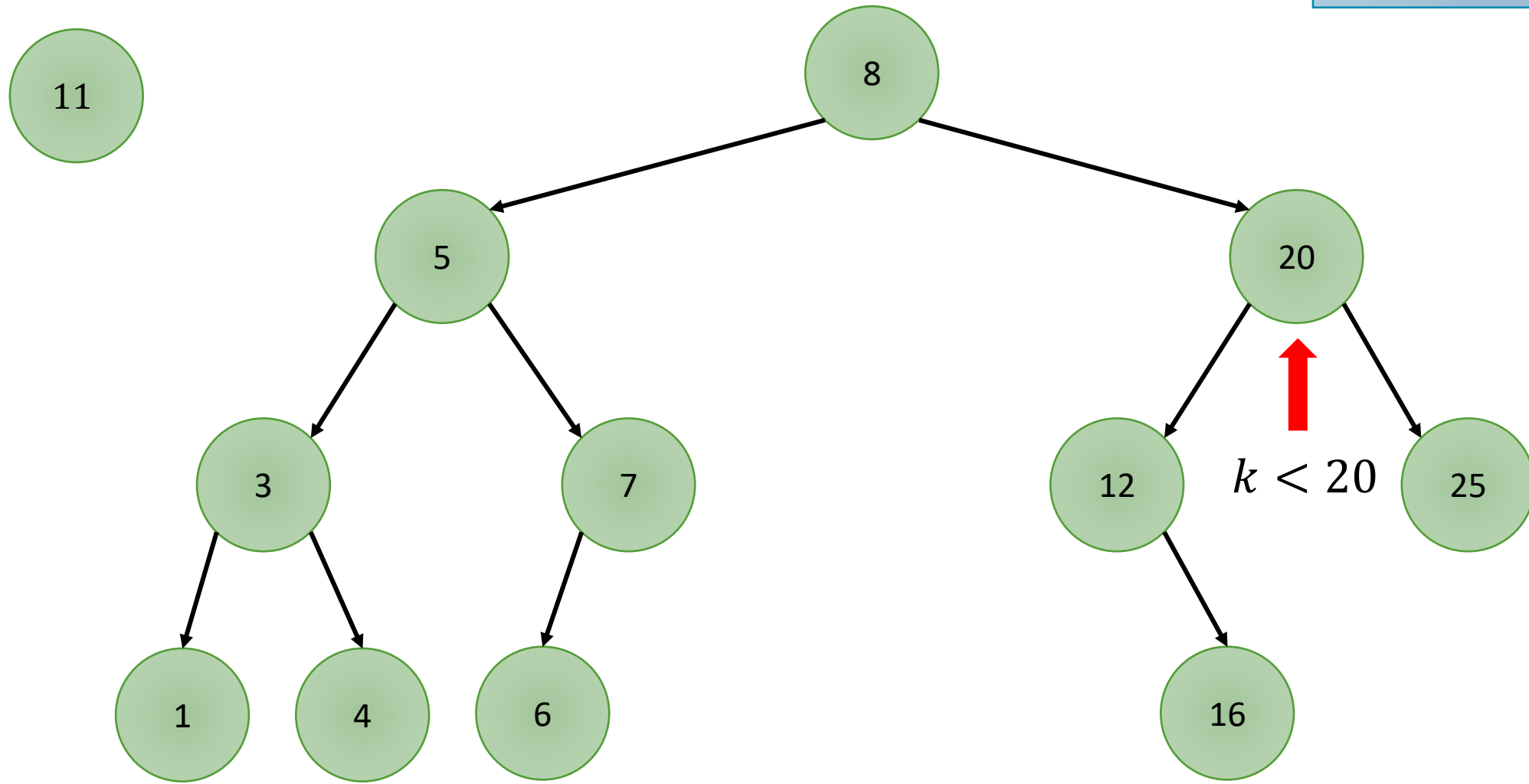
# Binary Search Trees (BST)

BST-Insert( $k$ )



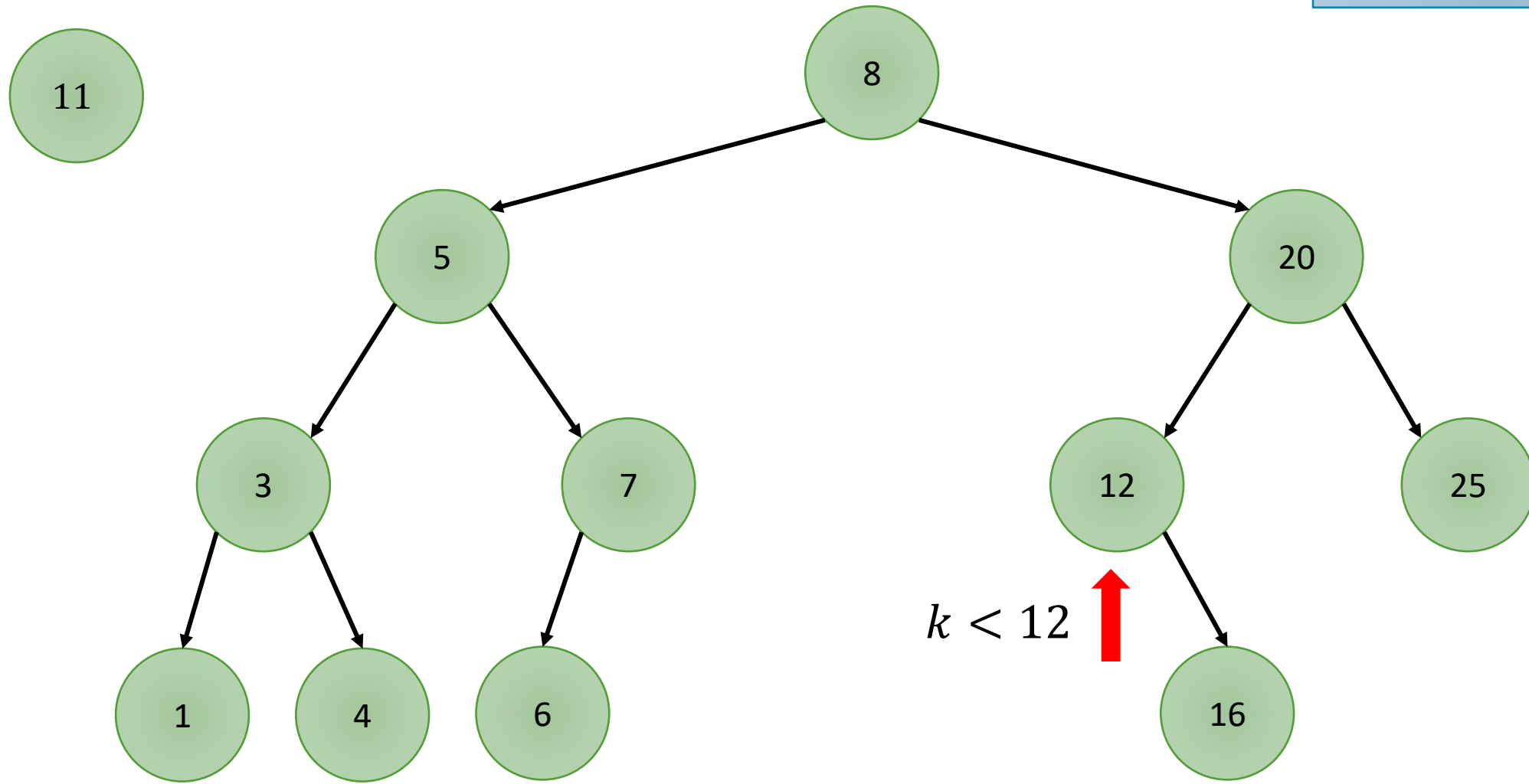
# Binary Search Trees (BST)

BST-Insert( $k$ )



# Binary Search Trees (BST)

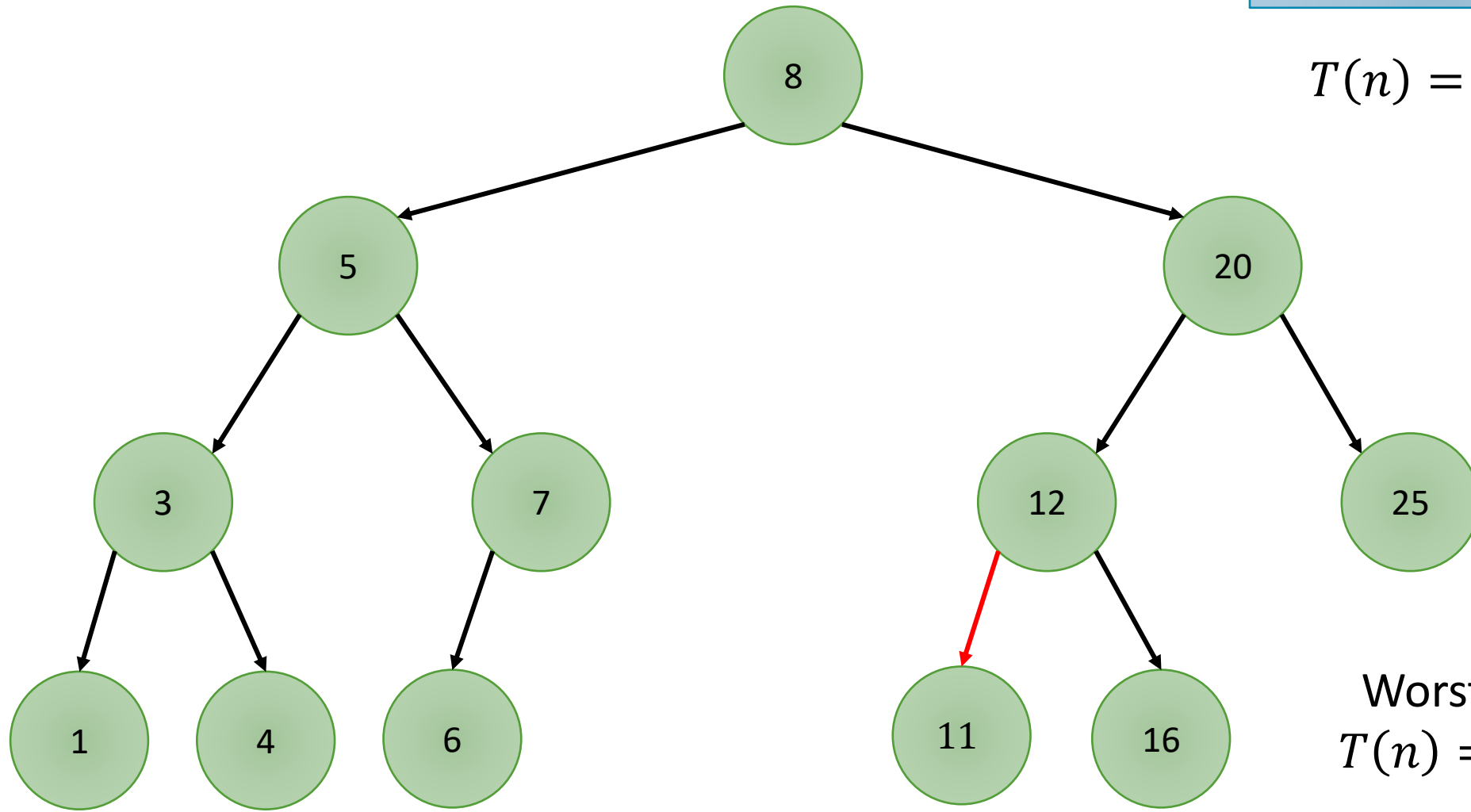
BST-Insert( $k$ )



# Binary Search Trees (BST)

BST-Insert( $k$ )

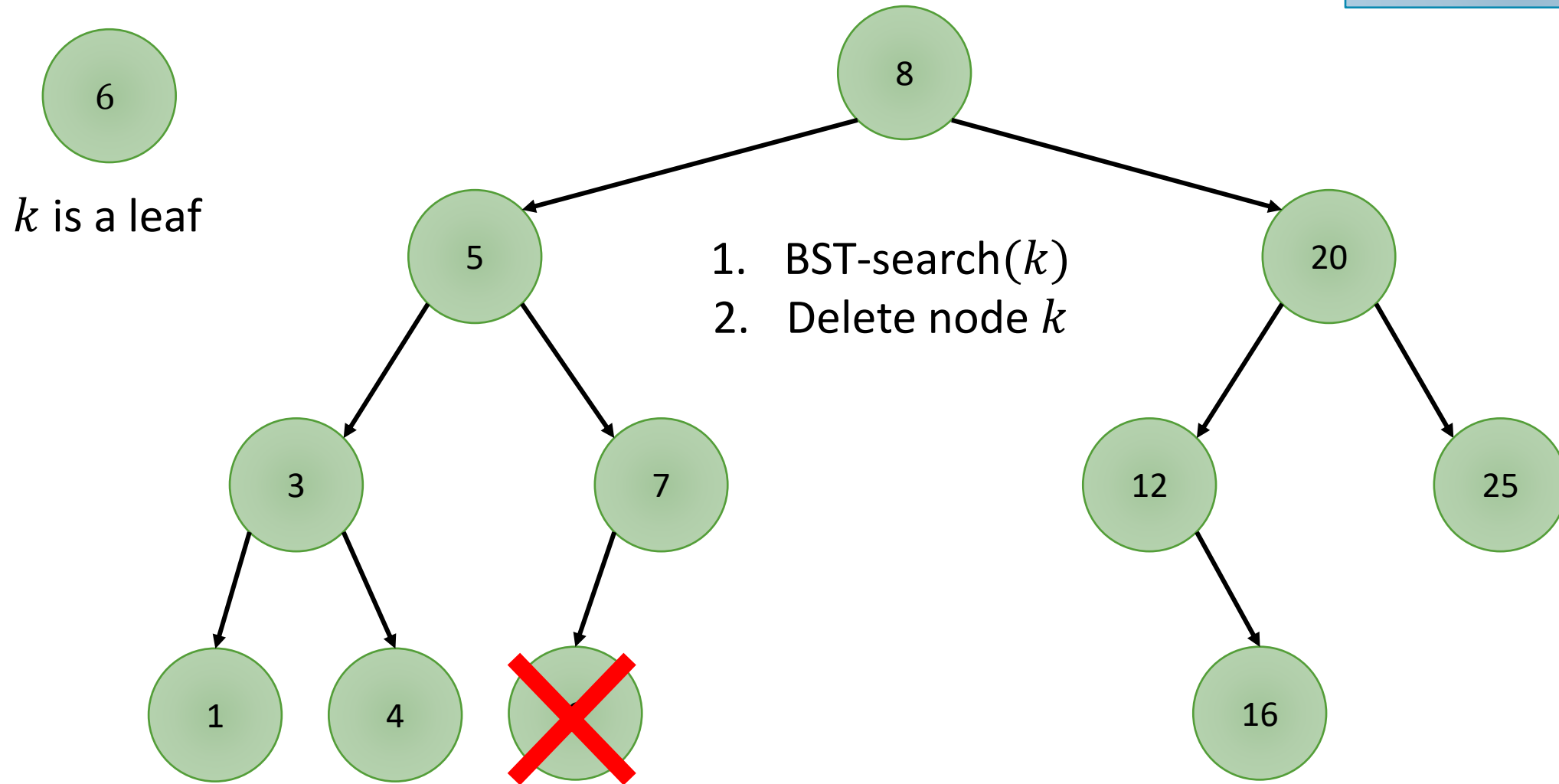
$$T(n) = O(h)$$



Worst-case  
 $T(n) = \Theta(n)$

# Binary Search Trees (BST)

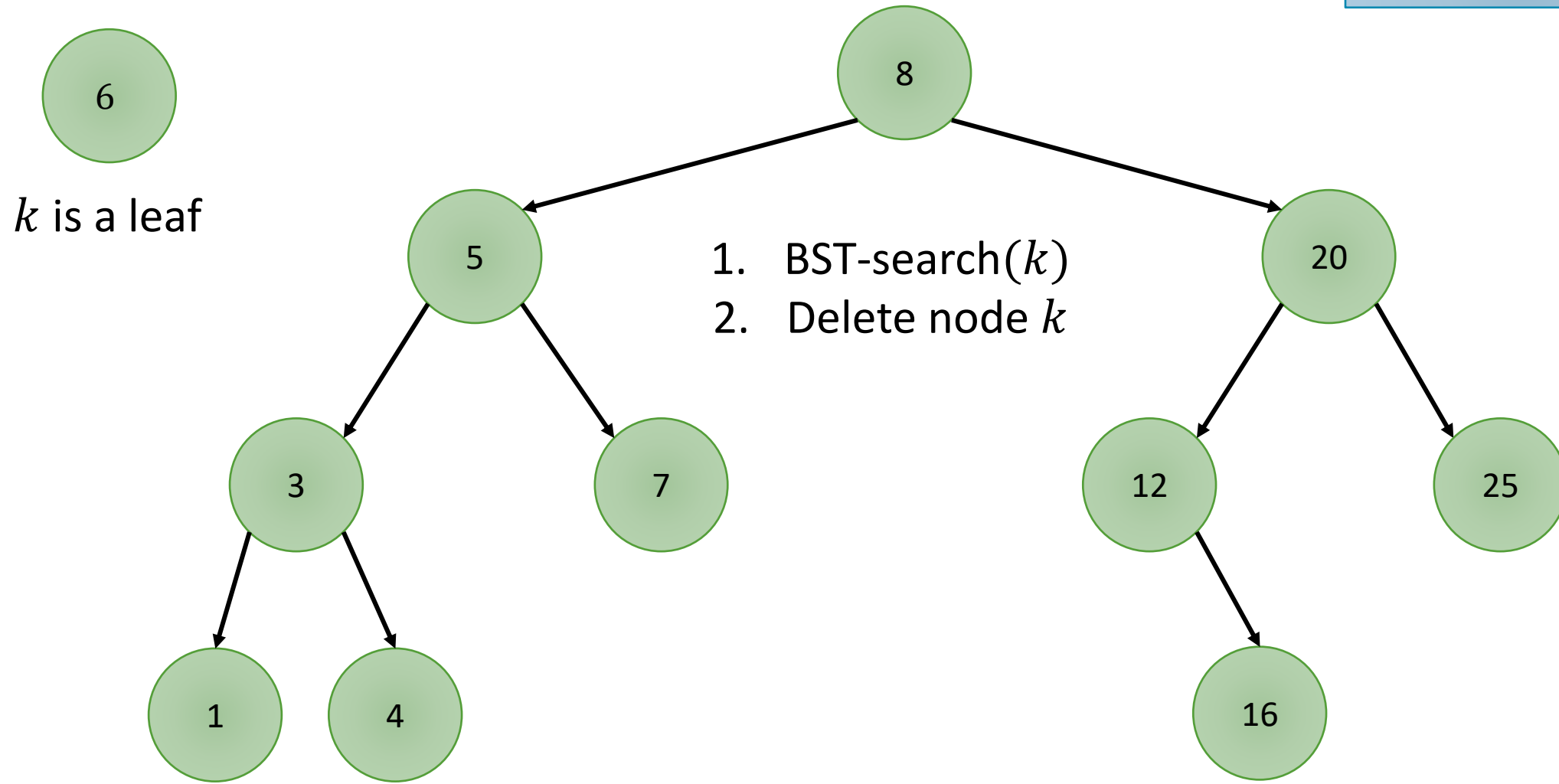
BST-delete( $k$ )





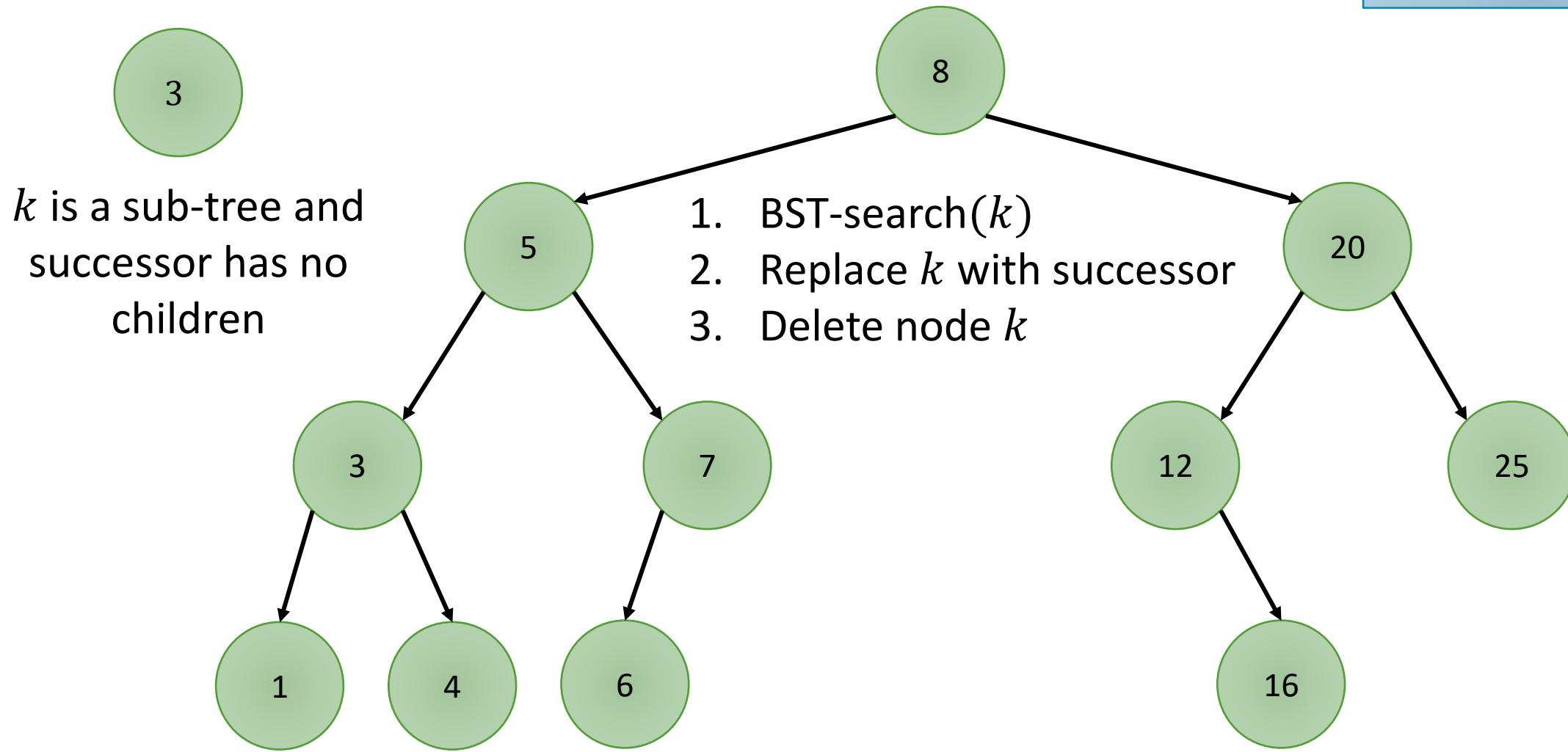
# Binary Search Trees (BST)

