

Faculty of Information Technology, Monash University

COMMONWEALTH OF AUSTRALIA

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FIT2004, S2/2016

Week 3: Sorting Algorithms

Lecturer: Muhammad Aamir Cheema

ACKNOWLEDGMENTS

The slides are based on the material developed by [Arun Konagurthu](#) and [Lloyd Allison](#).

Overview

- In the previous week's lectures, we focused on precise reasoning to solve problems using **loop invariants**.
- We have also analysed them, proved them etc.
- In this lecture, we will discuss and analyze more efficient $O(N \log N)$ time sorting algorithms.
- Our discussion will cover
 - Stability of sorting and In-place algorithms
 - Heap data-structure and Heap sort
 - Merge sort
 - Quicksort

Recommended reading

- Priority Queue and Heap data structure reference:
<http://www.csse.monash.edu.au/~lloyd/tildeAlgDS/Priority-Q/>
- Merge sort:
<http://www.csse.monash.edu.au/~lloyd/tildeAlgDS/Sort/Merge/>
- Dutch National Flag partitioning problem:
<http://www.csse.monash.edu.au/~lloyd/tildeAlgDS/Sort/Flag/>
- Quick sort:
<http://www.csse.monash.edu.au/~lloyd/tildeAlgDS/Sort/Quick/>
- Weiss, Data structures and Algorithm analysis in Java, Chapters 6 & 7

Stable sorting algorithms

A sorting algorithm is called stable if it maintains the relative ordering of elements that have equal keys.

Input is sorted by names

Input

Marks	80	75	70	90	85	75
Name	Alice	Bill	John	Geoff	Leo	Maria

Sort on Marks using a stable algorithm



Output

Marks	70	75	75	80	85	90
Name	John	Bill	Maria	Alice	Leo	Geoff

Note: Output is sorted on marks then names.

Unstable sorting cannot guarantee this (e.g., Maria may appear before Bill)

Selection sort is unstable. Insertion sort is stable!

Priority Queue

- Priority Queue is an Abstract Data Type usually implemented with a heap
- The operations of a Priority Queue are:
 - **create** an empty Priority Queue
 - **insert** an element having a certain priority to the Priority Queue
 - **remove** the element having the highest priority.

Do not confuse Priority Queue with ordinary Queue

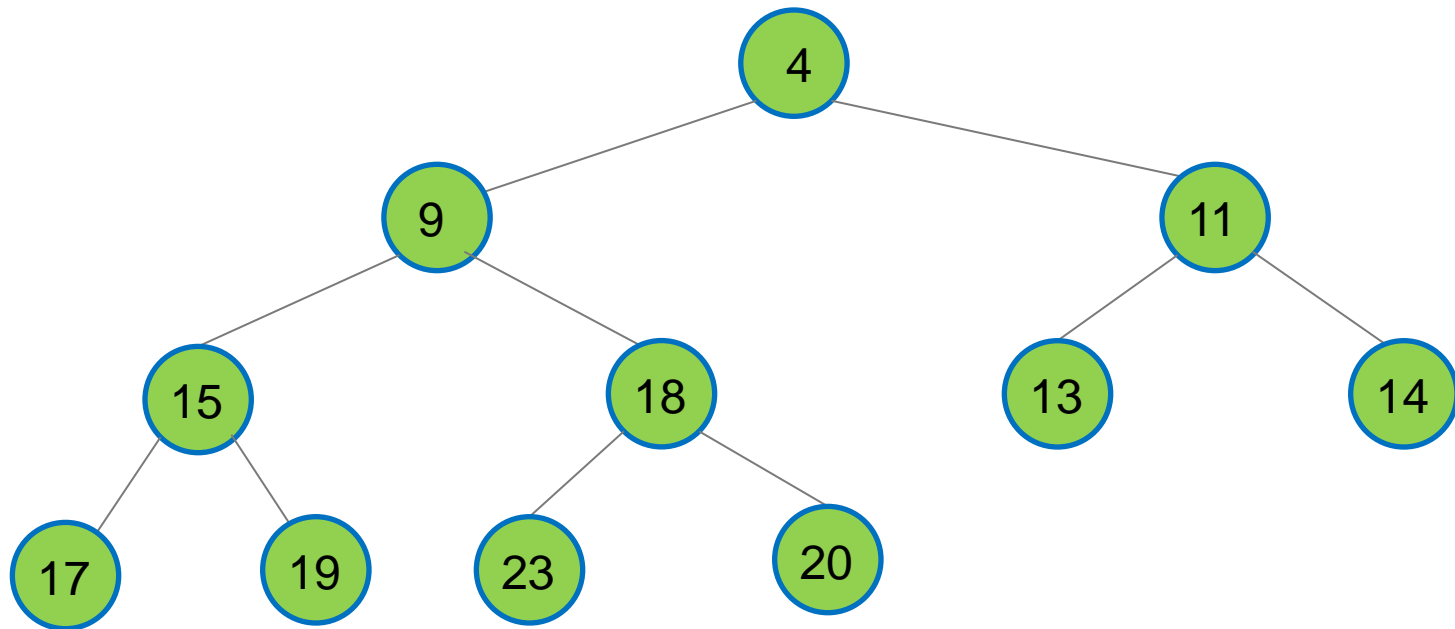
Heap data structure

- Heap is an implementation of a Priority Queue
- Heap is so common for Priority Queue implementation that, in context of Priority Queue, when the term **Heap** is used, it generally means an implementation of **Priority Queue**.
- In this unit, we may use Heap and Priority Queue interchangeably

Properties of Heap

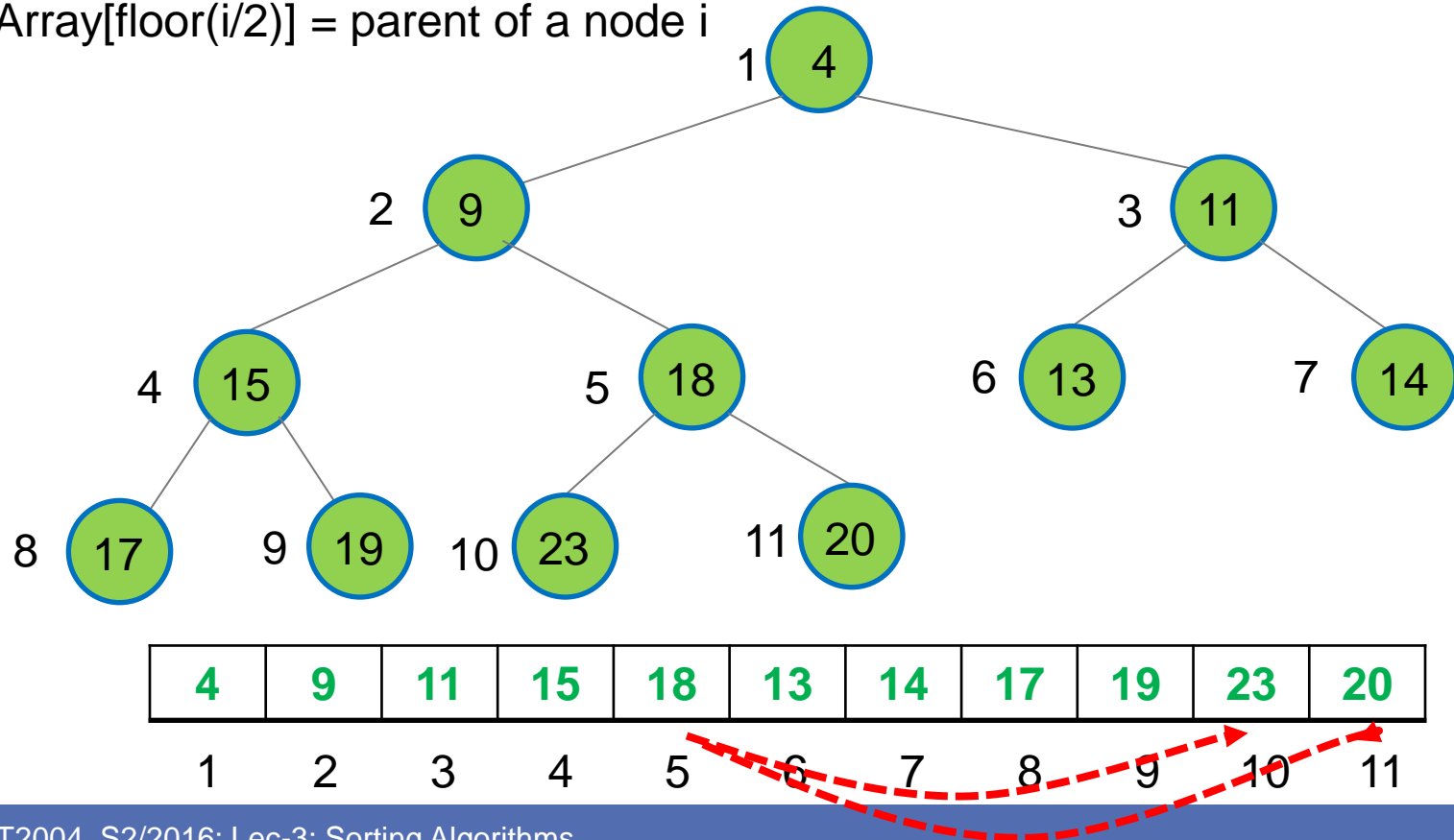
Heap is an implementation of the Priority Queue