BST-delete( $k$ )



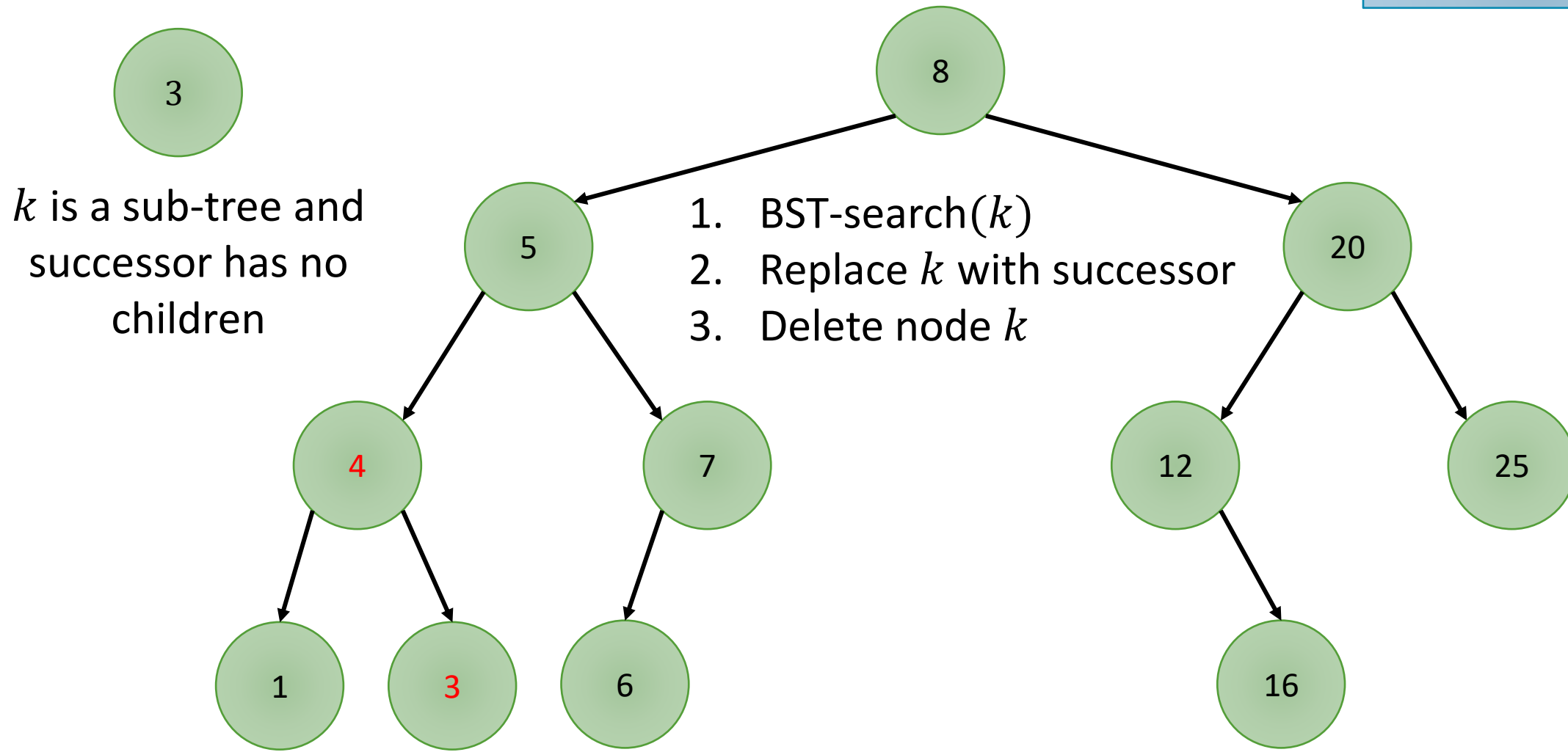
# Binary Search Trees (BST)

BST-delete( $k$ )



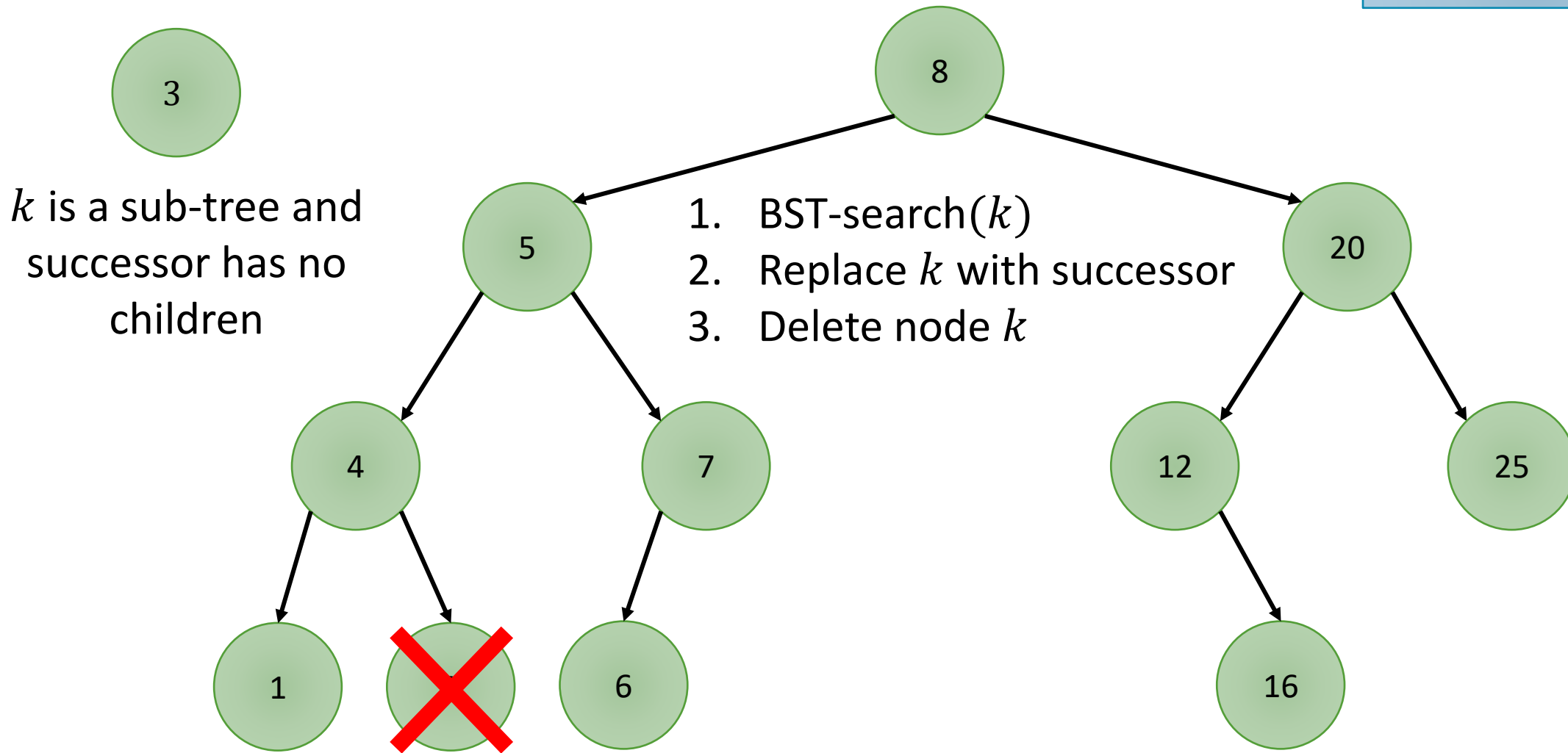
# Binary Search Trees (BST)

BST-delete( $k$ )



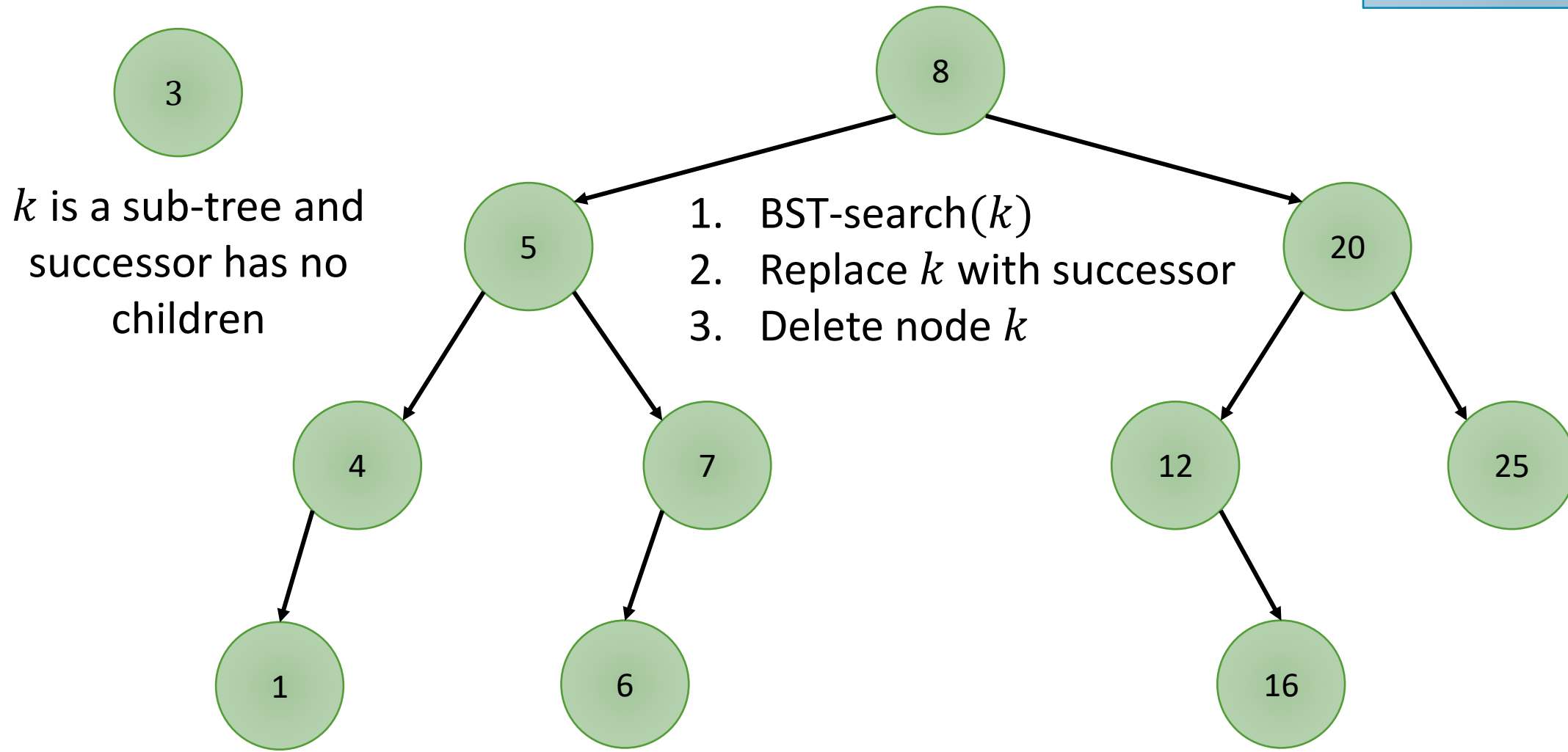
# Binary Search Trees (BST)

BST-delete( $k$ )



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BST-delete( $k$ )

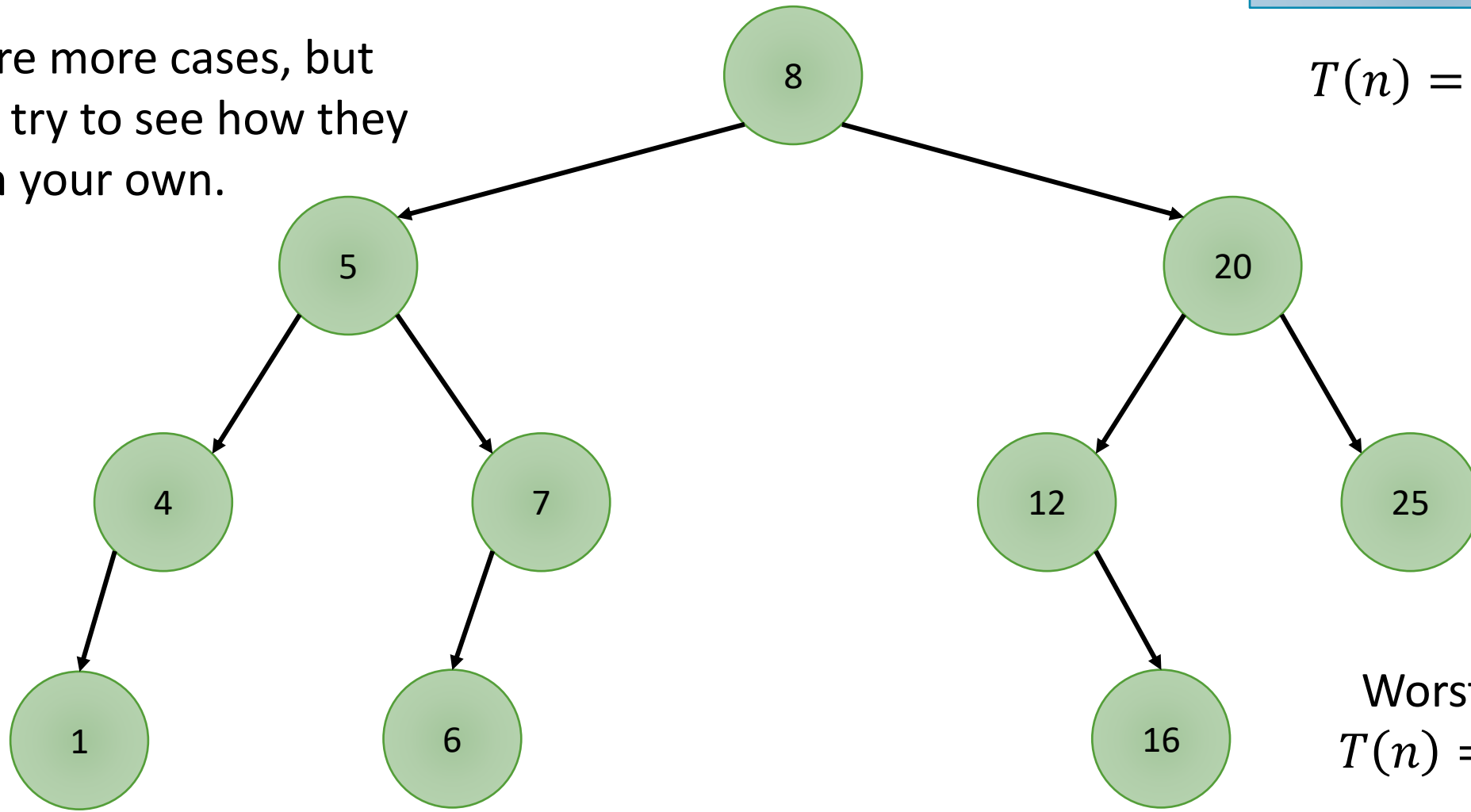


# Binary Search Trees (BST)

There are more cases, but you can try to see how they work on your own.

BST-delete( $k$ )

$$T(n) = O(h)$$



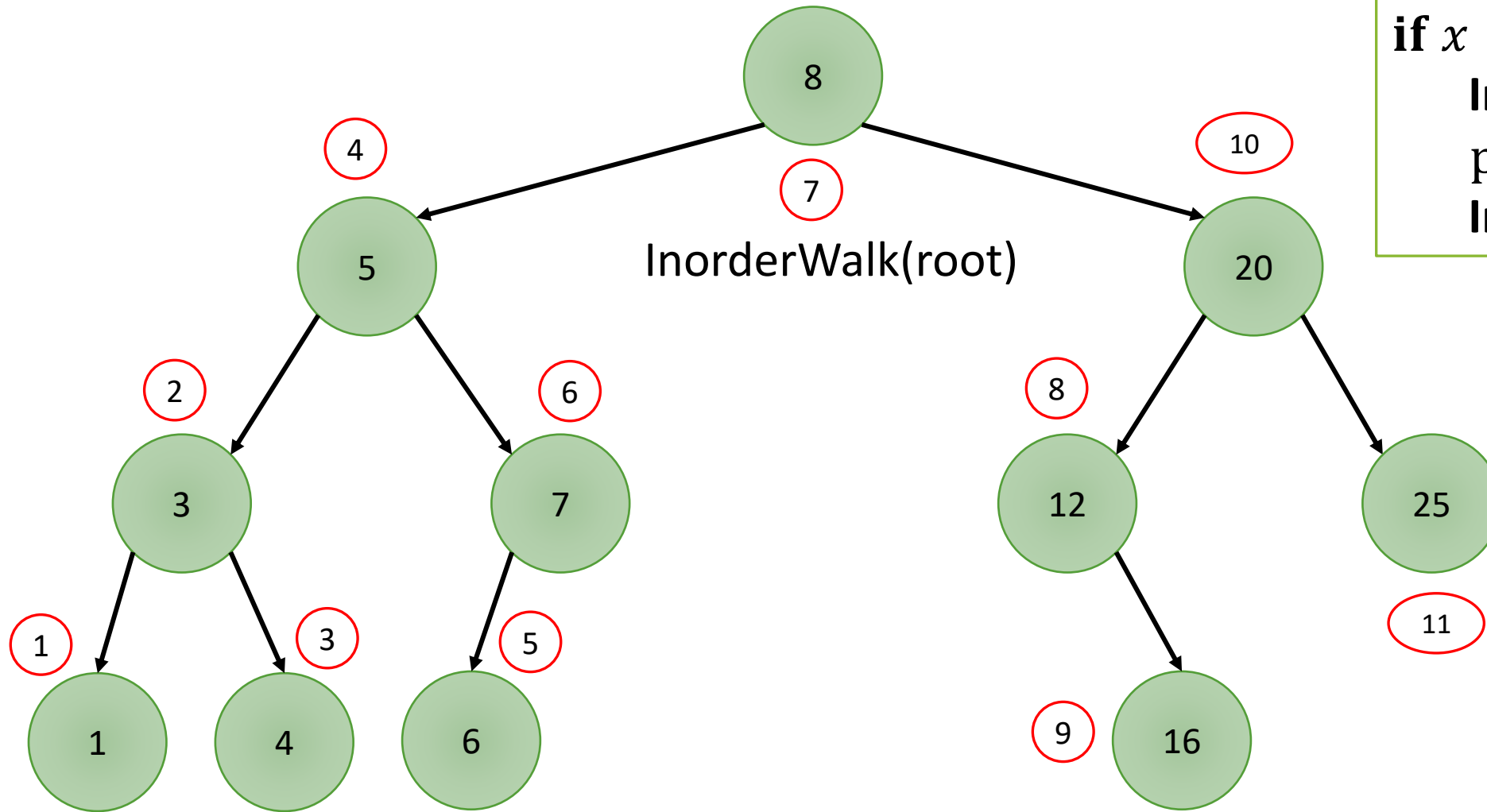
Worst-case  
 $T(n) = \Theta(n)$

# Binary Search Trees (BST)

## In-Order Traversal

```
InorderWalk(x):  
if x ≠ NULL  
    InorderWalk(x.left)  
    print(x.key)  
    InorderWalk(x.right)
```

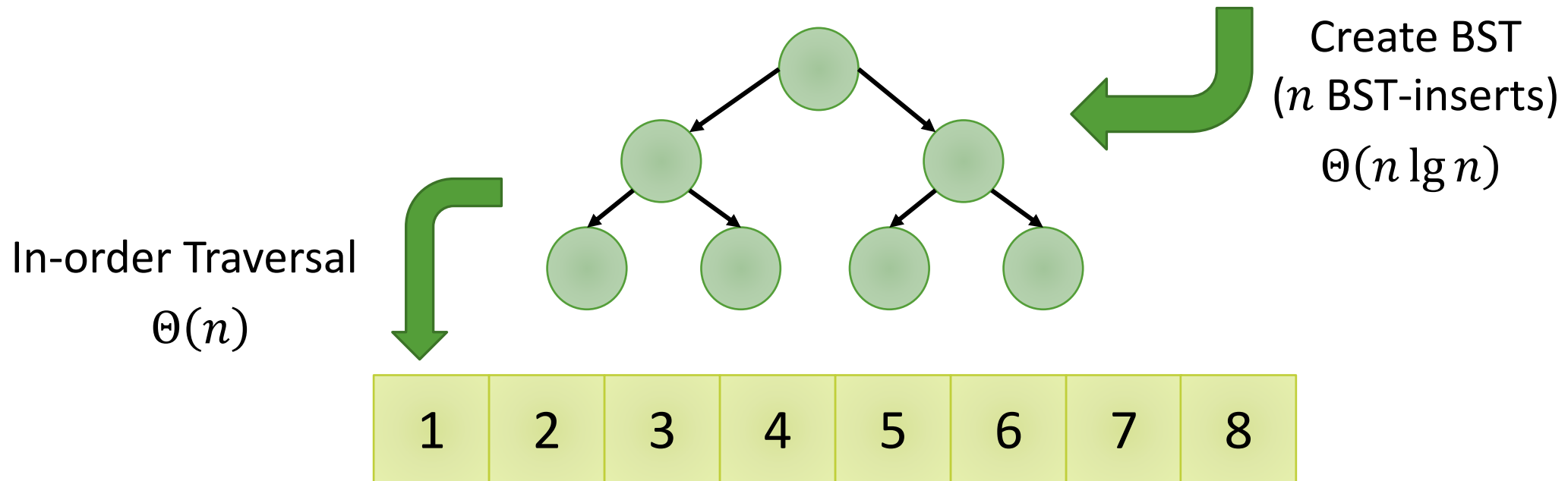
$$T(n) = O(n)$$



Assuming Tree is  
Balanced!

BST Sort  $T(n) = \Theta(n \lg n) + \Theta(n) = \Theta(n \lg n)$

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# BST Sort

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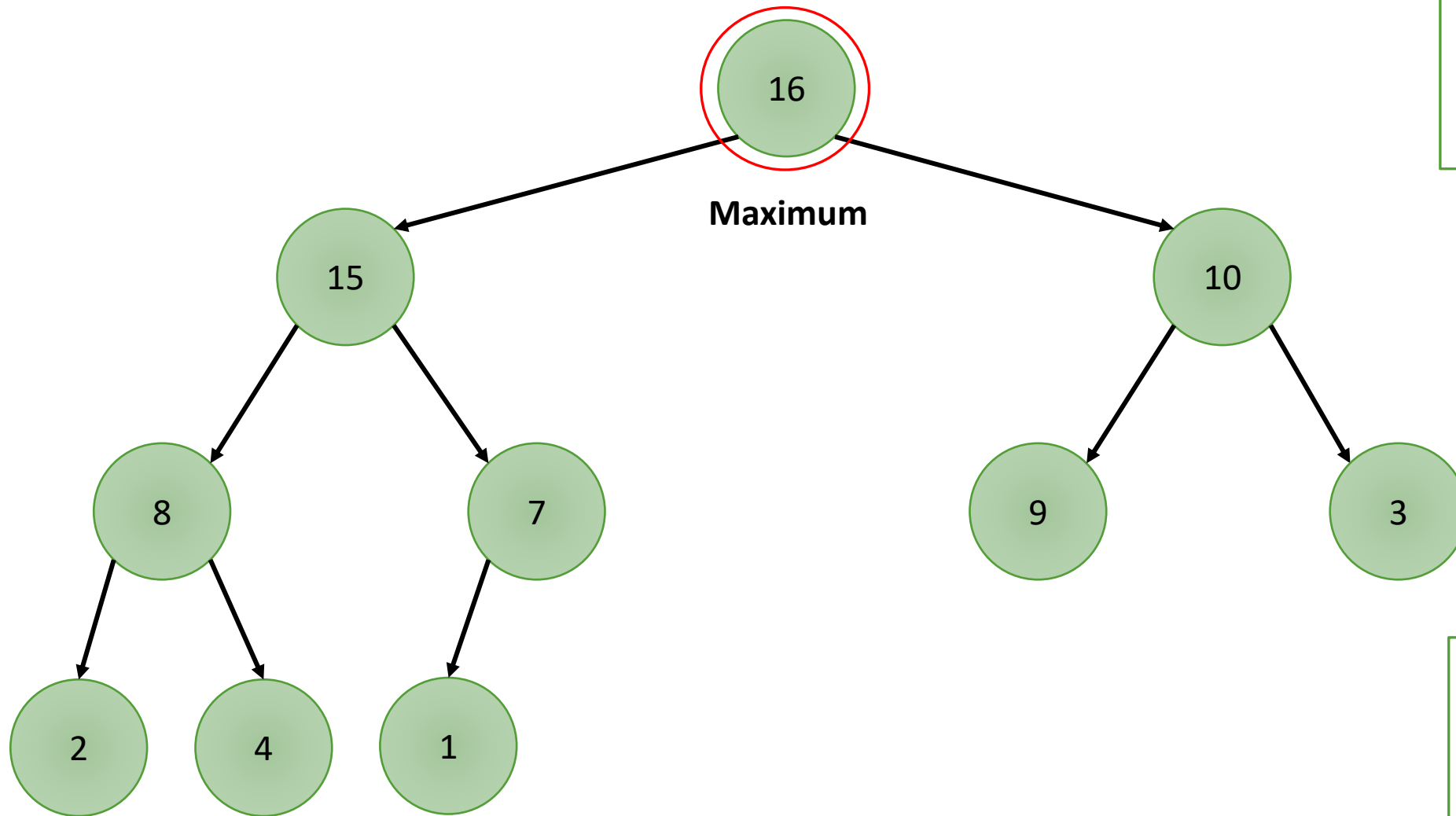
- In order to make sure that the tree is balanced, we can use various other tree-based data structures:
  - Red-black Trees
  - AVL Trees (better than RB-trees for search-intensive applications)
  - Splay Trees (pushes frequently accessed elements closer to the root; useful for caching)
  - Treap (a combination of a tree and a heap)
  - B-Trees (for large multi-degree data)
  - 2-3 / 2-3-4 trees (special cases of B-trees)
- But on average, randomly inserting elements in a standard BST gives  $O(n \lg n)$

# Heapsort: At a glance

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- Worst-case  $O(n \lg n)$  - like merge-sort, but unlike insertion sort
- Sorts in place – like insertion, but unlike merge-sort
- Heapsort combines the best attributes of these deterministic sorting algorithms.

# Binary Max-Heap



## Key Property:

Max-Heap a binary tree data structure where for every node  $x$ :

$$\text{Parent}(x).key \geq x.key$$

## Max-Heap operations:

- 1.) HEAPIFY( $i$ )
- 2.) HEAP-getmax
- 3.) HEAP-insert( $k$ )

# Binary Max-Heap

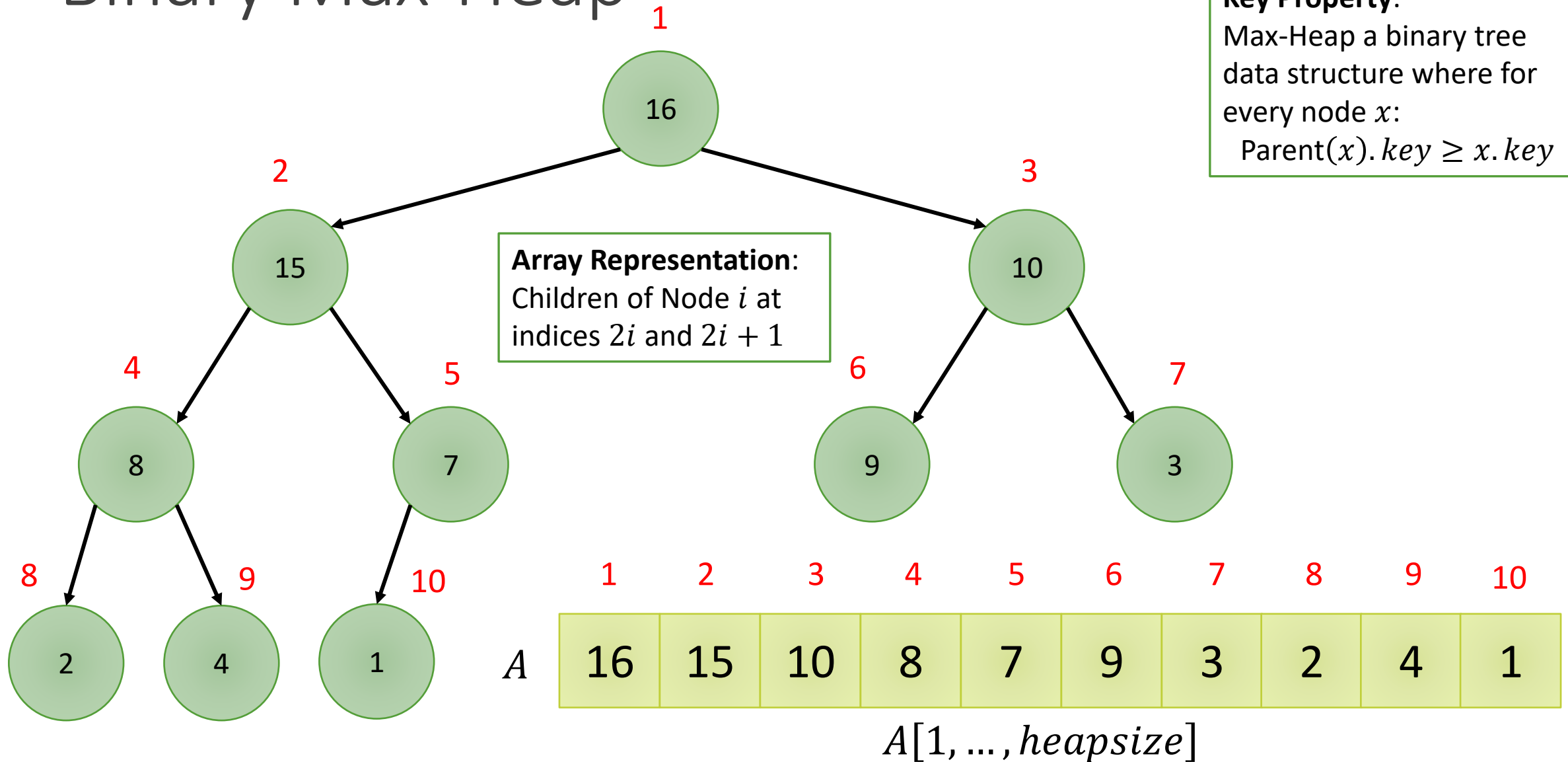
**Key Property:**

Max-Heap a binary tree data structure where for every node  $x$ :

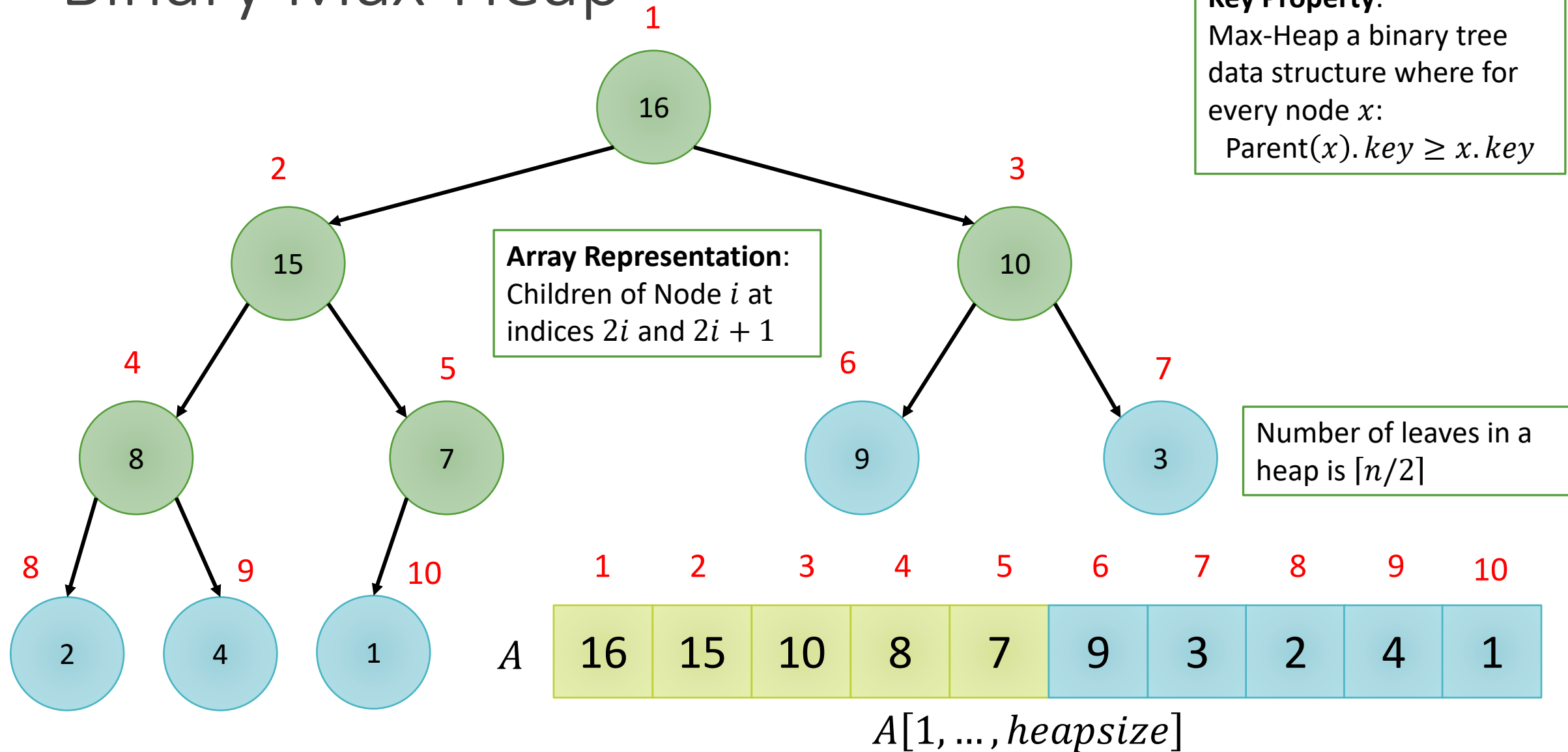
$$\text{Parent}(x).key \geq x.key$$

**Array Representation:**

Children of Node  $i$  at indices  $2i$  and  $2i + 1$

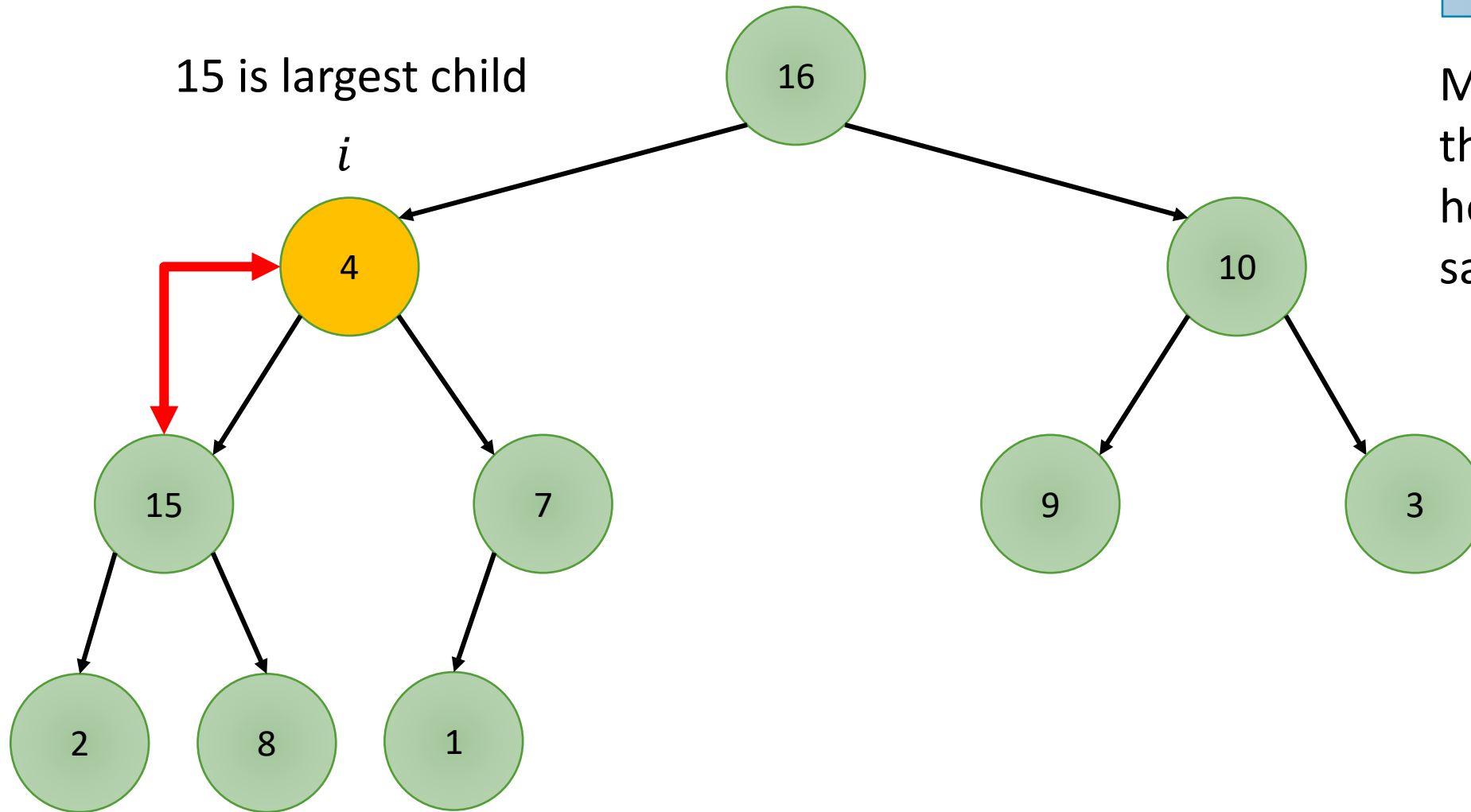


# Binary Max-Heap



# Max-Heap

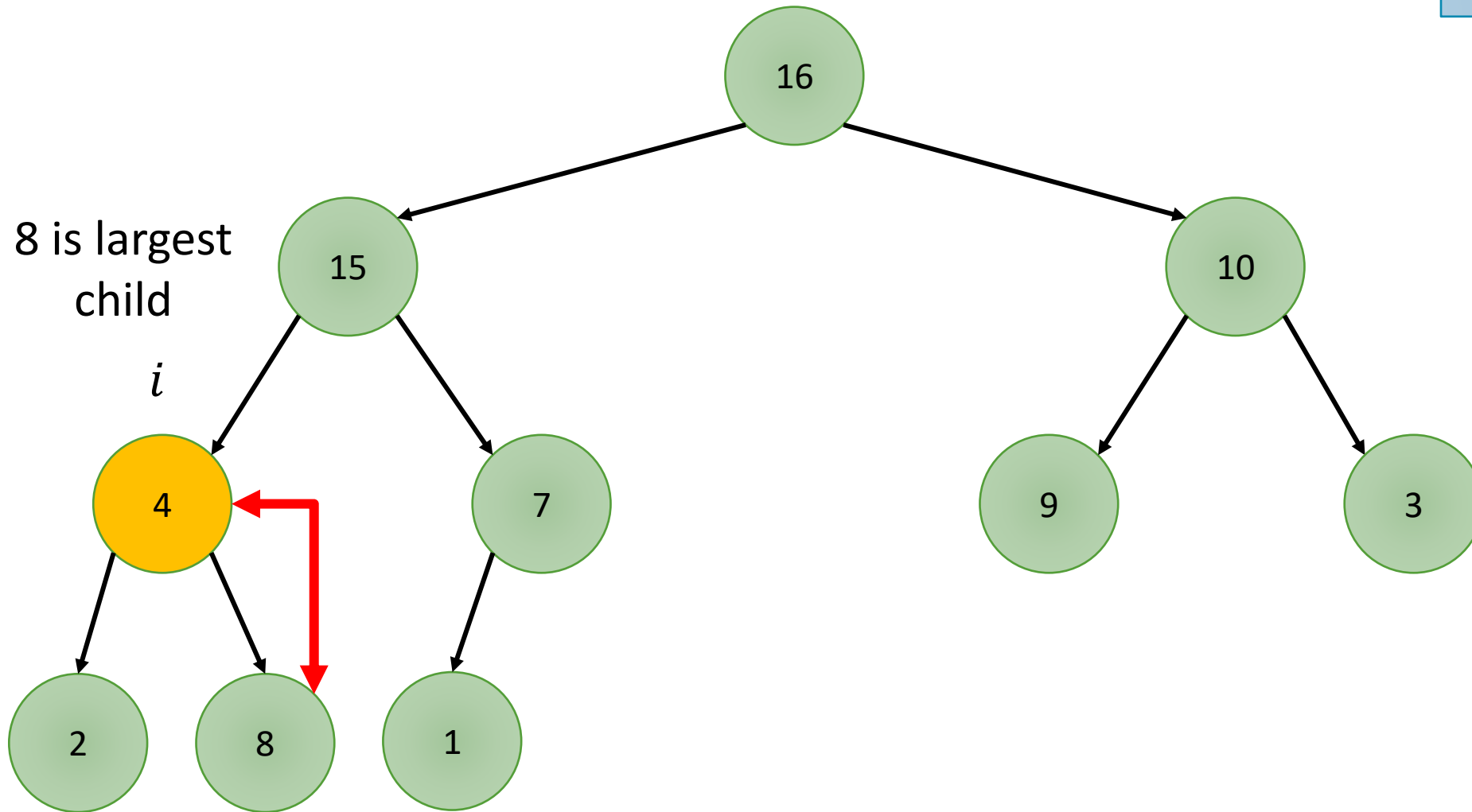
HEAPIFY( $i$ )



Move node  $i$  down the tree until the heap property is satisfied.