- Heap is a **balanced** binary tree
- A parent is always smaller than or equal to its children (this implies that the root is the smallest element in the heap)



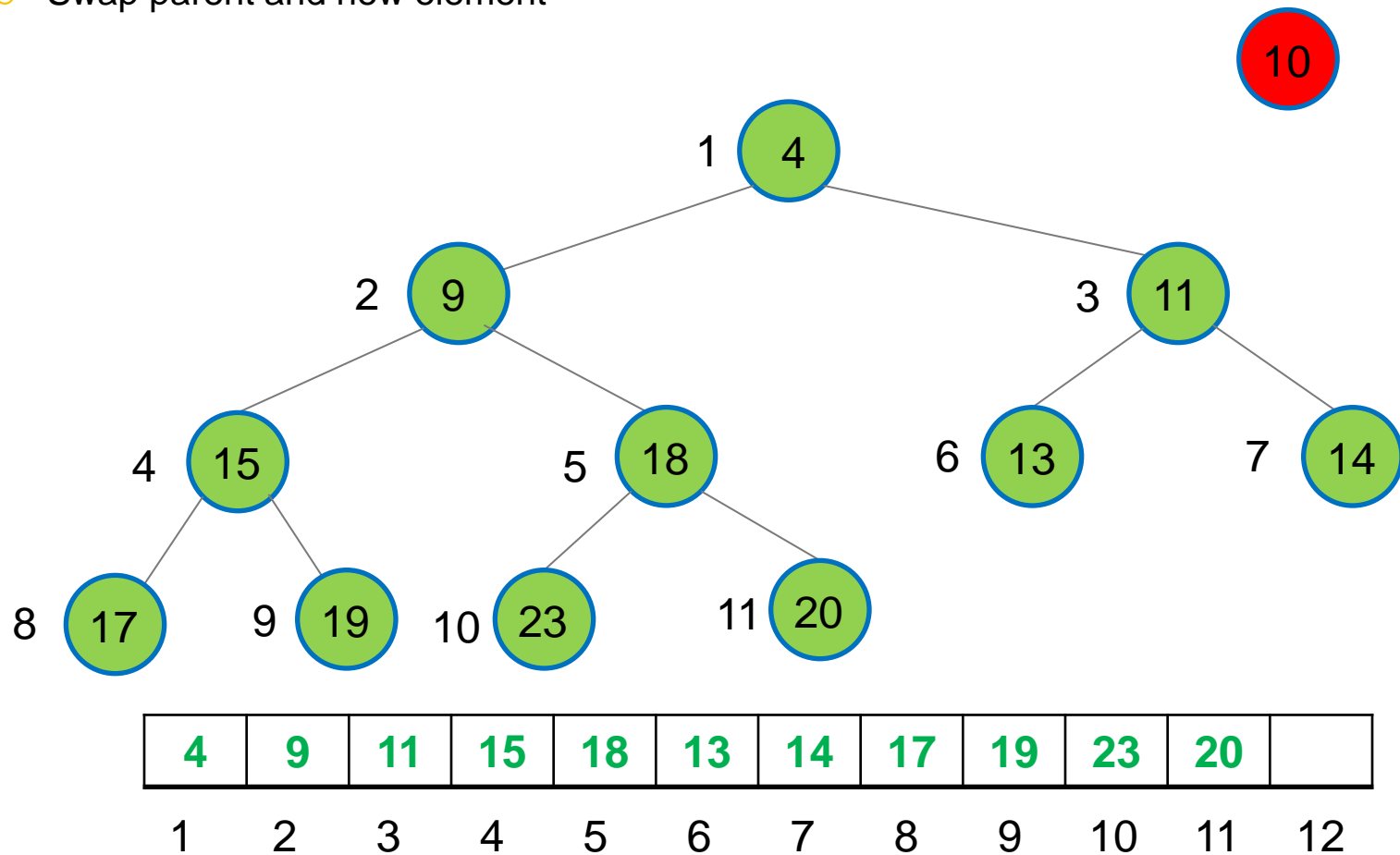
Heap can be represented as an array

- $\text{Array}[1]$ = root of the heap
- $\text{Array}[i]$ = an arbitrary node i
- $\text{Array}[2i]$ = left child of node i
- $\text{Array}[2i + 1]$ = right child of node i
- $\text{Array}[\text{floor}(i/2)]$ = parent of a node i