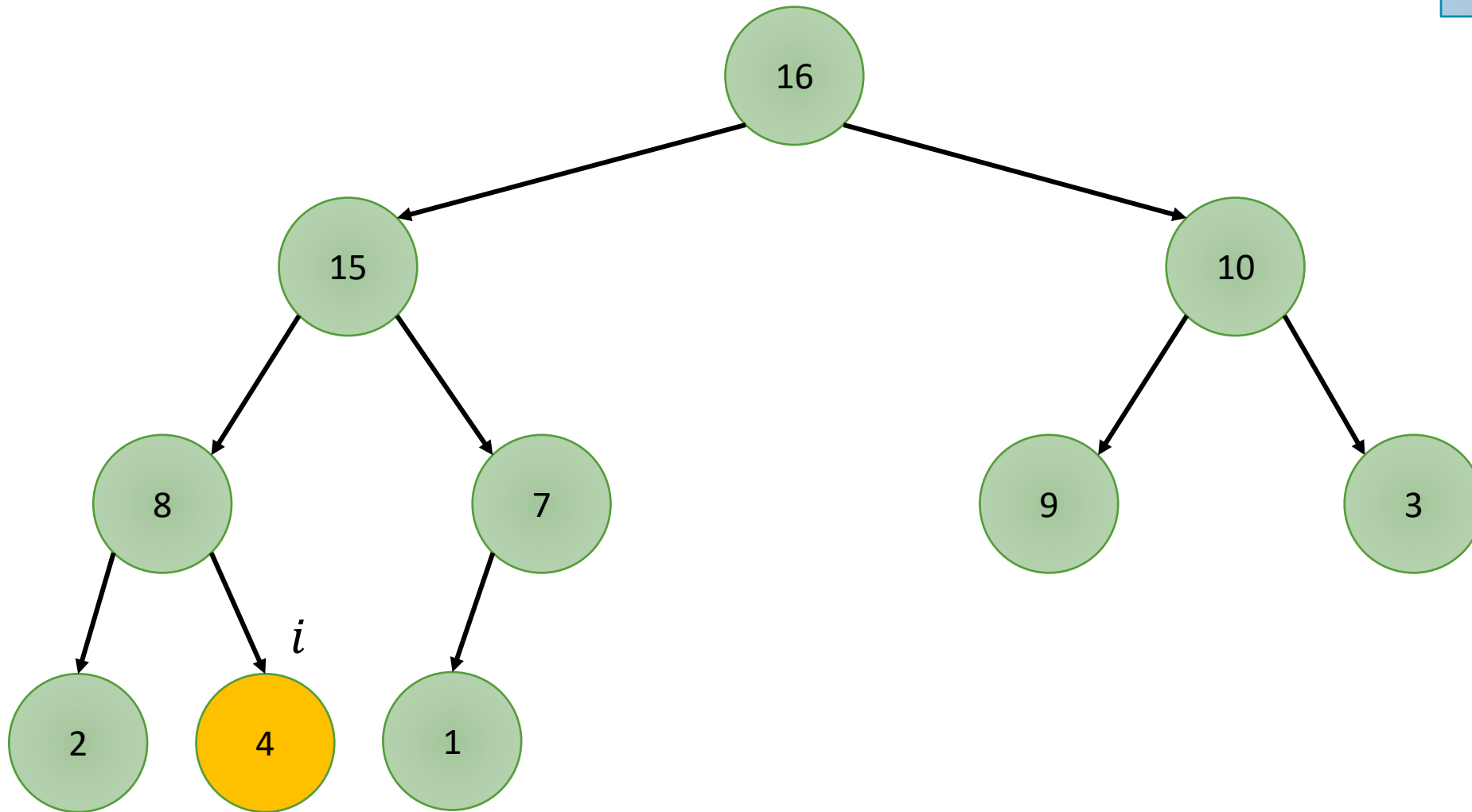
# Max-Heap

HEAPIFY( $i$ )



# Max-Heap

HEAPIFY( $i$ )

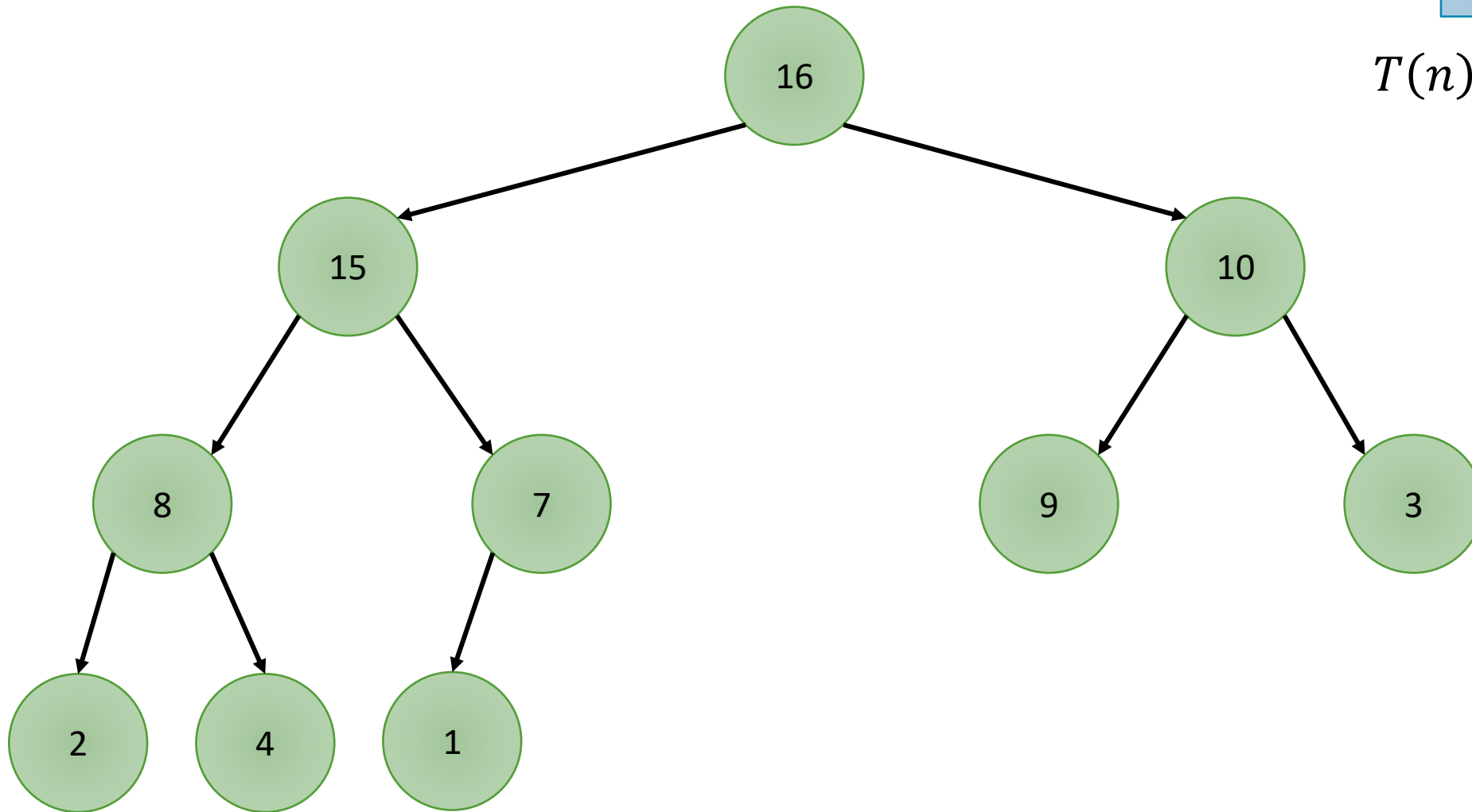




# Max-Heap

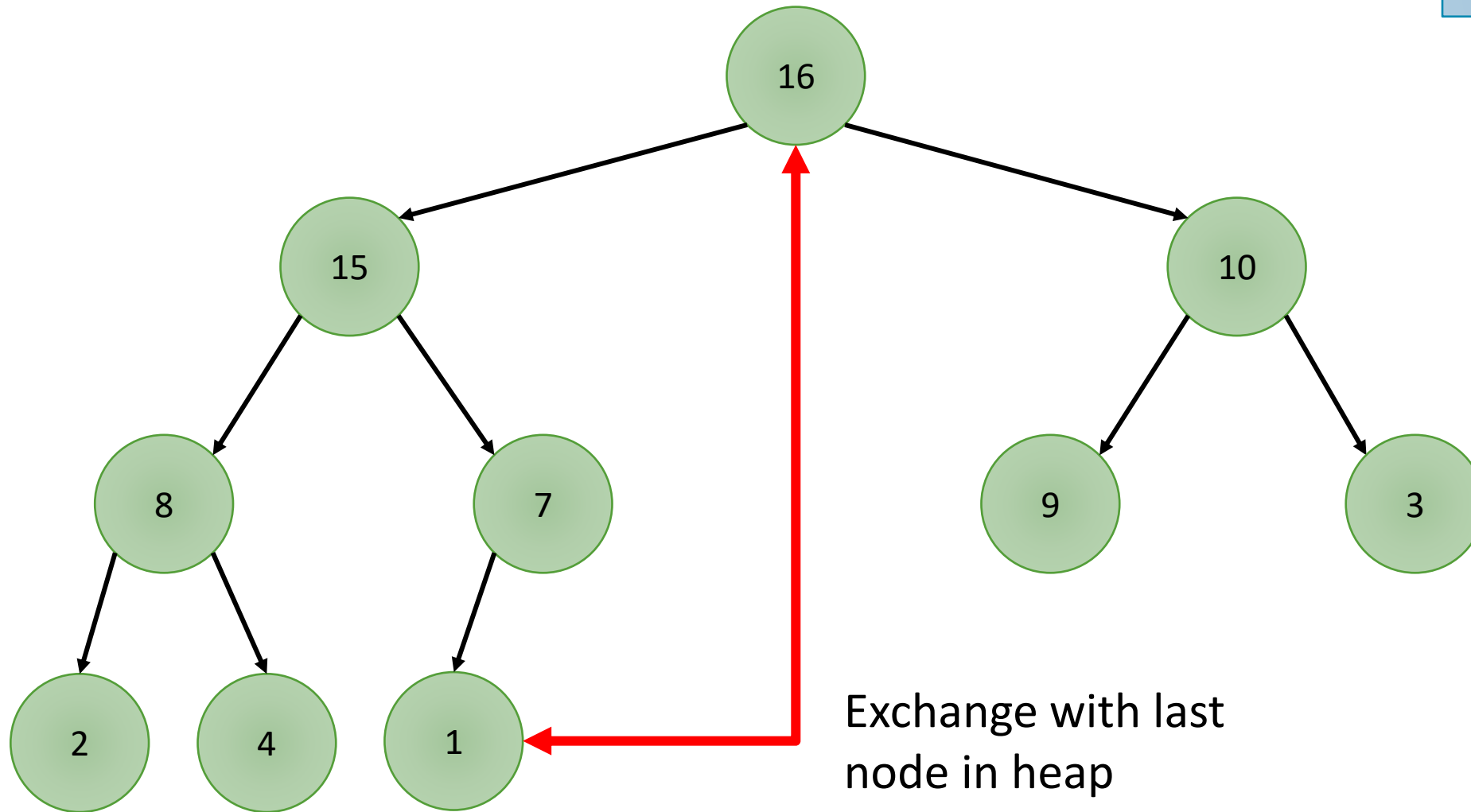
HEAPIFY( $i$ )

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



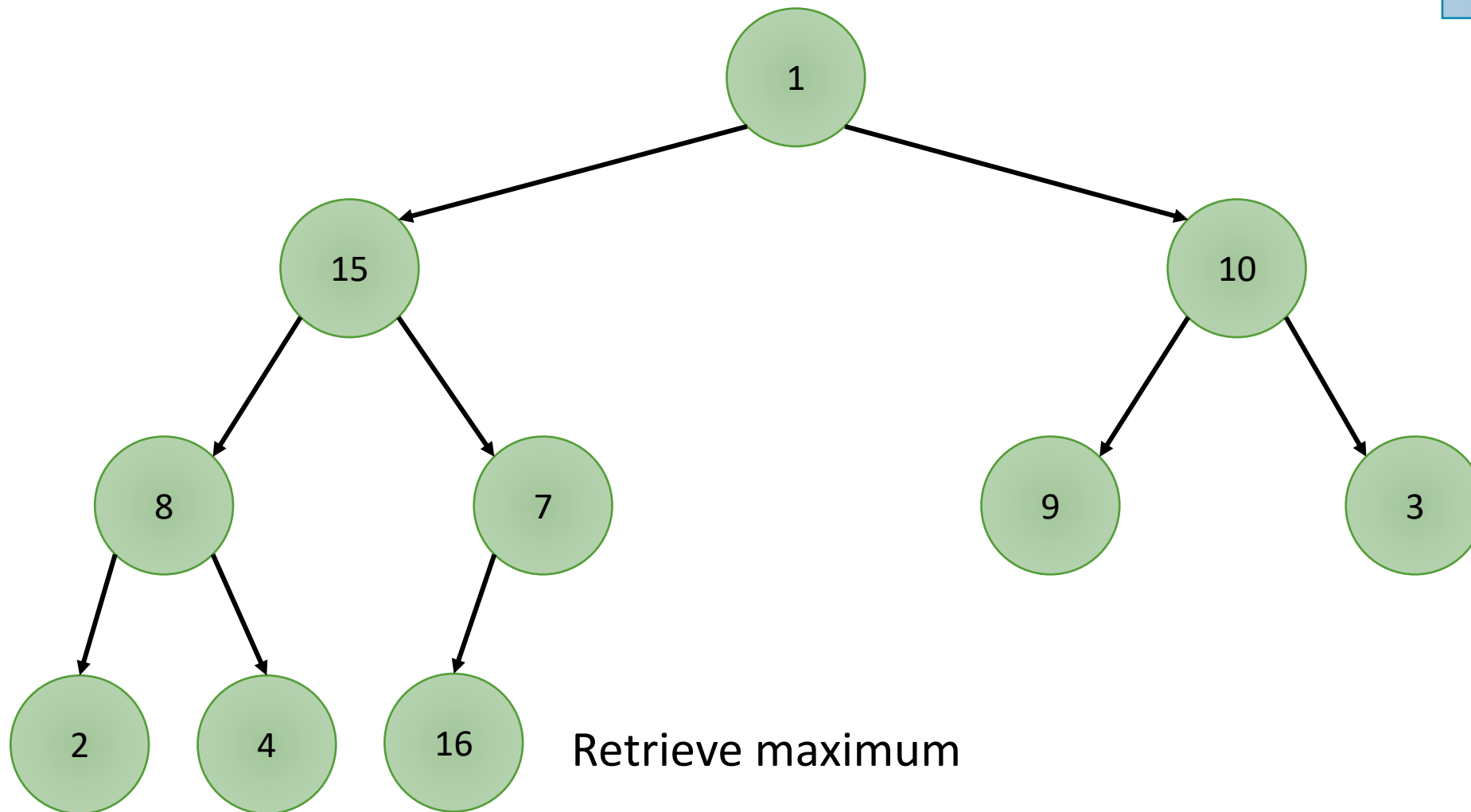
# Max-Heap

HEAP-getmax



# Max-Heap

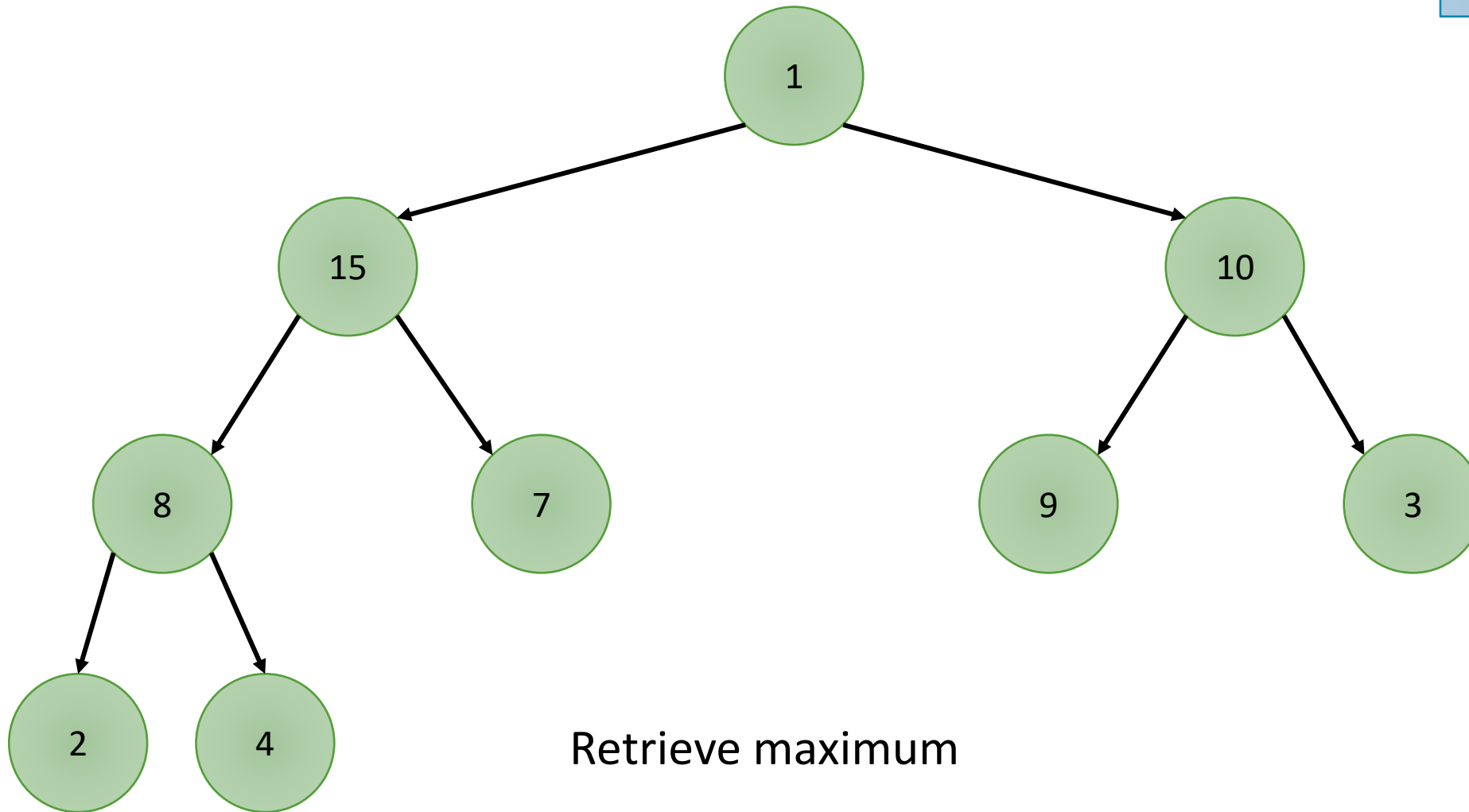
HEAP-getmax



Retrieve maximum

# Max-Heap

HEAP-getmax

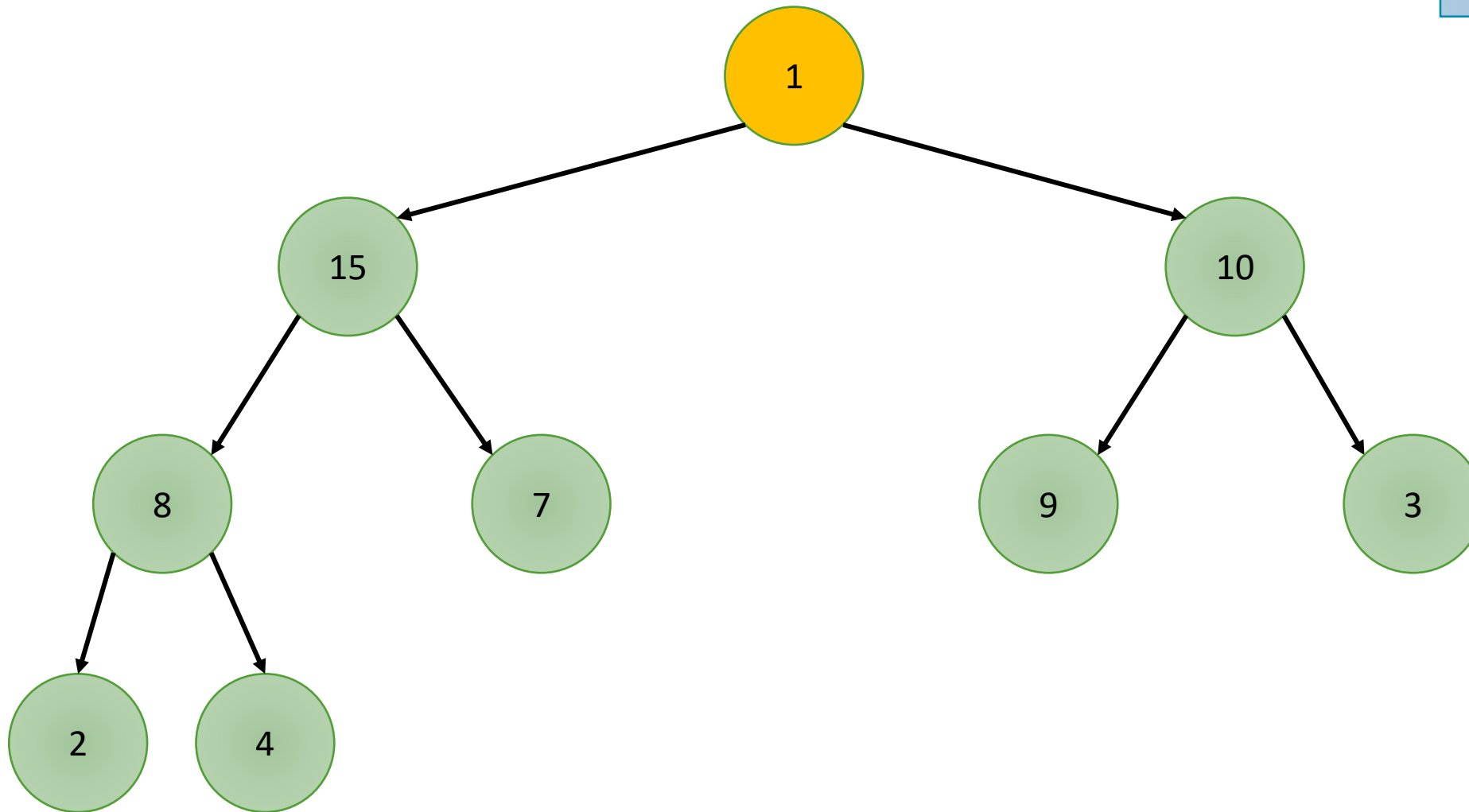


Retrieve maximum

# Max-Heap

Call HEAPIFY(1)

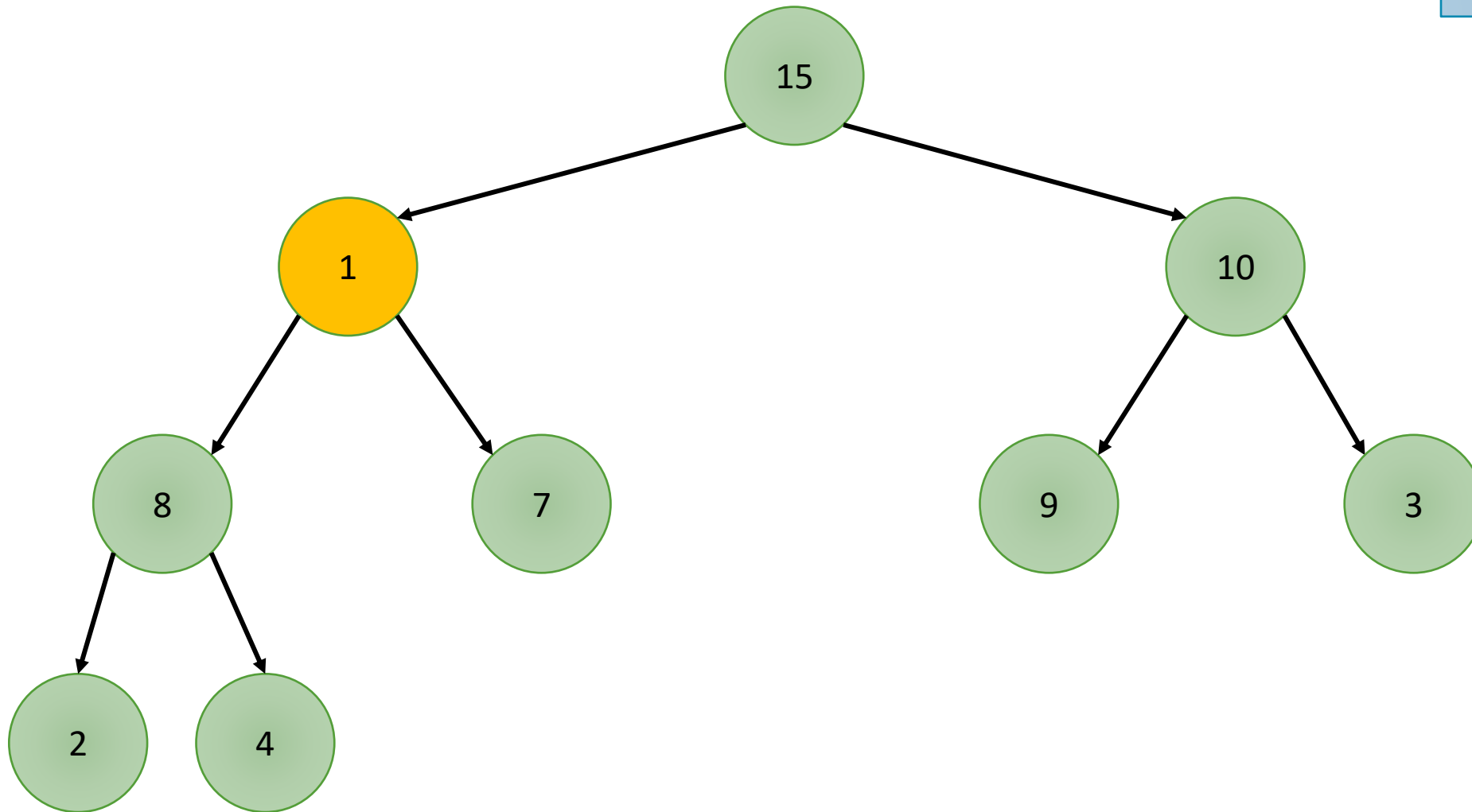
HEAP-getmax



# Max-Heap

Call HEAPIFY(1)

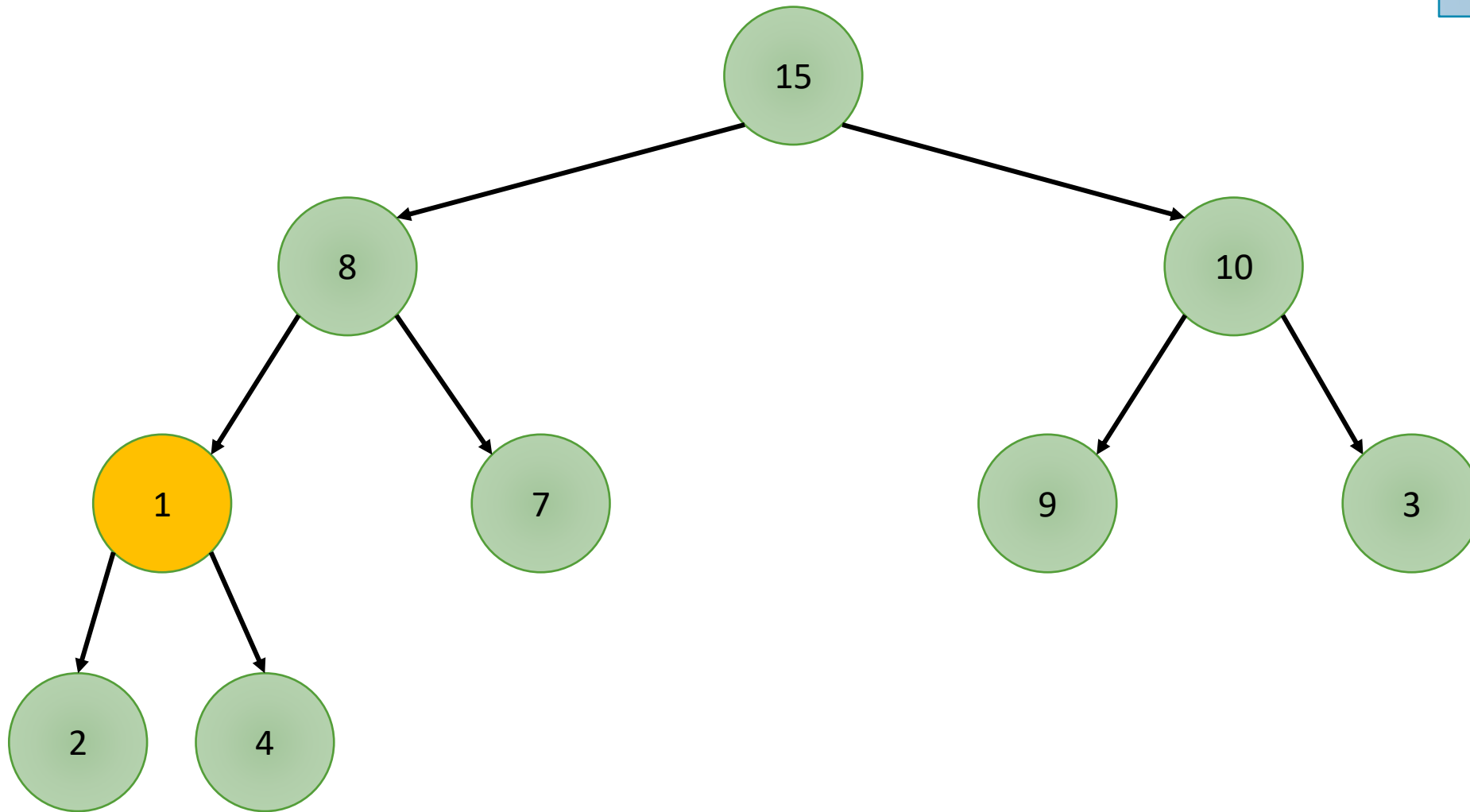
HEAP-getmax



# Max-Heap

Call HEAPIFY(1)

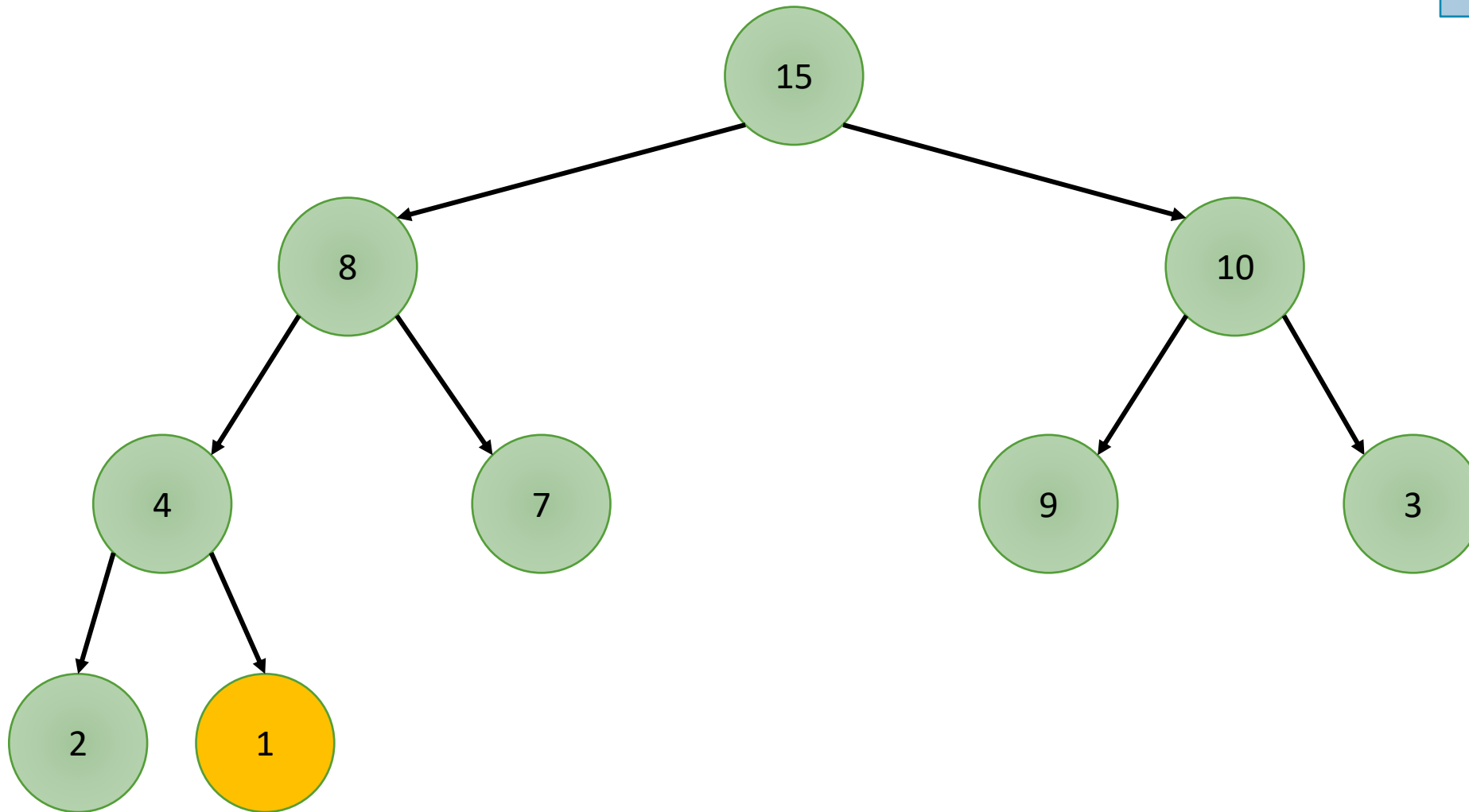
HEAP-getmax



# Max-Heap

Call HEAPIFY(1)

HEAP-getmax



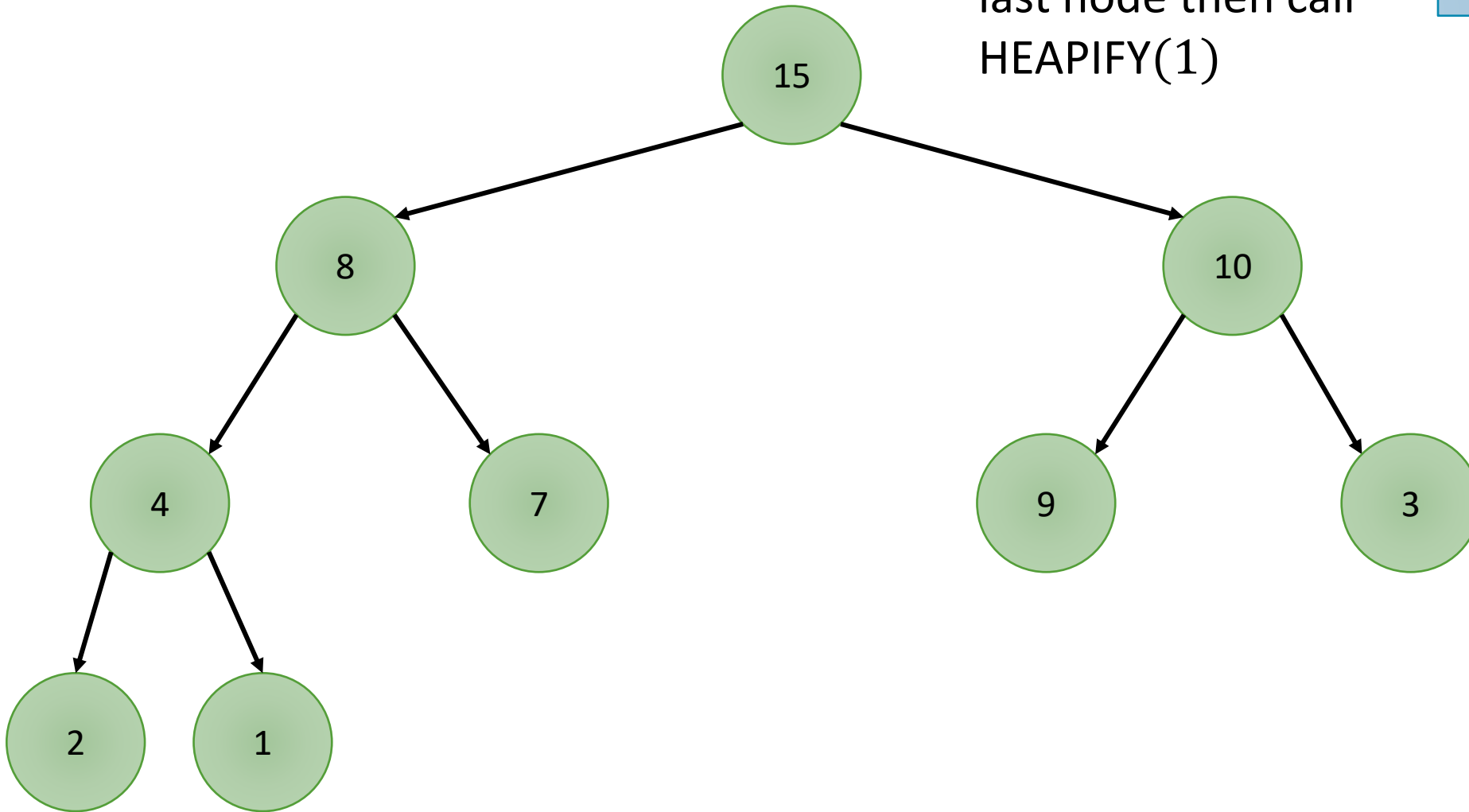


# Max-Heap

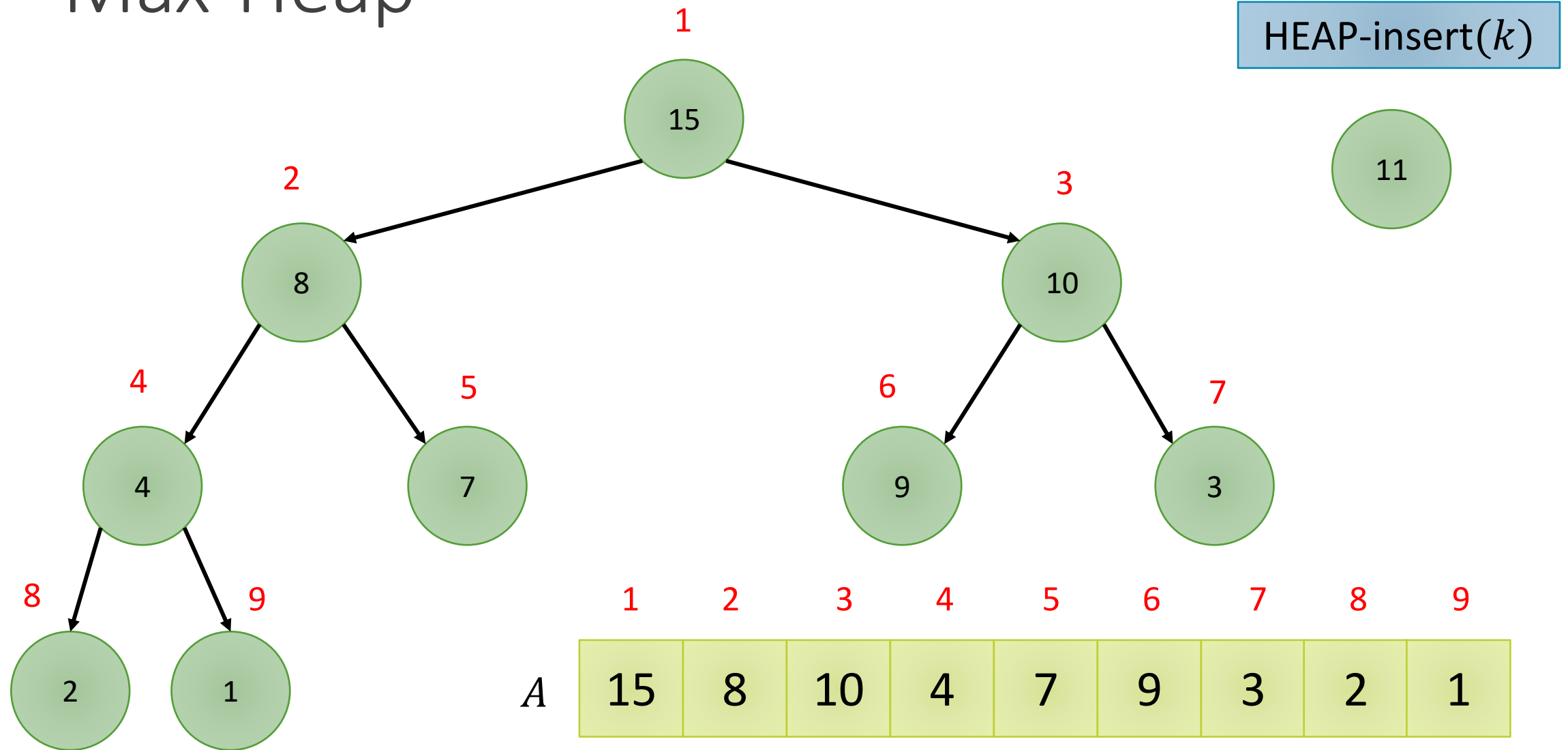
Exchange  $A[1]$  with  
last node then call  
 $\text{HEAPIFY}(1)$