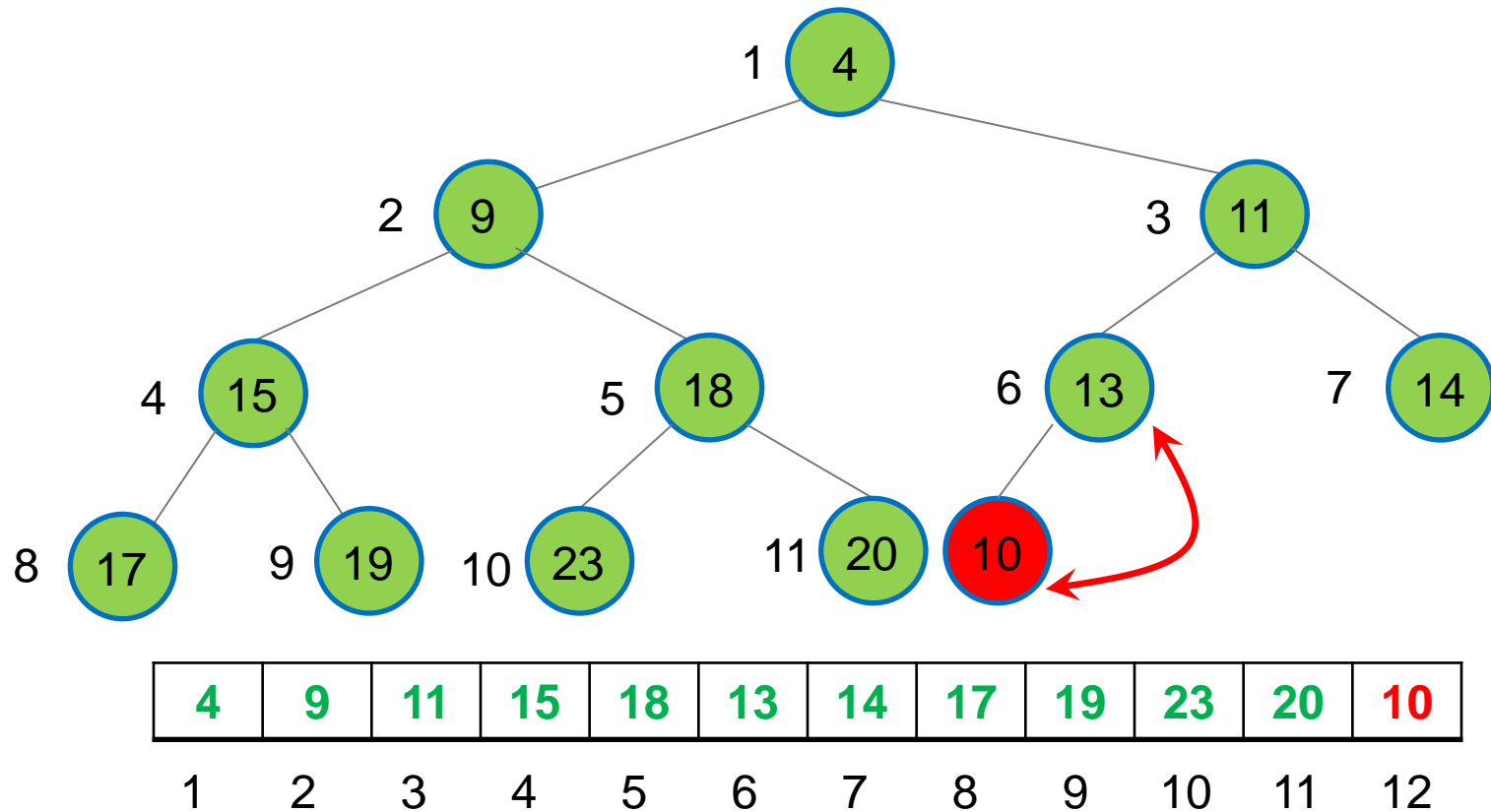
Insertion in Heap (up-Heap)

- Insert new element at Array[N+1]
- While parent(new) > new
 - Swap parent and new element