HEAP-getmax

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

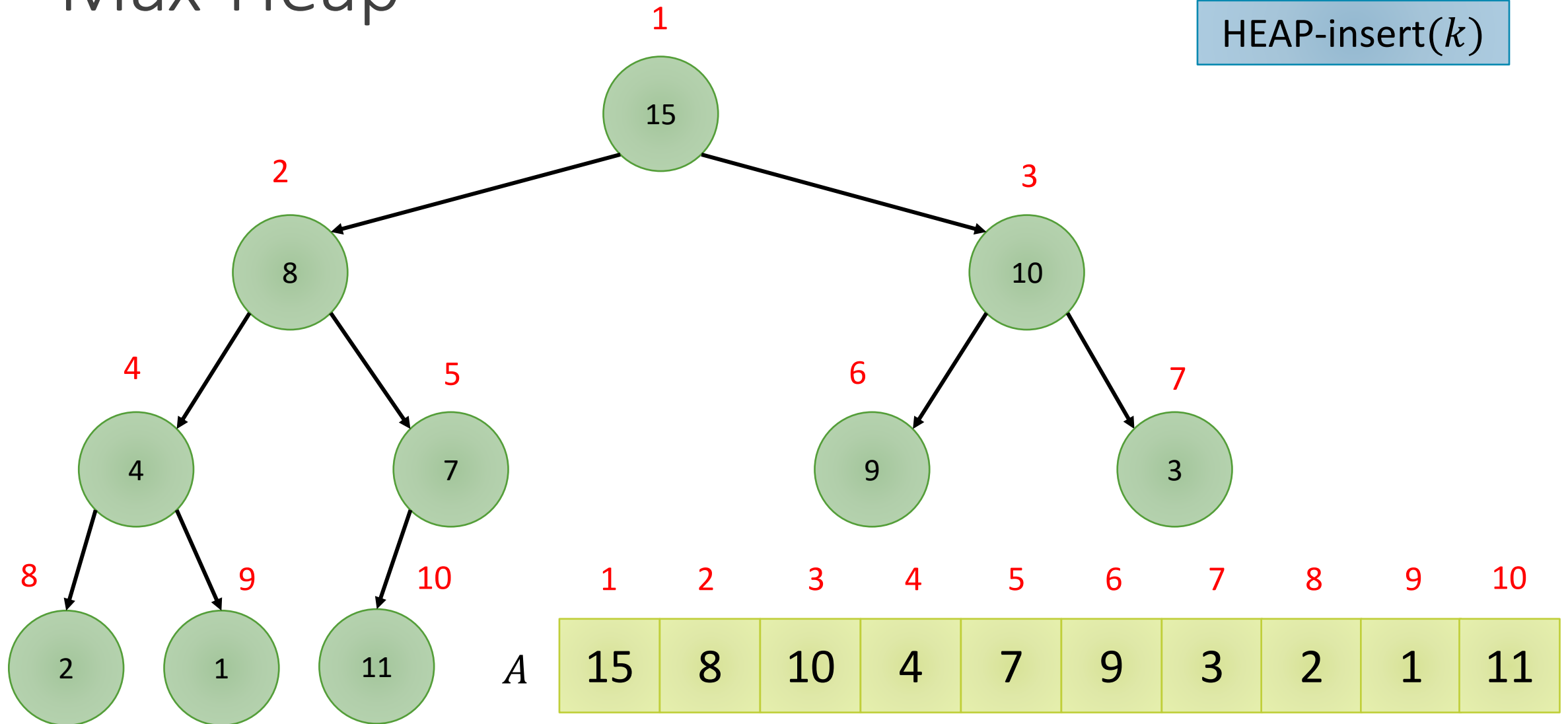


# Max-Heap



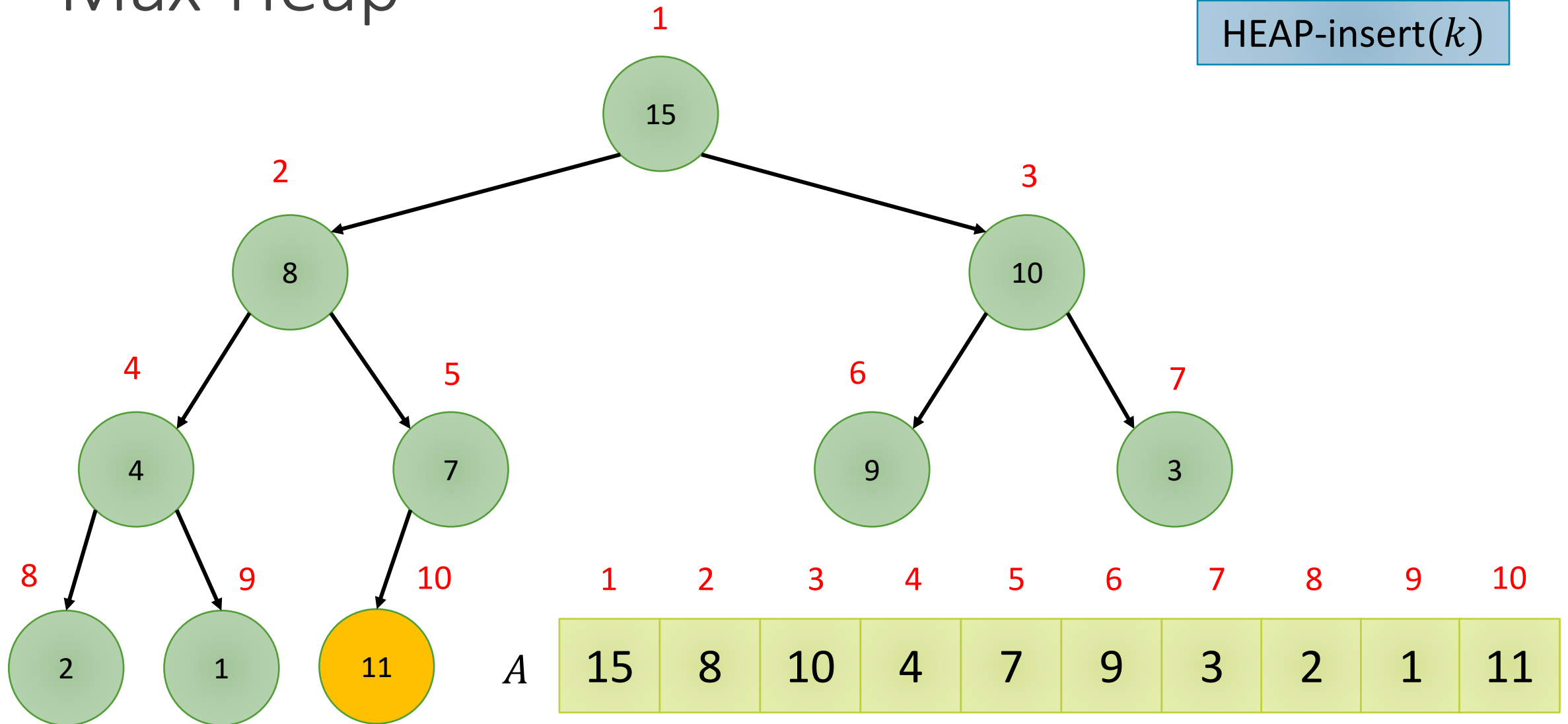
# Max-Heap

HEAP-insert( $k$ )



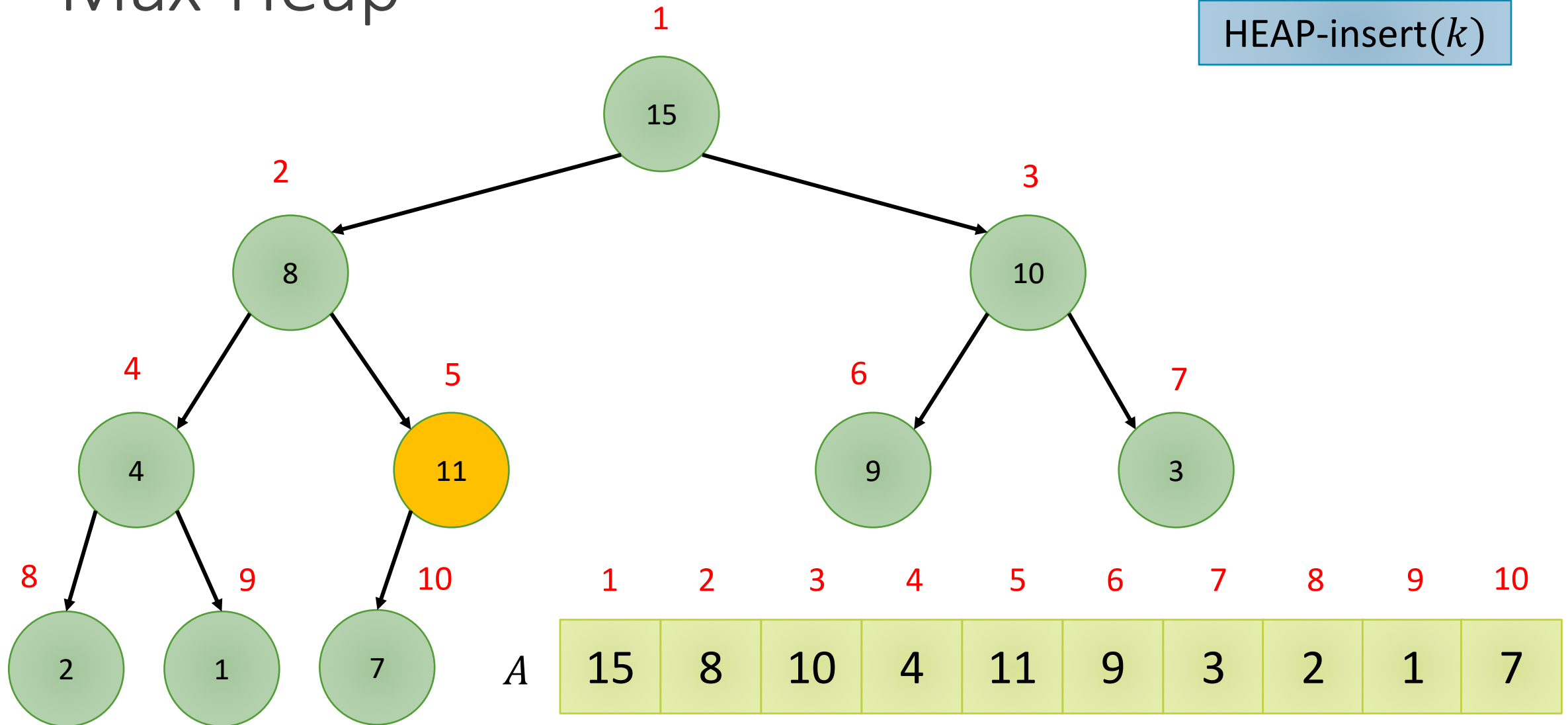
# Max-Heap

HEAP-insert( $k$ )



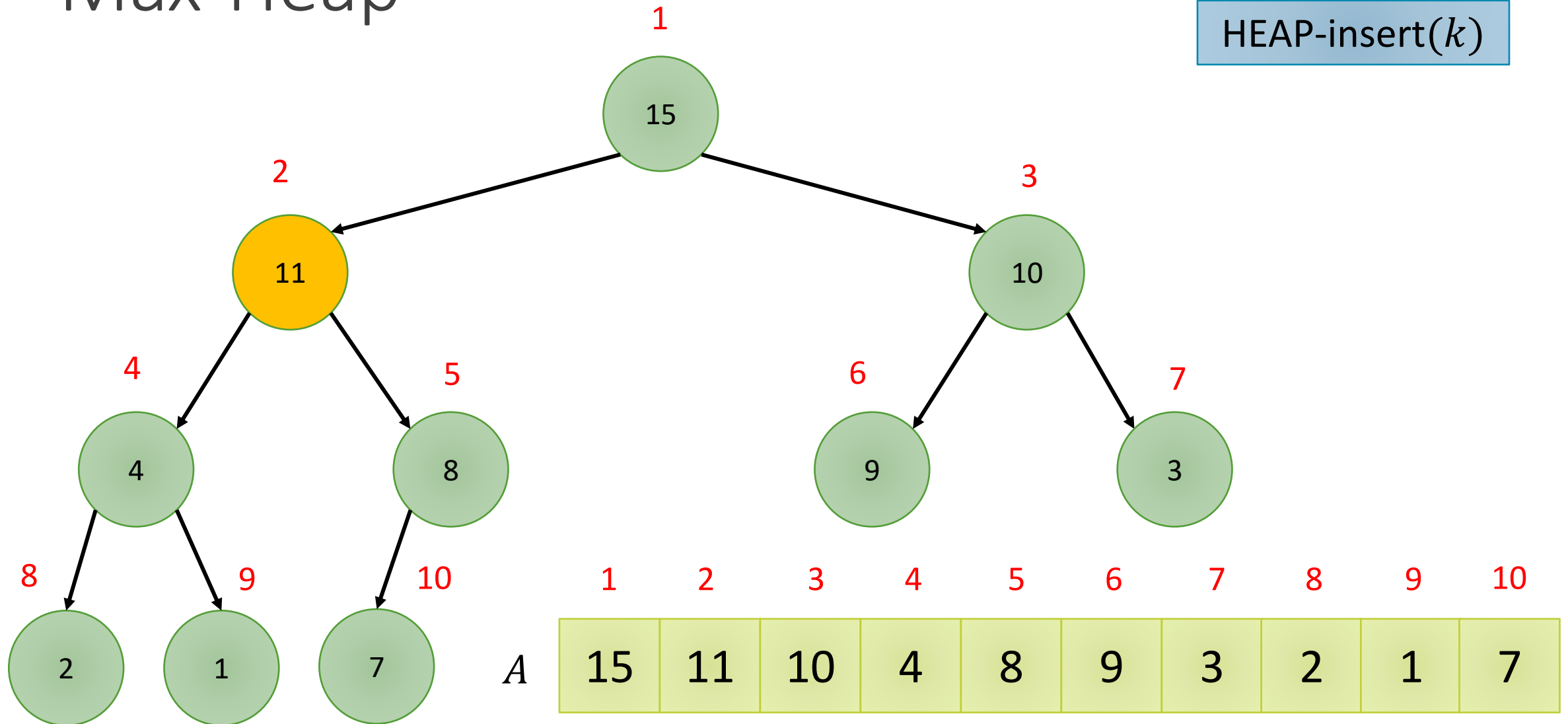
# Max-Heap

HEAP-insert( $k$ )



# Max-Heap

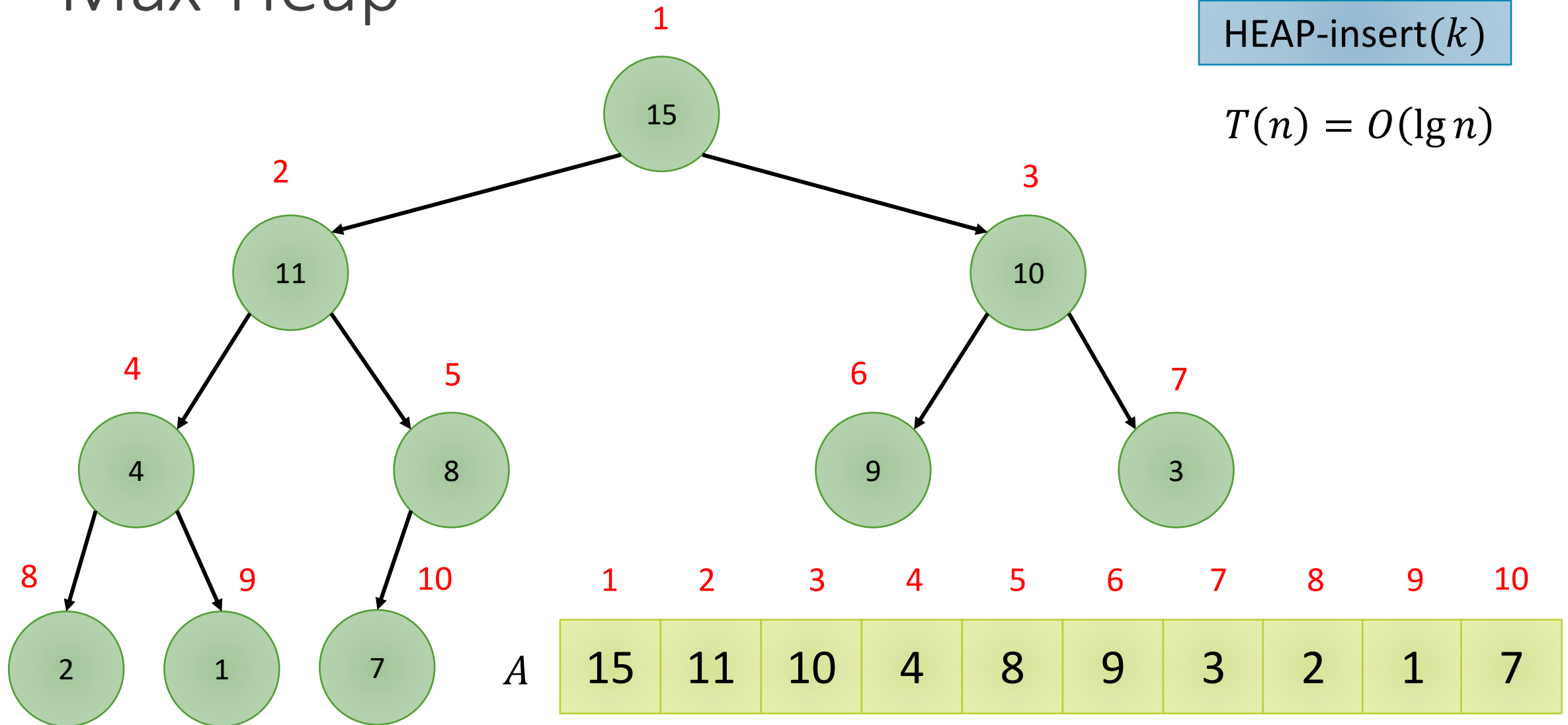
HEAP-insert( $k$ )



# Max-Heap

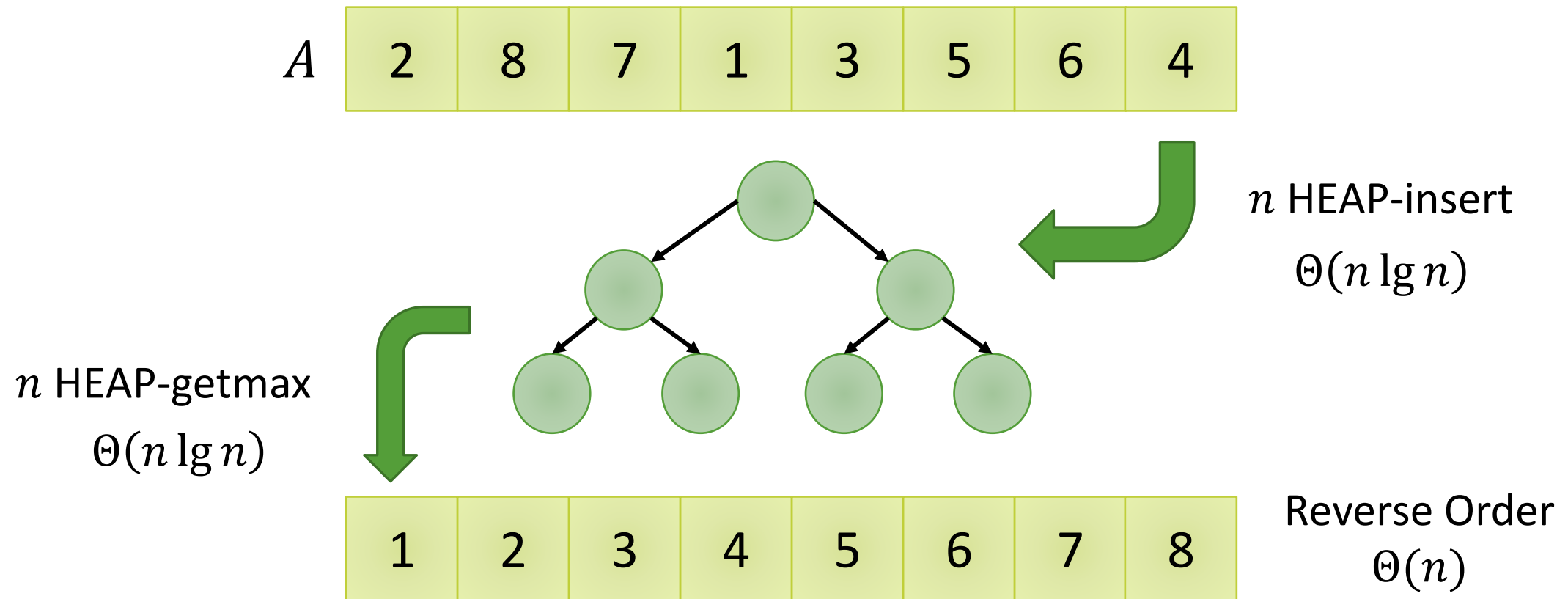
HEAP-insert( $k$ )

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



# Heapsort

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

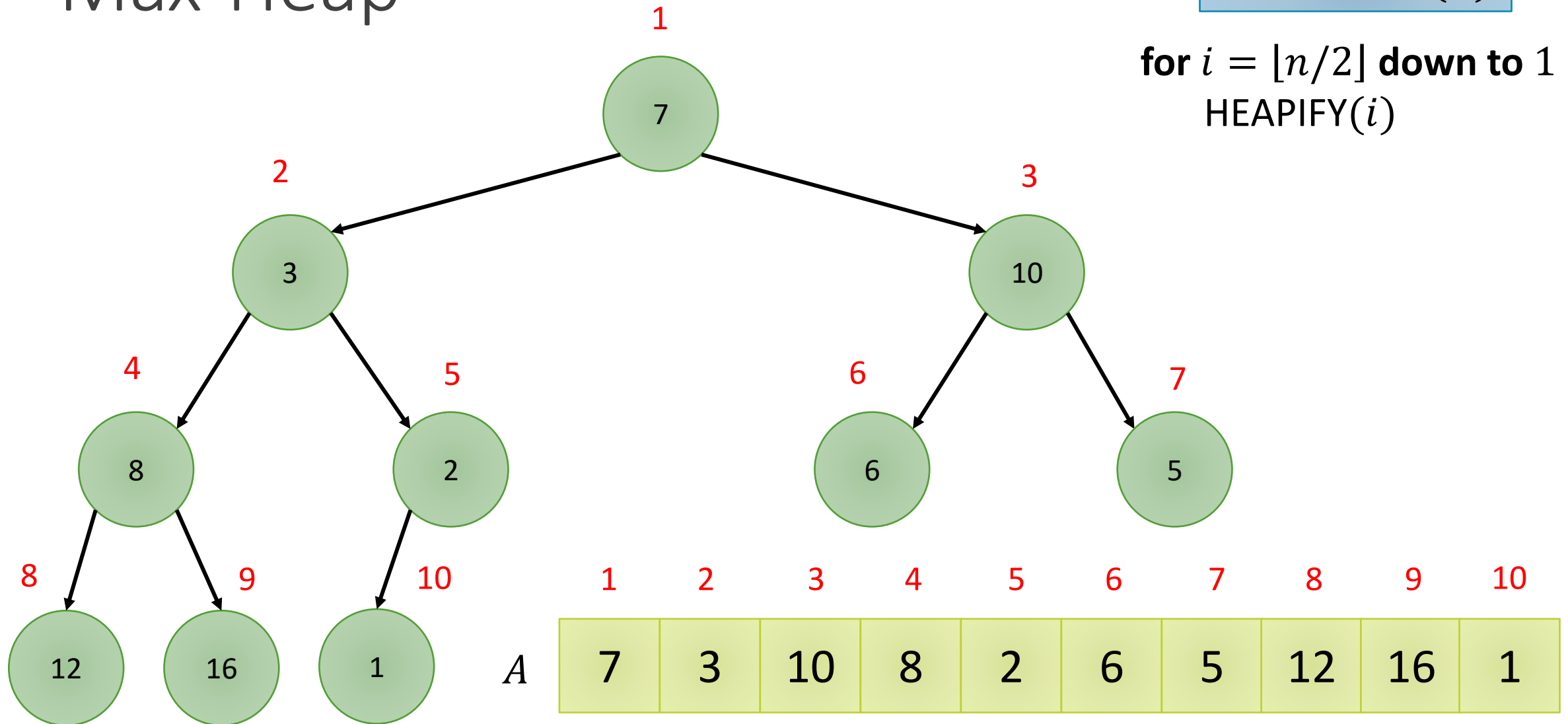




# Max-Heap

BUILD-HEAP( $A$ )

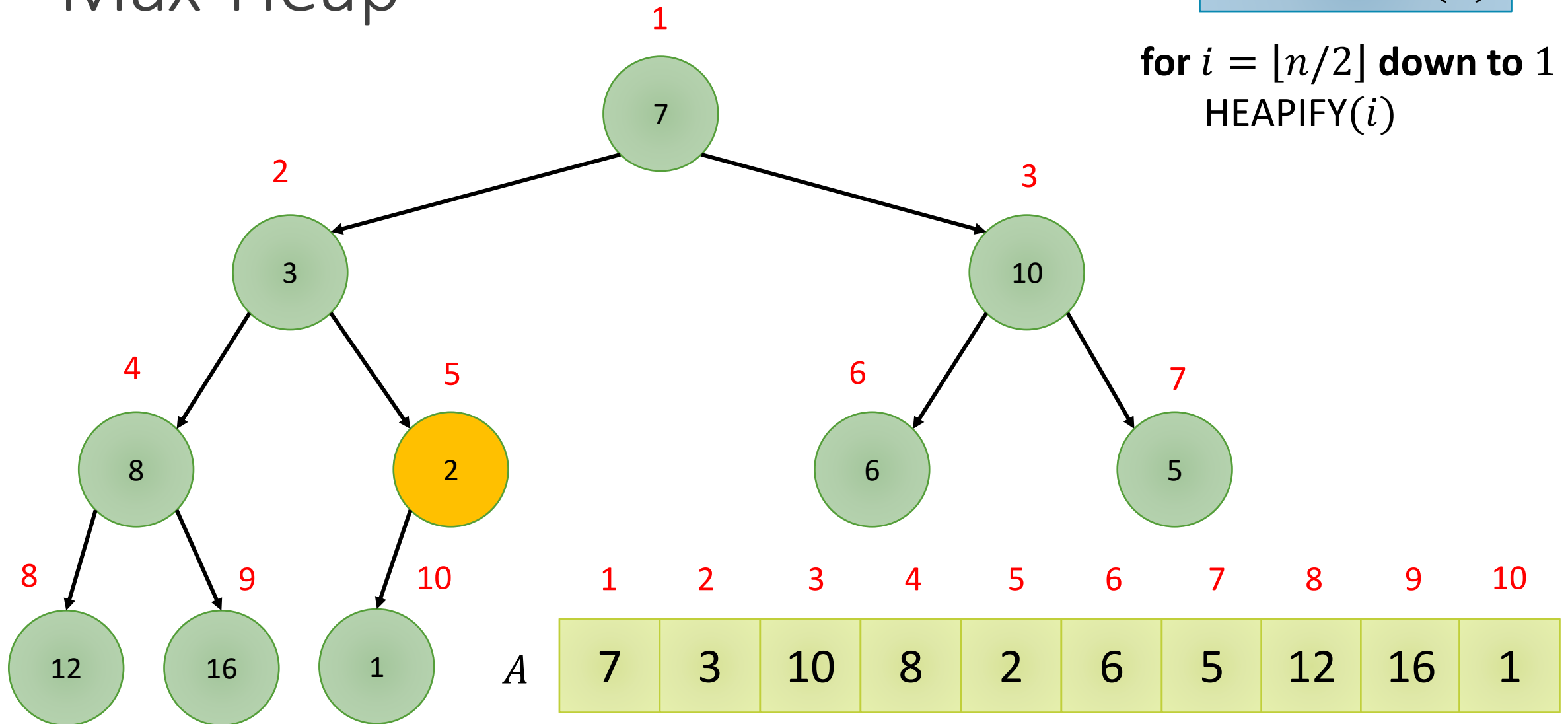
**for  $i = \lfloor n/2 \rfloor$  down to 1**  
**HEAPIFY( $i$ )**



# Max-Heap

BUILD-HEAP( $A$ )

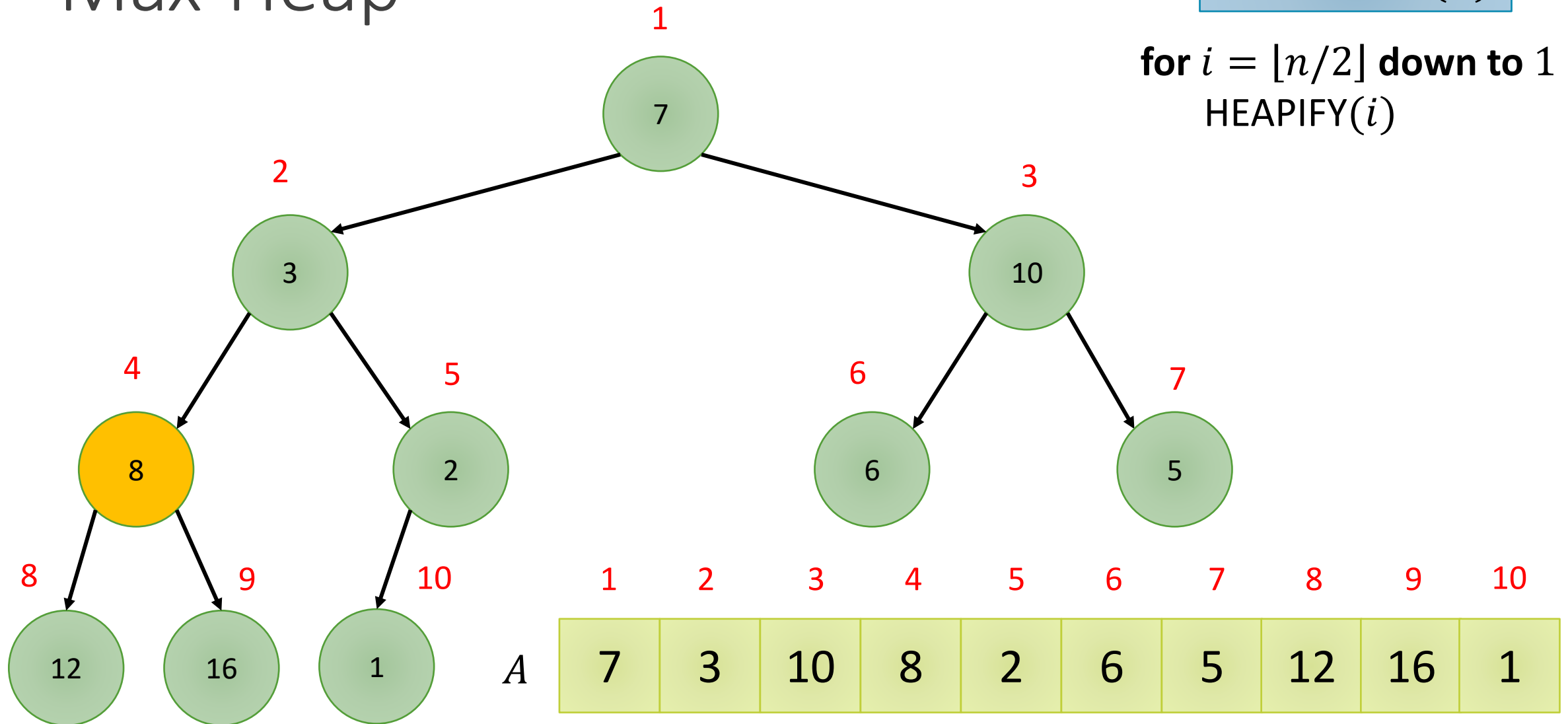
**for  $i = \lfloor n/2 \rfloor$  down to 1**  
**HEAPIFY( $i$ )**



# Max-Heap

BUILD-HEAP(*A*)

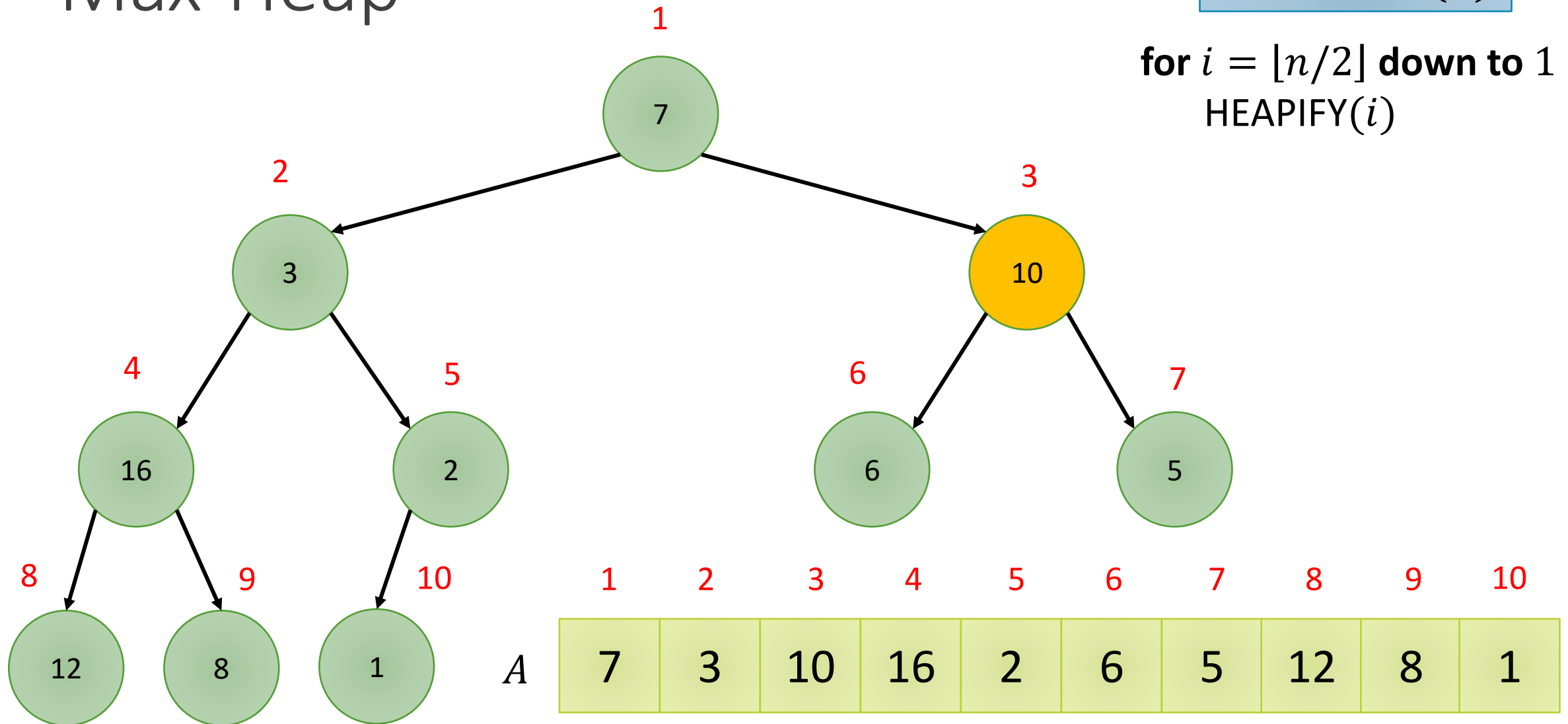
**for**  $i = \lfloor n/2 \rfloor$  **down to** 1  
    HEAPIFY(*i*)



# Max-Heap

BUILD-HEAP( $A$ )

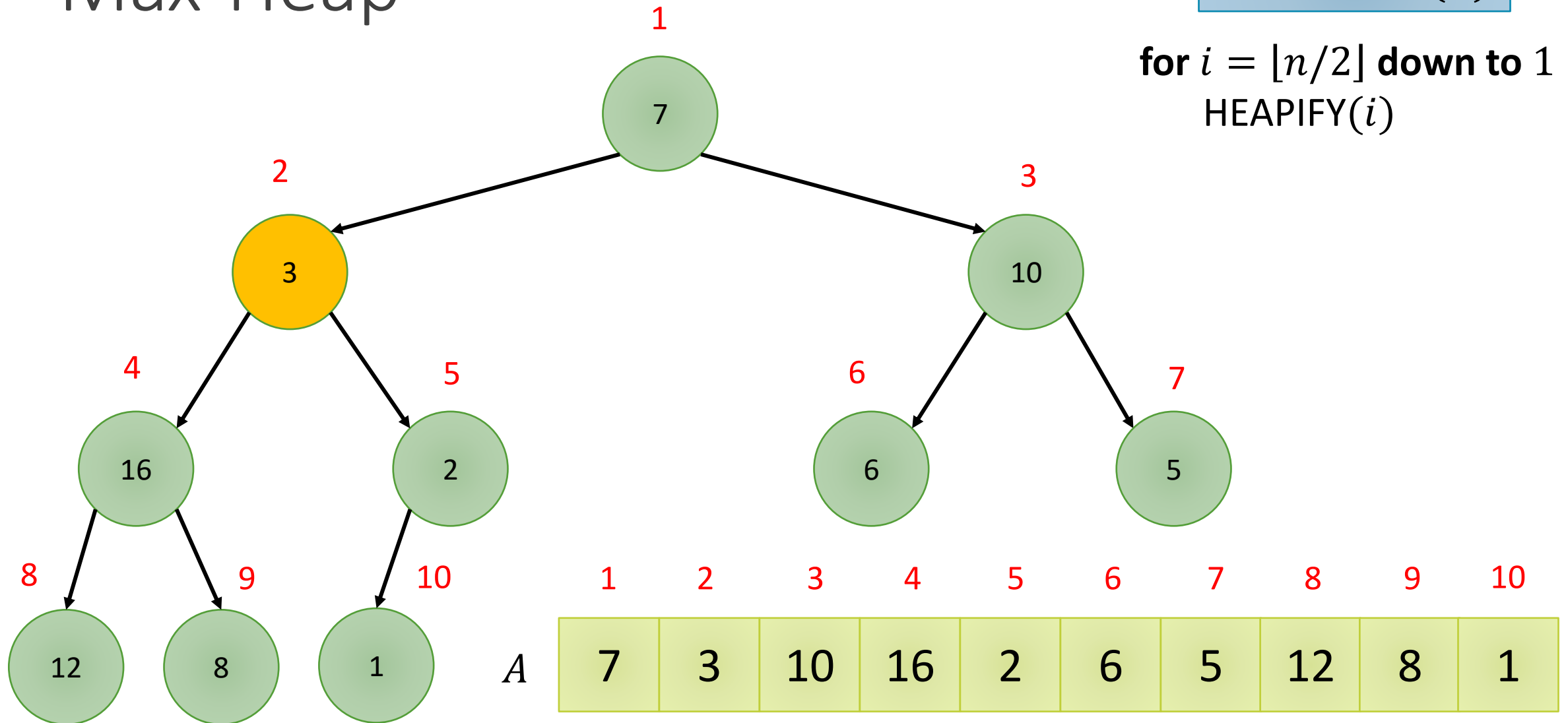
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BUILD-HEAP(*A*)

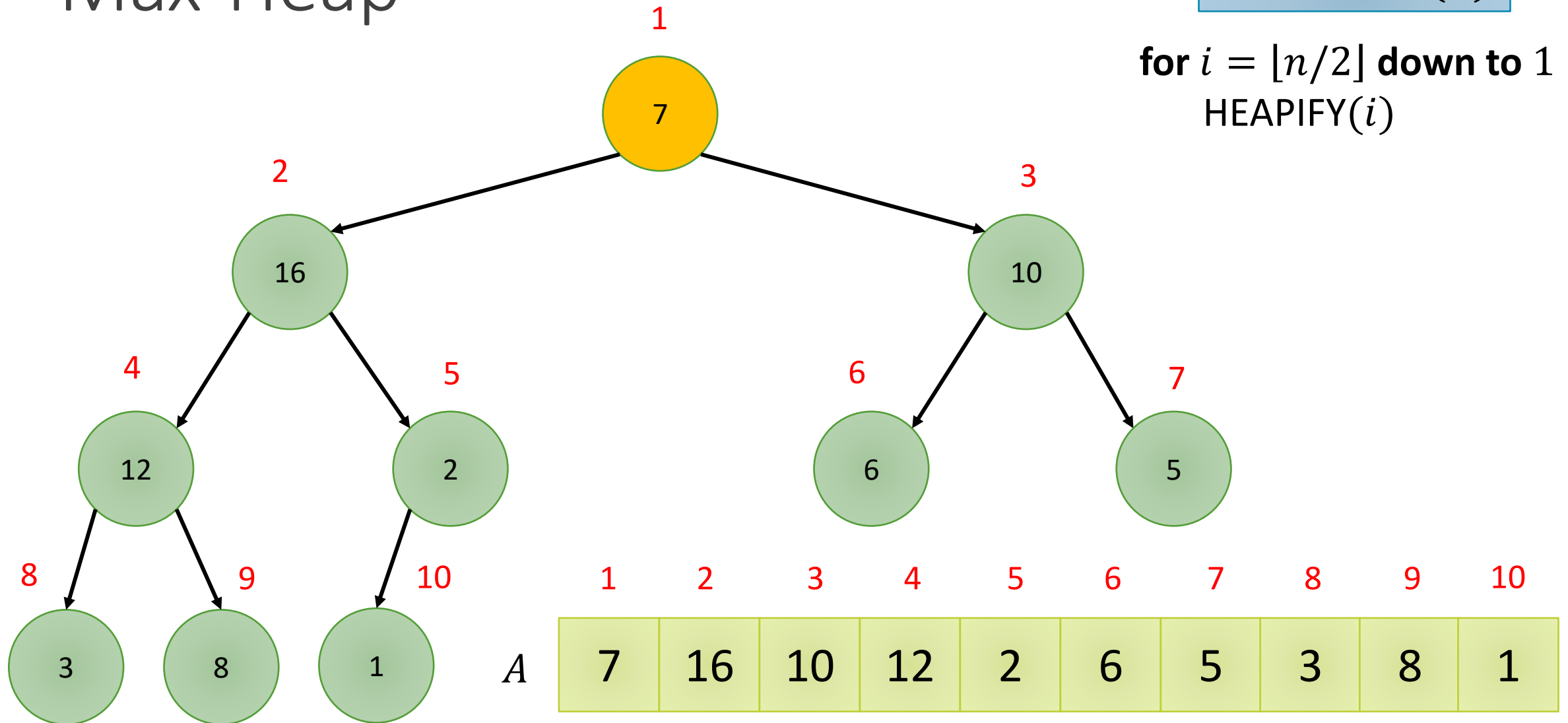
**for**  $i = \lfloor n/2 \rfloor$  **down to** 1  
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# Max-Heap