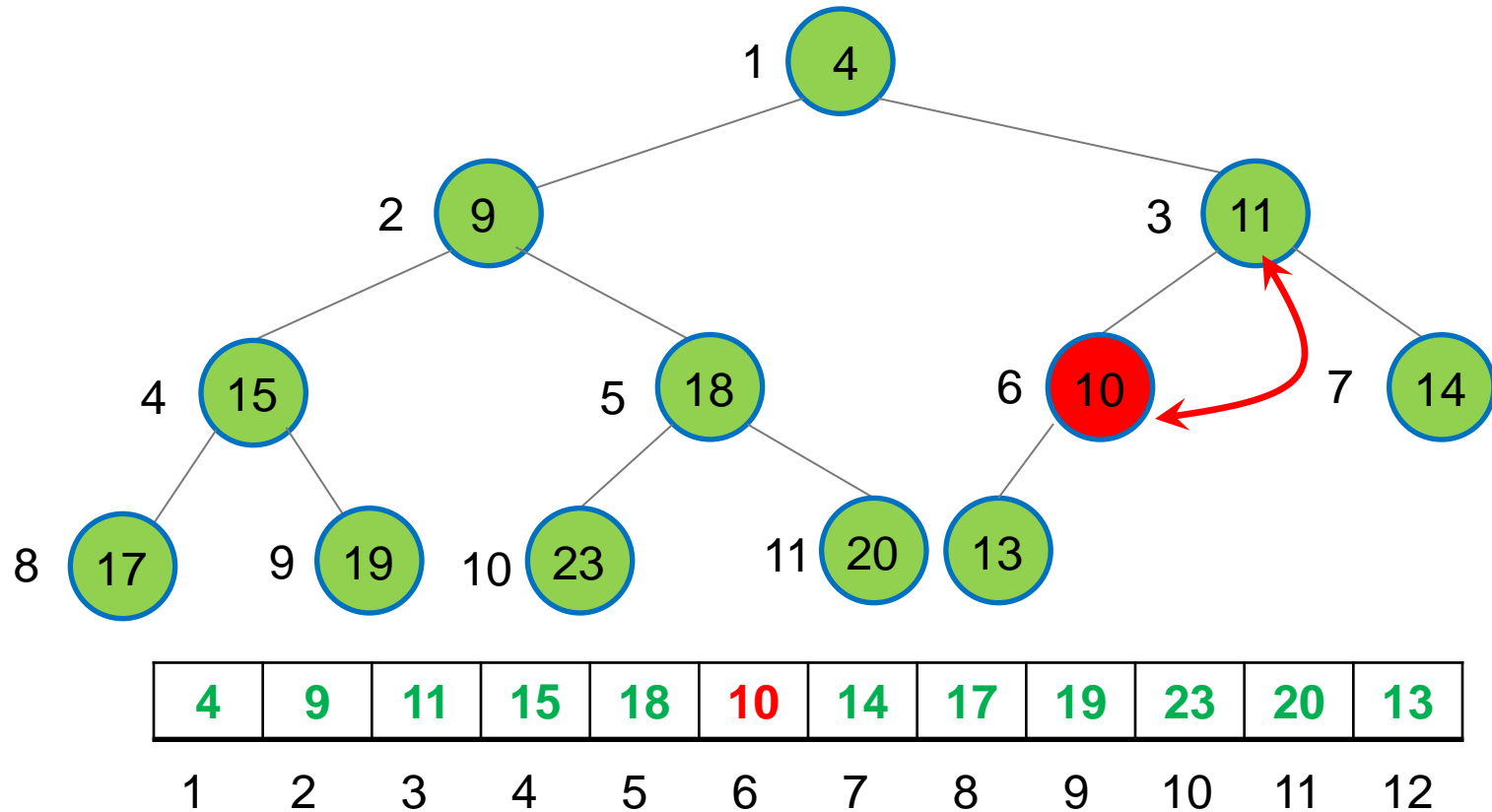
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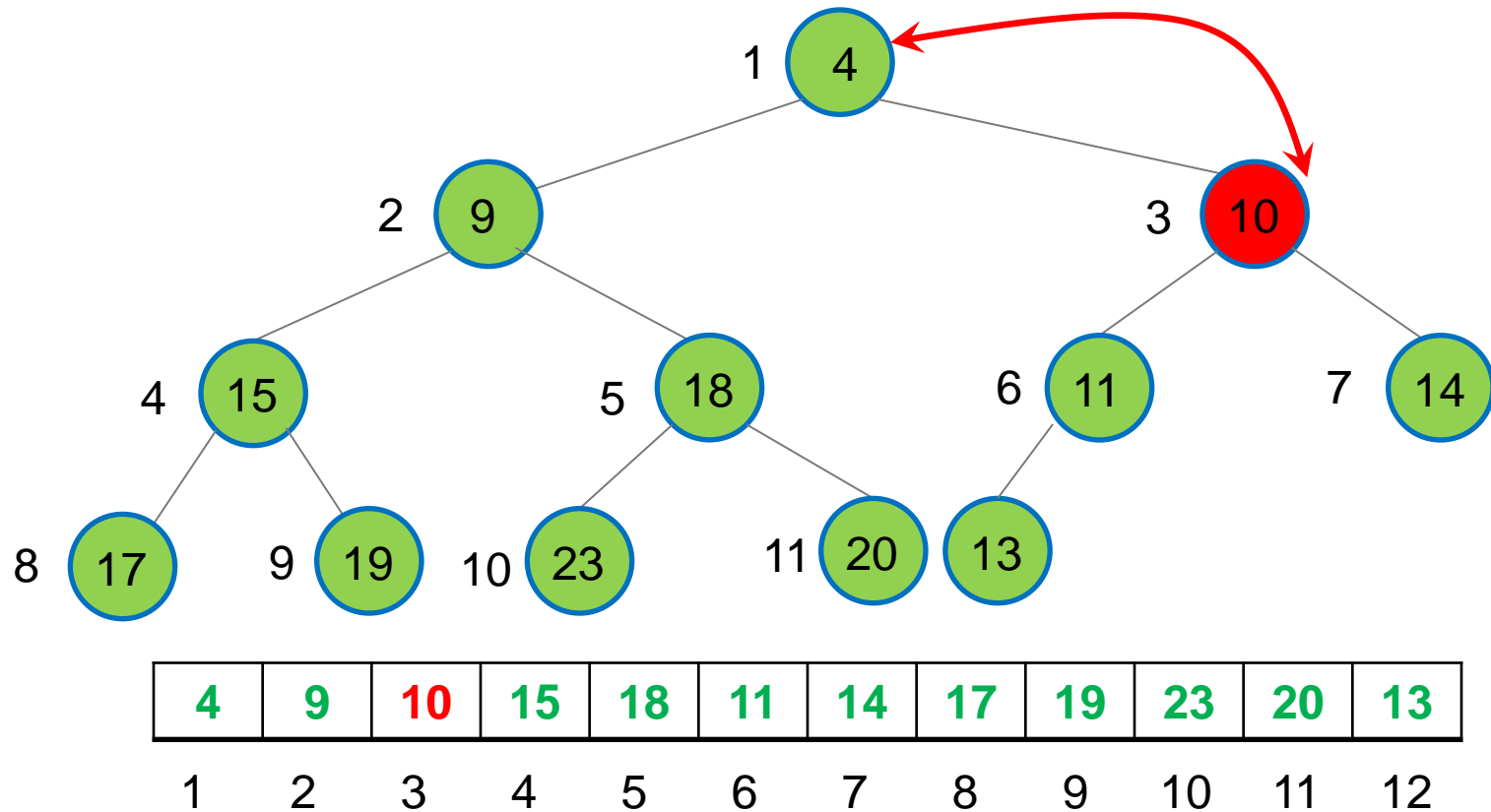
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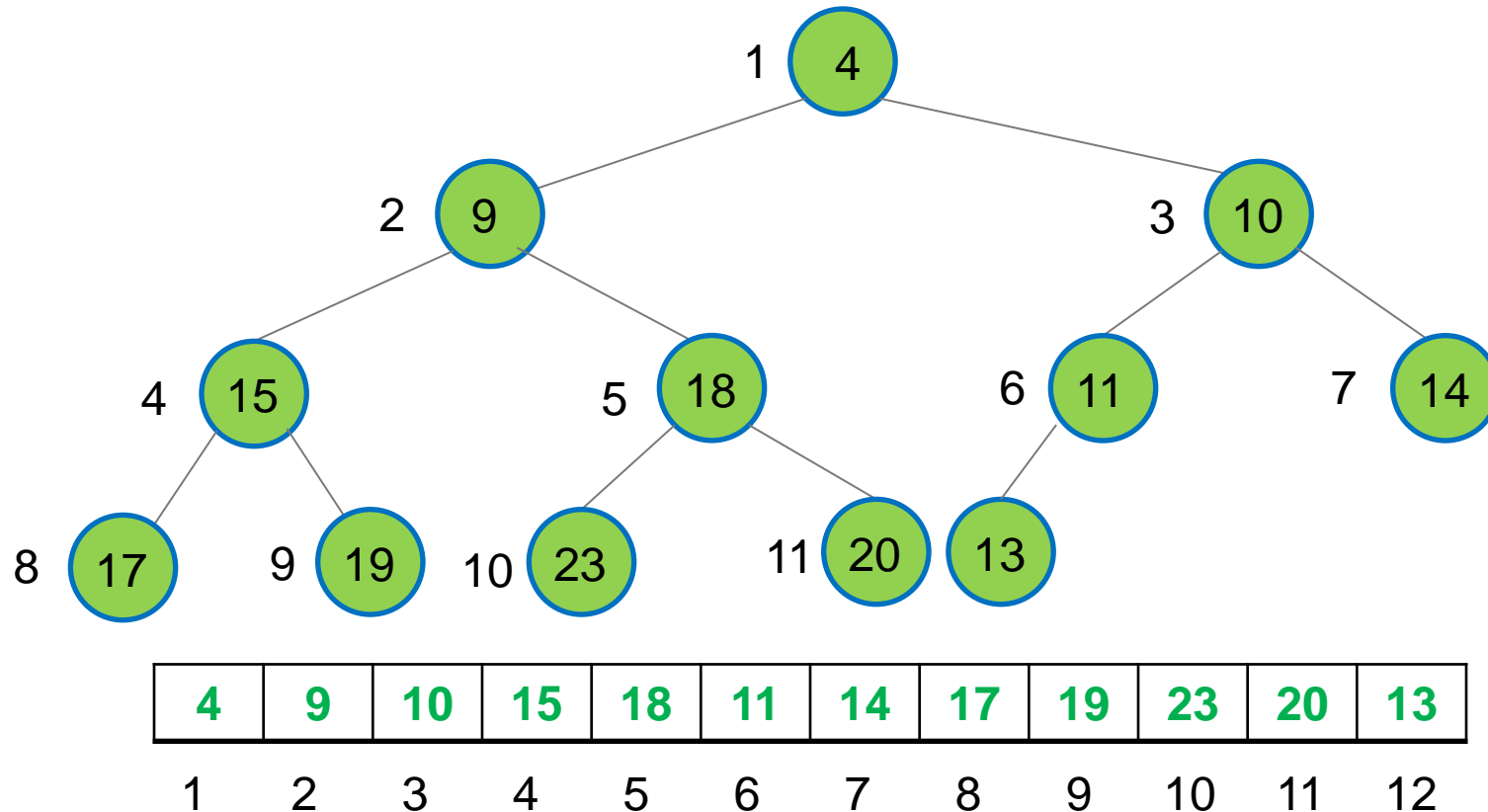
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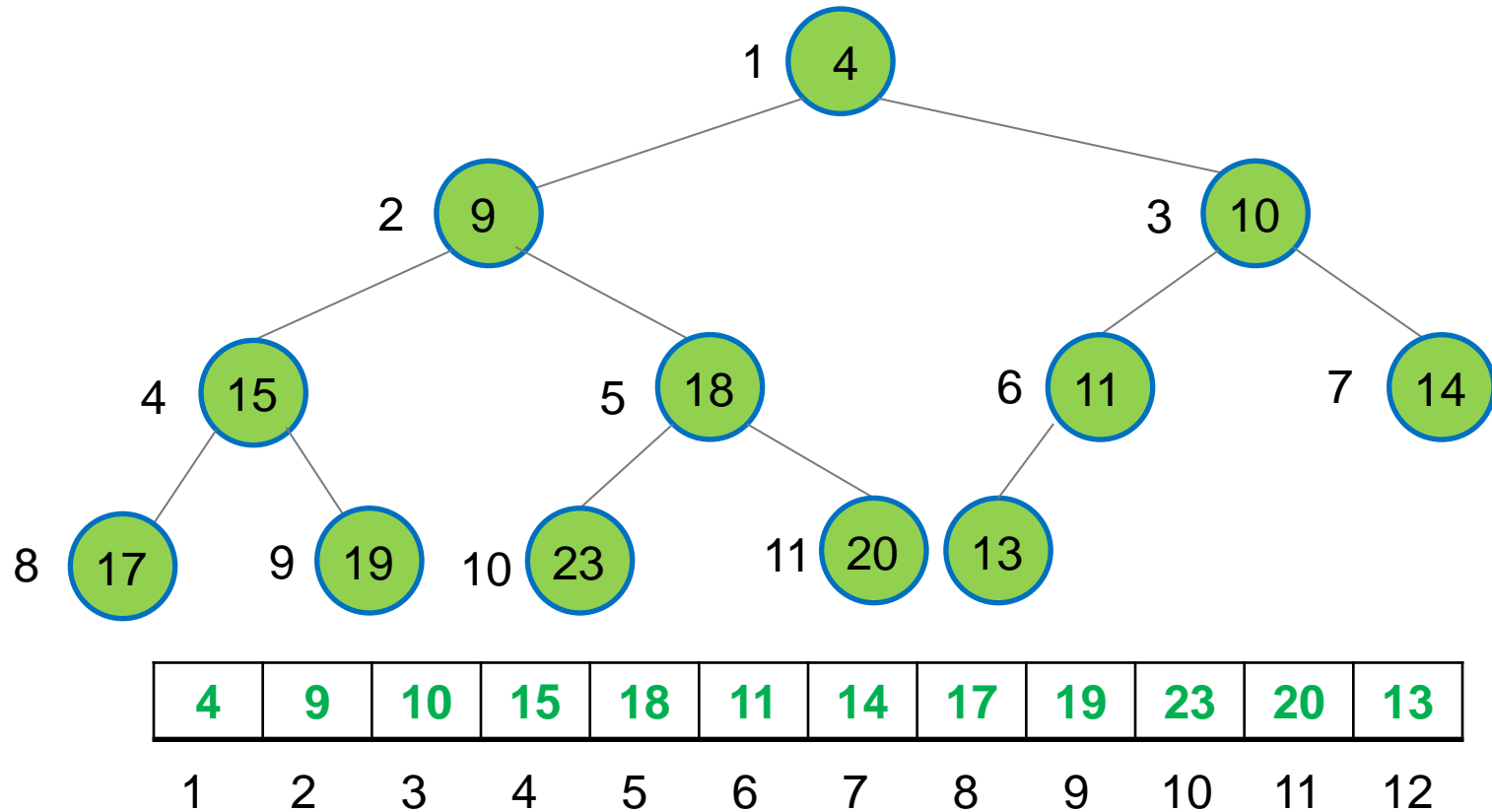
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Insertion in Heap (up-Heap)

- Insert new element at Array[N+1]
- While parent(new) > new **and new is not the root node**
 - Swap parent and new element

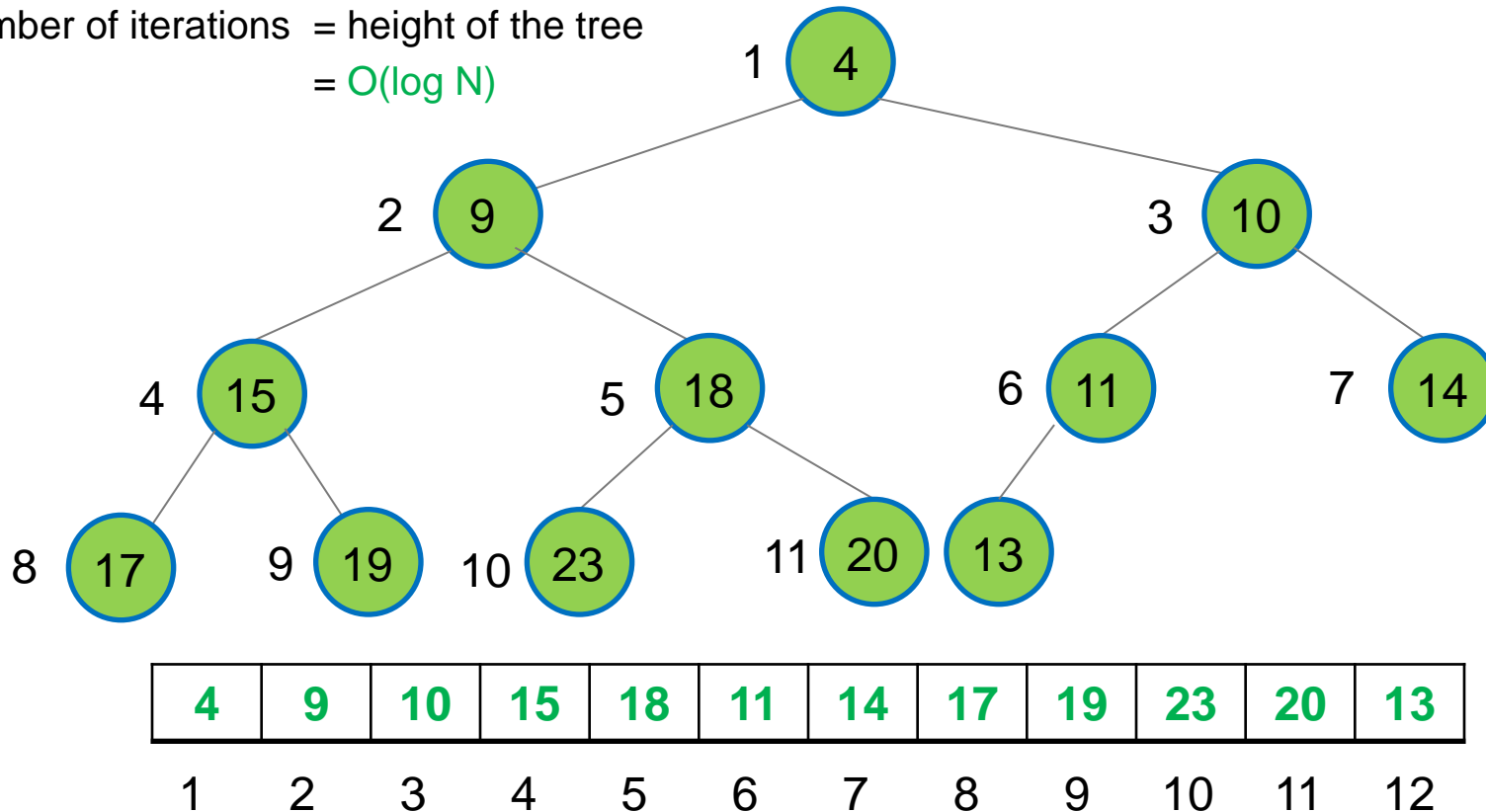


Complexity of up-heap

- Insert new element at `Array[N+1]`
- While `parent(new) > new` and new is not the root node
 - Swap parent and new element