BUILD-HEAP( $A$ )

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**HEAPIFY( $i$ )**

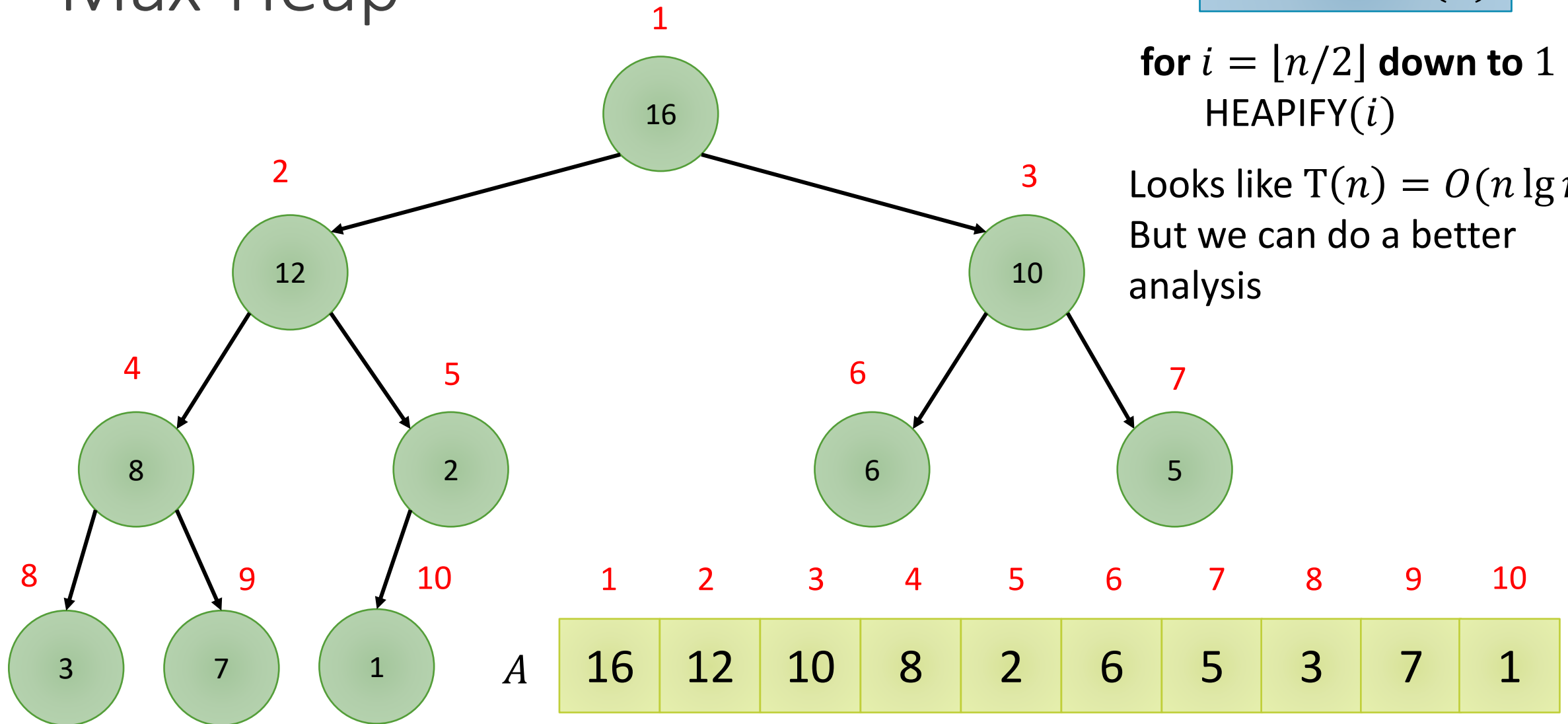


# Max-Heap

BUILD-HEAP(*A*)

**for**  $i = \lfloor n/2 \rfloor$  **down to** 1  
    HEAPIFY(*i*)

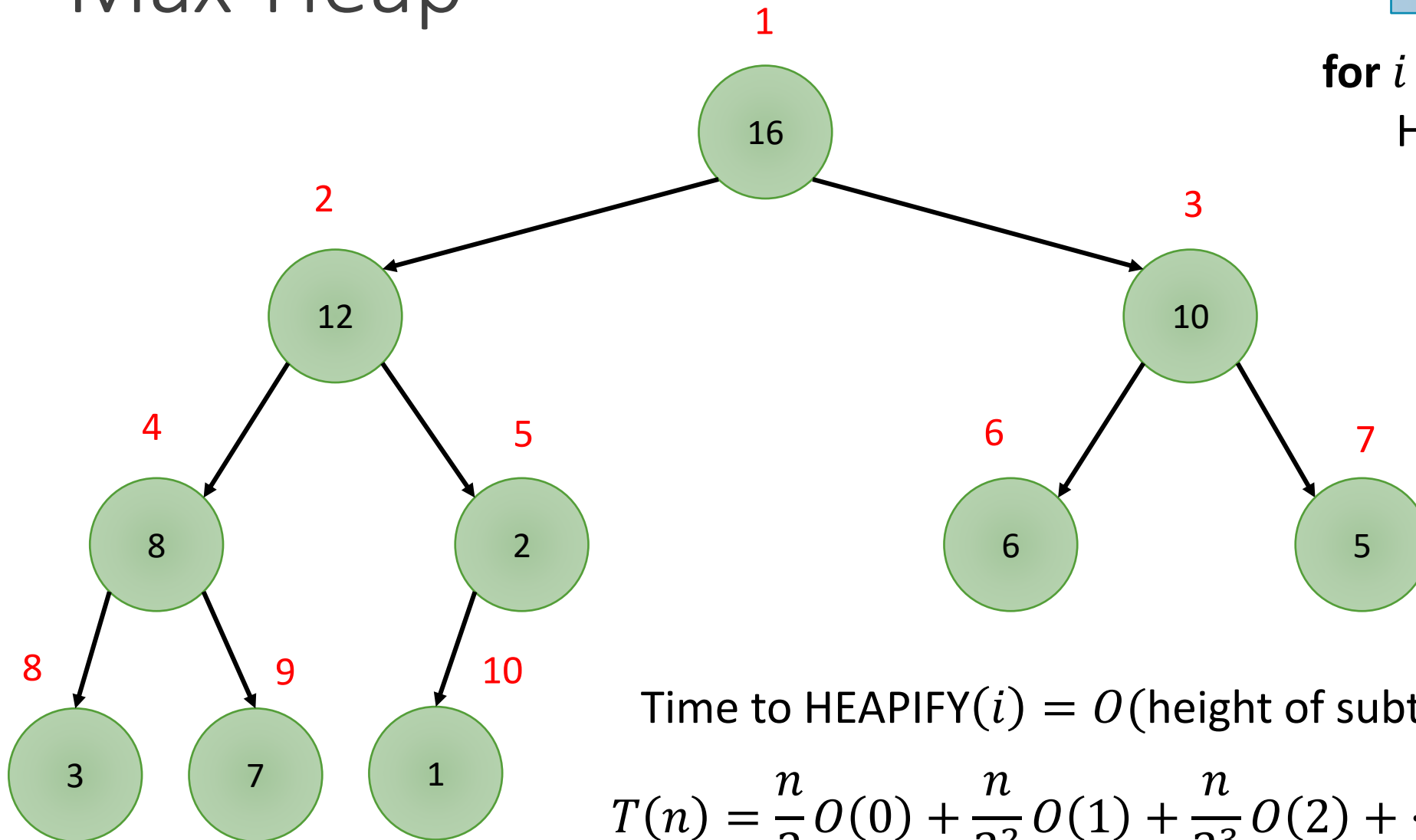
Looks like  $T(n) = O(n \lg n)$   
But we can do a better  
analysis



# Max-Heap

BUILD-HEAP(*A*)

**for**  $i = \lfloor n/2 \rfloor$  **down to** 1  
    HEAPIFY(*i*)



Time to HEAPIFY(*i*) =  $O(\text{height of subtree rooted at } i)$

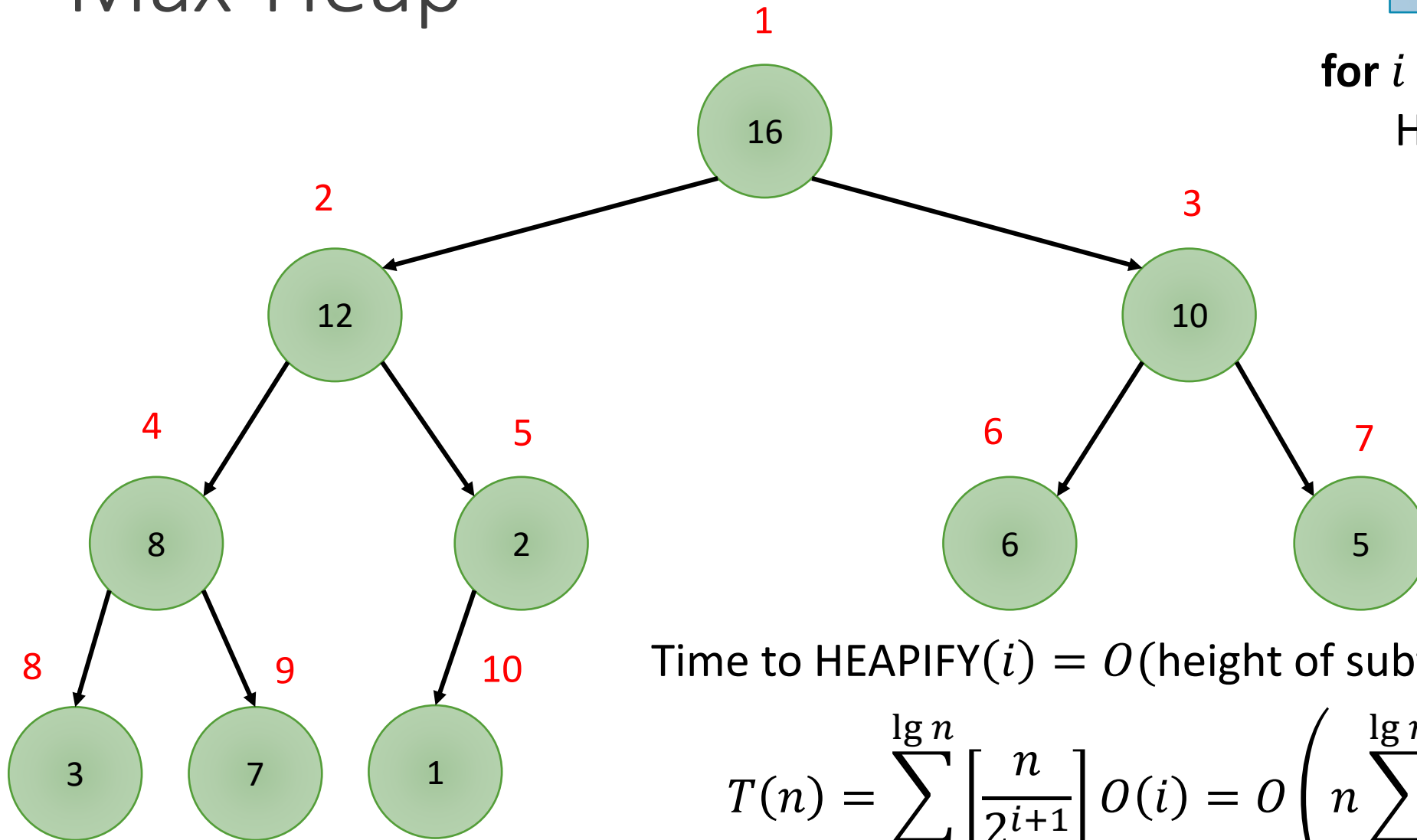
$$T(n) = \frac{n}{2} O(0) + \frac{n}{2^2} O(1) + \frac{n}{2^3} O(2) + \cdots + \frac{n}{2^{\lg n + 1}} O(\lg n)$$



# Max-Heap

BUILD-HEAP(*A*)

**for**  $i = \lfloor n/2 \rfloor$  **down to** 1  
    HEAPIFY(*i*)



Time to HEAPIFY(*i*) =  $O(\text{height of subtree rooted at } i)$

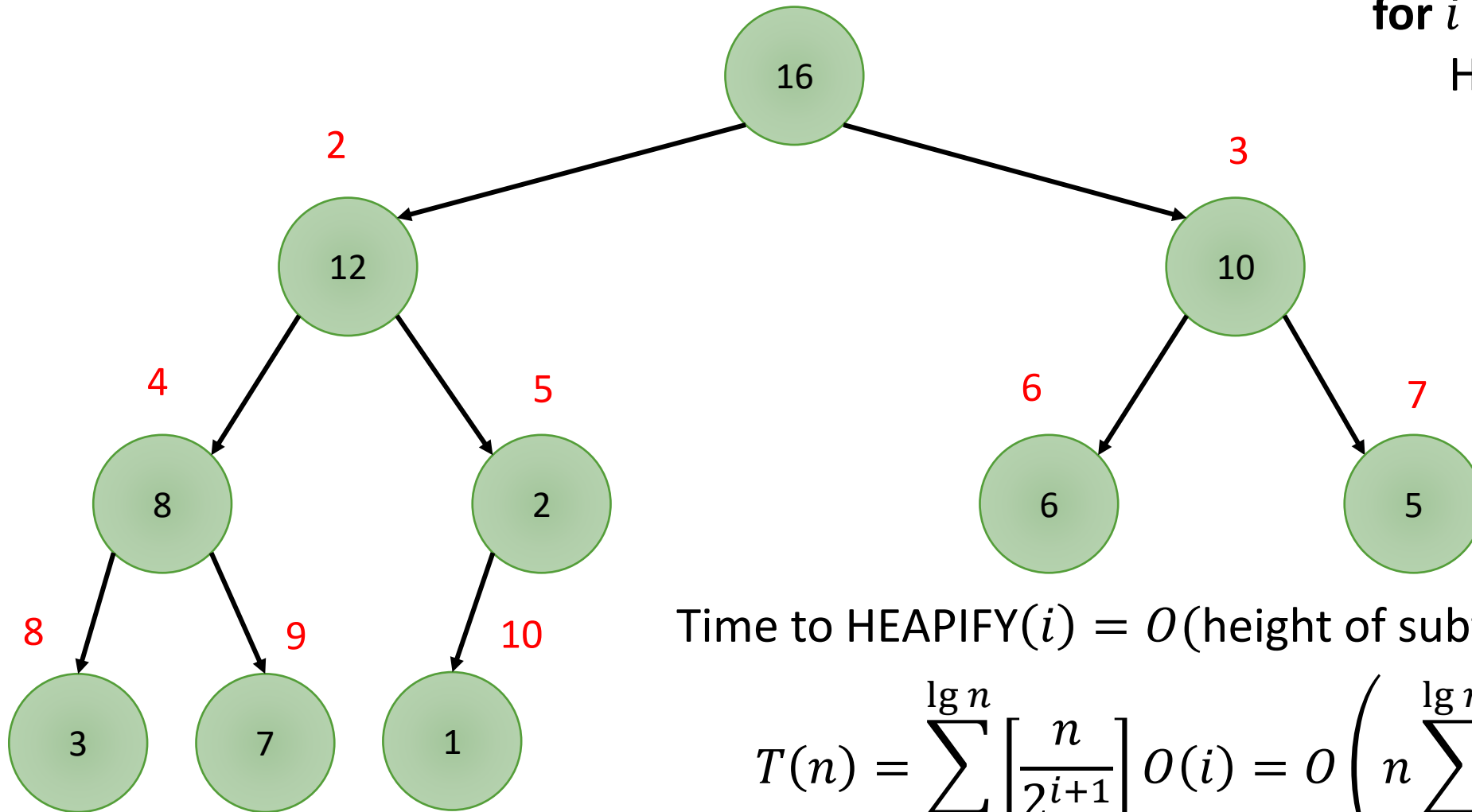
$$T(n) = \sum_{i=0}^{\lg n} \left\lfloor \frac{n}{2^{i+1}} \right\rfloor O(i) = O\left(n \sum_{i=0}^{\lg n} \left\lfloor \frac{i}{2^i} \right\rfloor\right)$$

# Max-Heap

$$\sum_{k=1}^{\infty} kx^k \leq \frac{x}{(1-x)^2} \text{ for } |x| < 1$$

BUILD-HEAP(*A*)

**for**  $i = \lfloor n/2 \rfloor$  **down to** 1  
    HEAPIFY(*i*)



Time to HEAPIFY(*i*) =  $O(\text{height of subtree rooted at } i)$

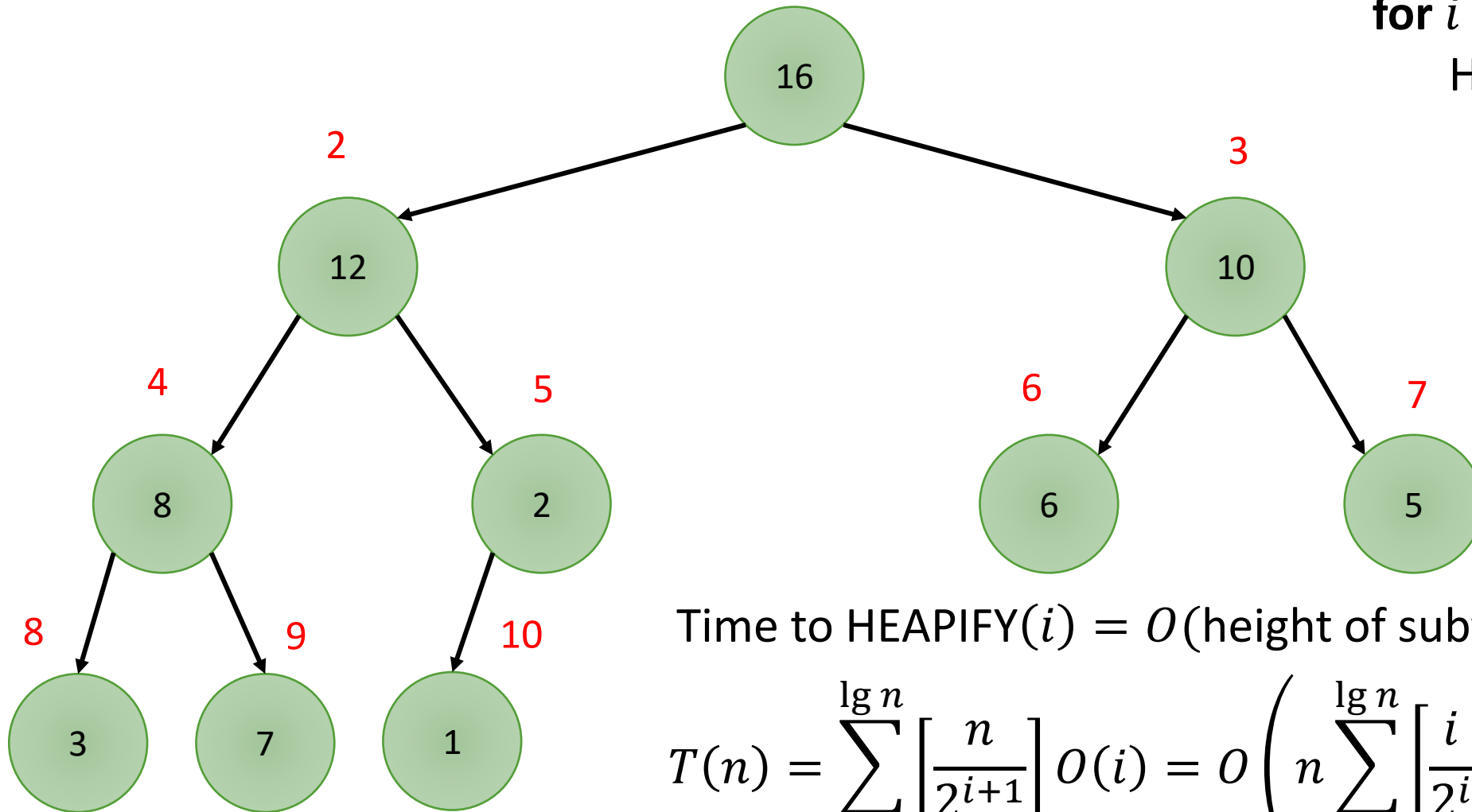
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# Max-Heap

$$\sum_{k=1}^{\infty} kx^k \leq \frac{x}{(1-x)^2} \text{ for } |x| < 1$$

BUILD-HEAP(*A*)

**for**  $i = \lfloor n/2 \rfloor$  **down to** 1  
    HEAPIFY(*i*)



Time to HEAPIFY(*i*) =  $O(\text{height of subtree rooted at } i)$

$$T(n) = \sum_{i=0}^{\lg n} \left\lfloor \frac{n}{2^{i+1}} \right\rfloor O(i) = O\left(n \sum_{i=0}^{\lg n} \left\lfloor \frac{i}{2^i} \right\rfloor\right) = O(n)$$

# Heapsort

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