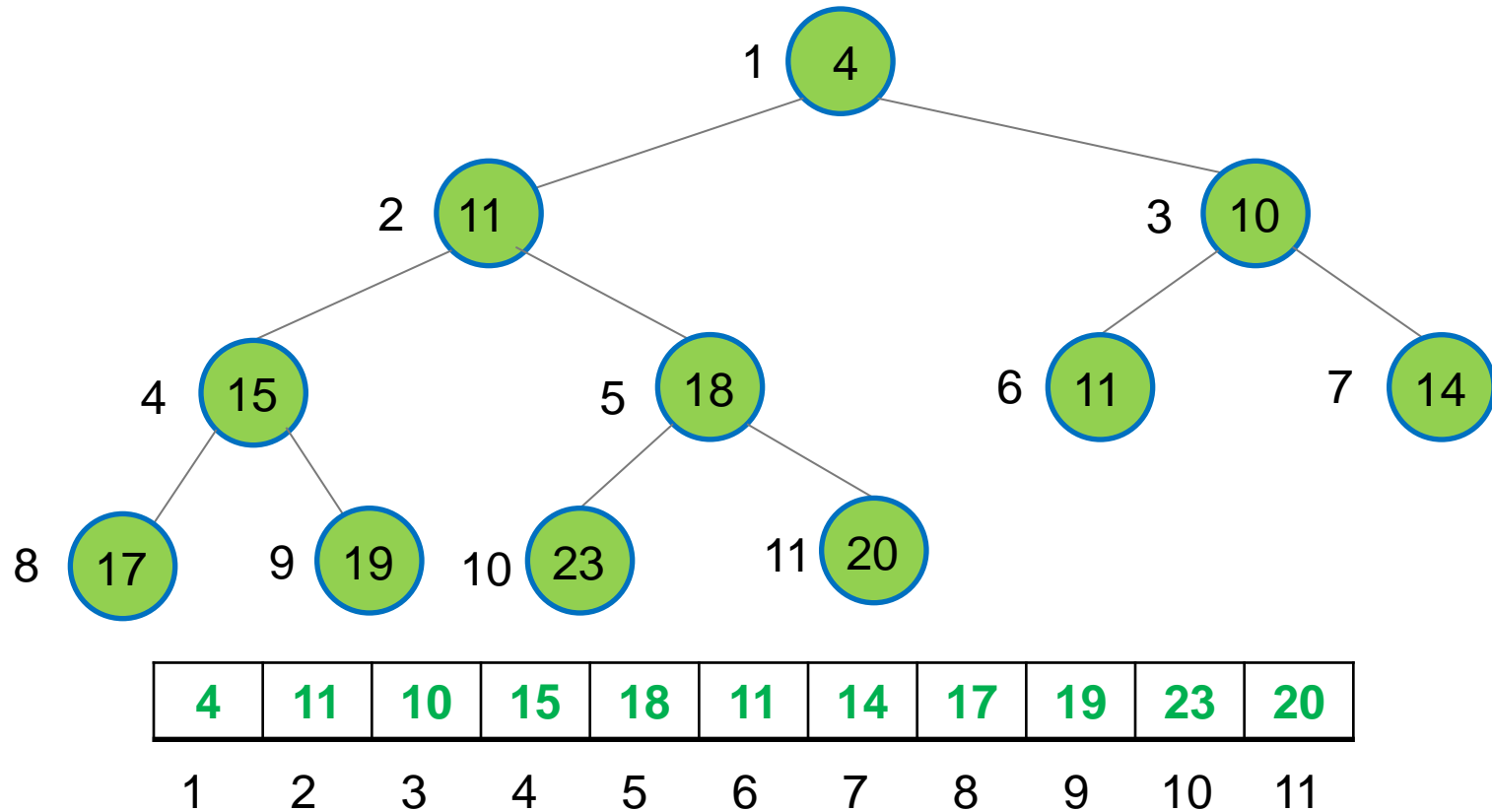
Worst-case time complexity:

Number of iterations = height of the tree
= $O(\log N)$



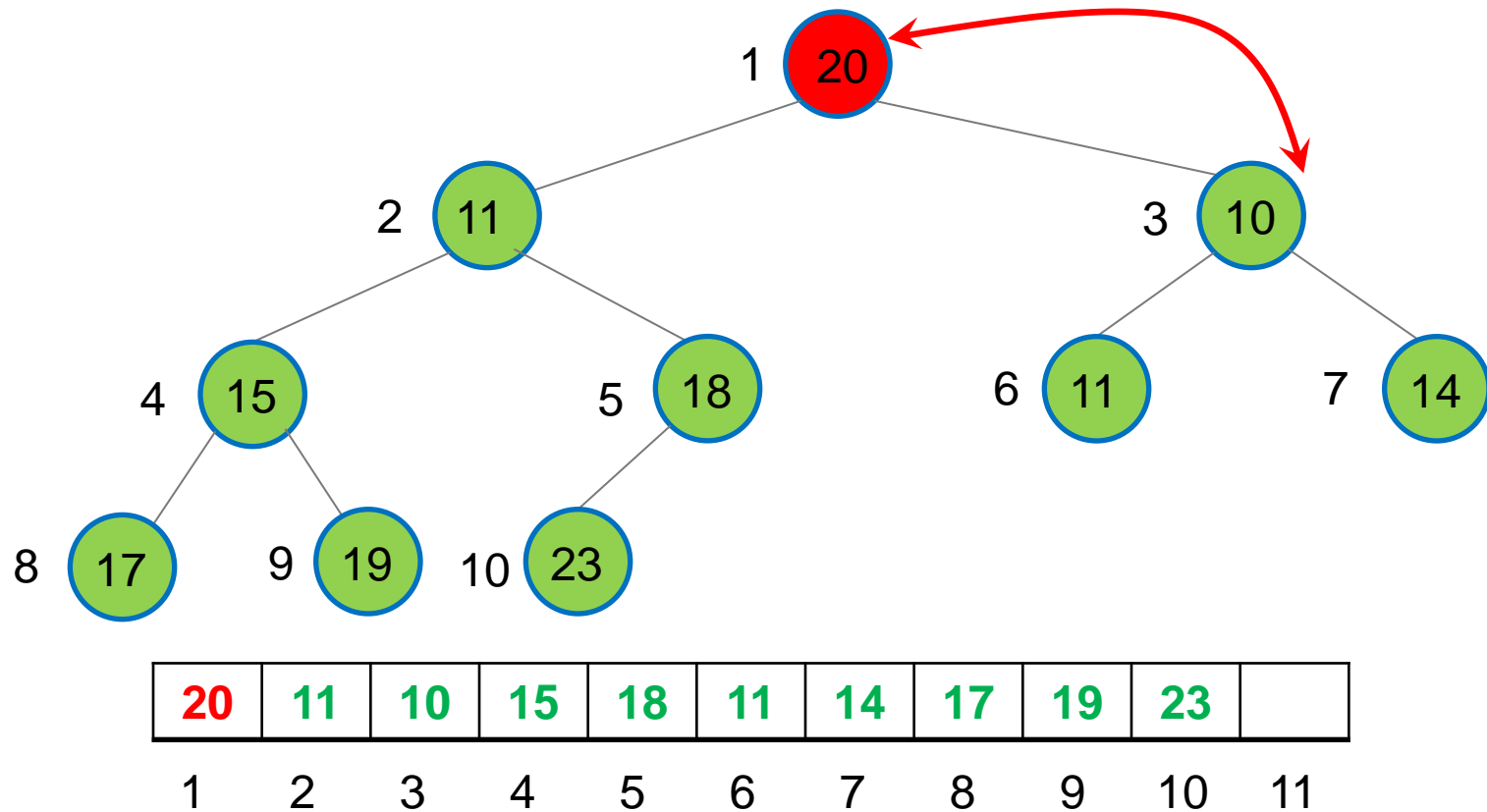
Deletion from Heap (downHeap)

- Delete Array[1]
- Move Array[N] (called last) to Array[1]



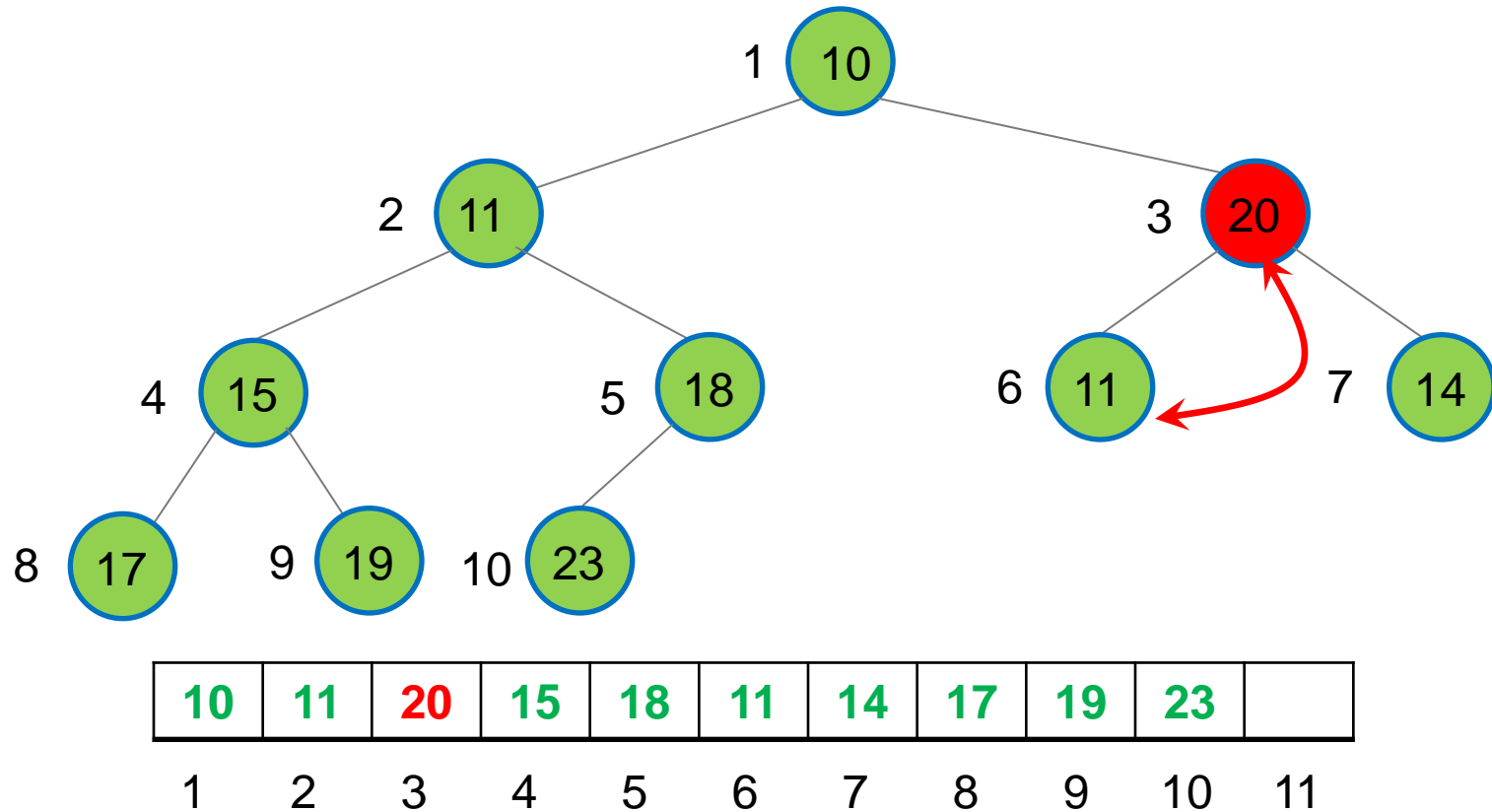
Deletion from Heap (downHeap)

- Delete Array[1]
- Move Array[N] (called last) to Array[1]
- While leftchild(last) < last or rightchild(last) < last
 - Swap last with minimum(leftchild, rightchild)



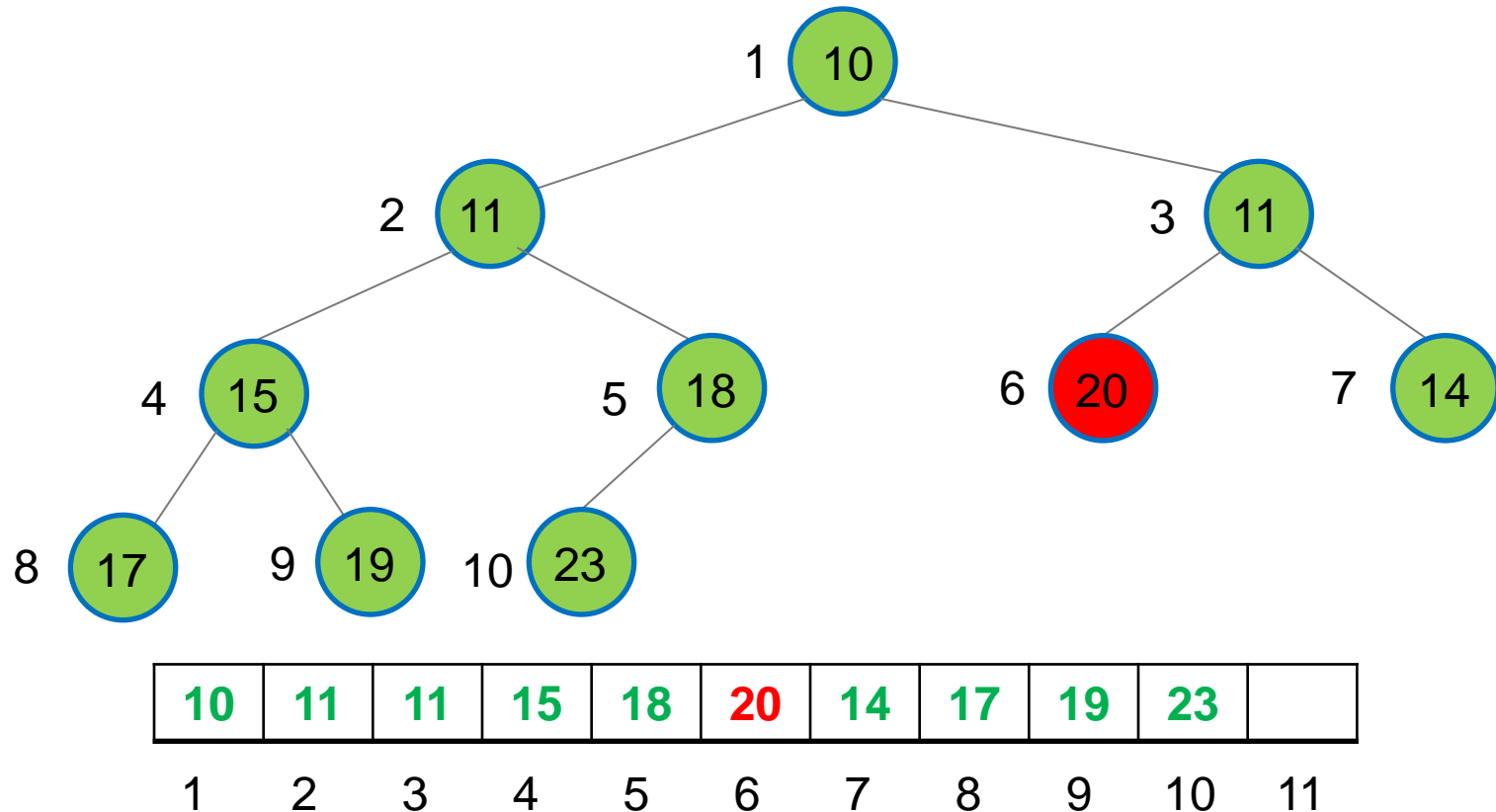
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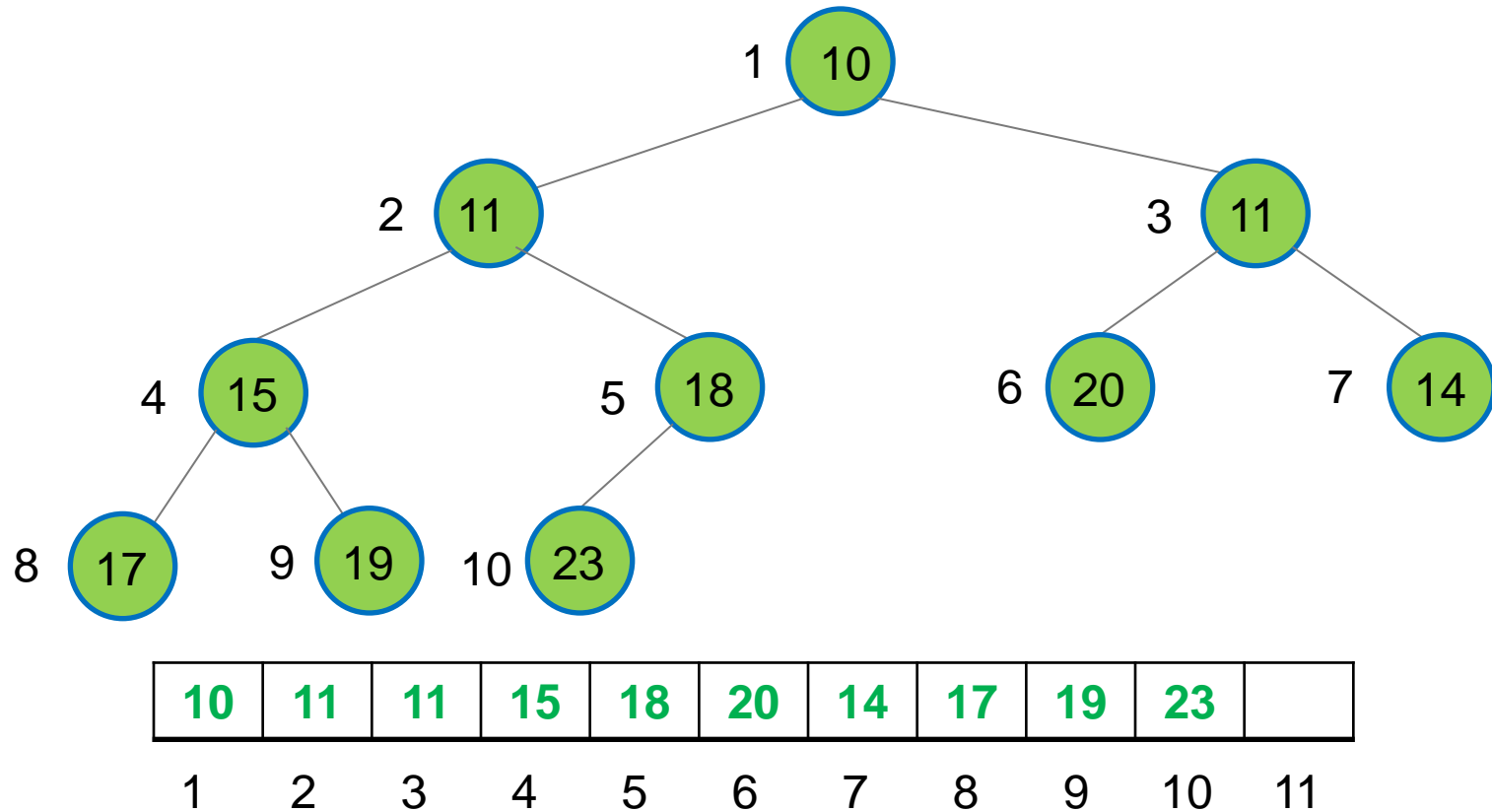
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Deletion from Heap (downHeap)

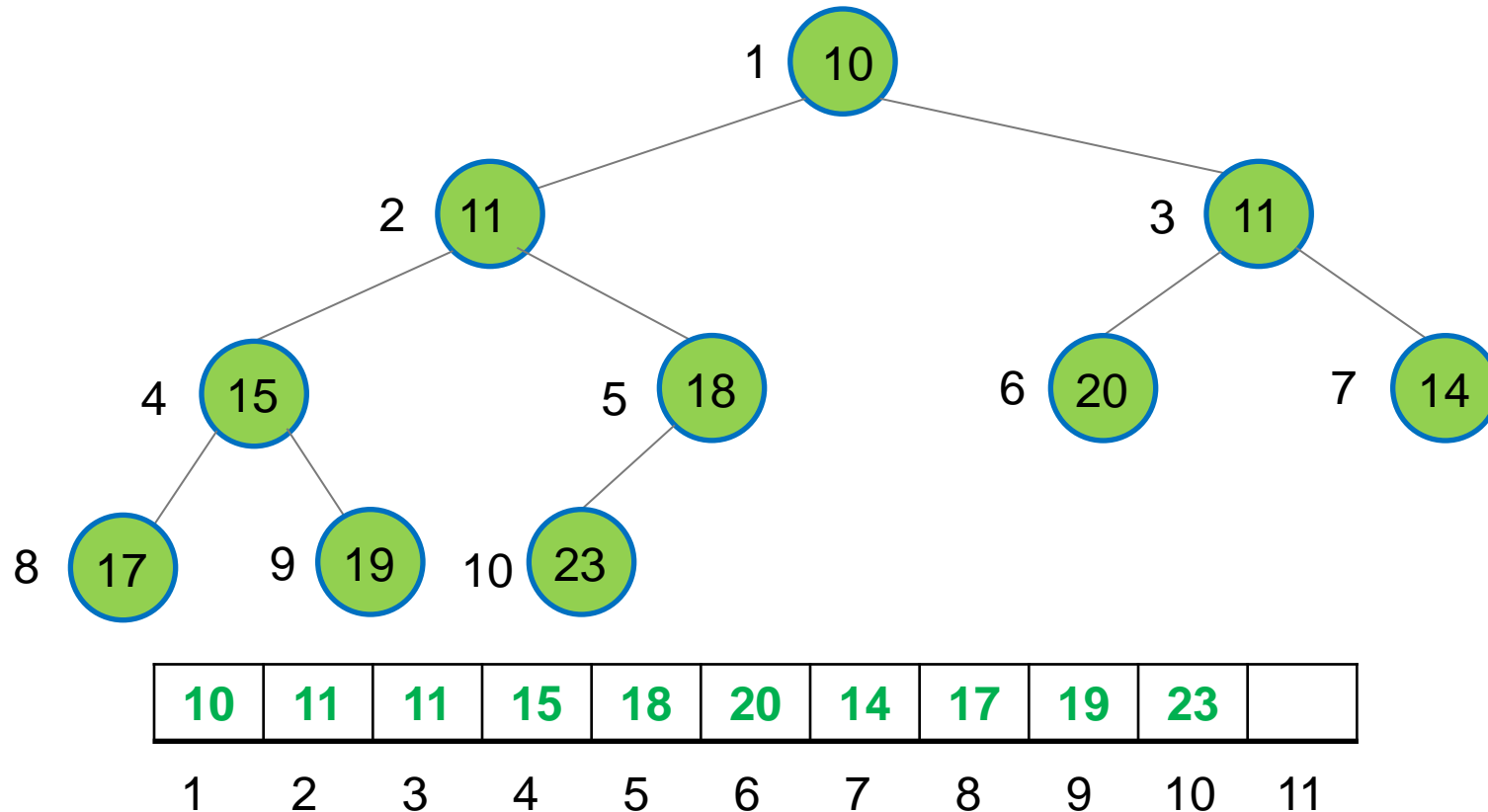
- Delete Array[1]
- Move Array[N] (called last) to Array[1]
- While leftchild(last) < last or rightchild(last) < last
 - Swap last with minimum(leftchild, rightchild)



Complexity of downHeap?

- Delete Array[1]
- Move Array[N] (called last) to Array[1]
- While leftchild(last) < last or rightchild(last) < last
 - Swap last with minimum(leftchild, rightchild)

$O(\log N)$



Heapify

The heapify procedure builds a heap from an unsorted array

A straightforward approach:

- Initialize an empty heap
- **for** $i = 1; i \leq N; i++$
- insert array[i] **in** heap

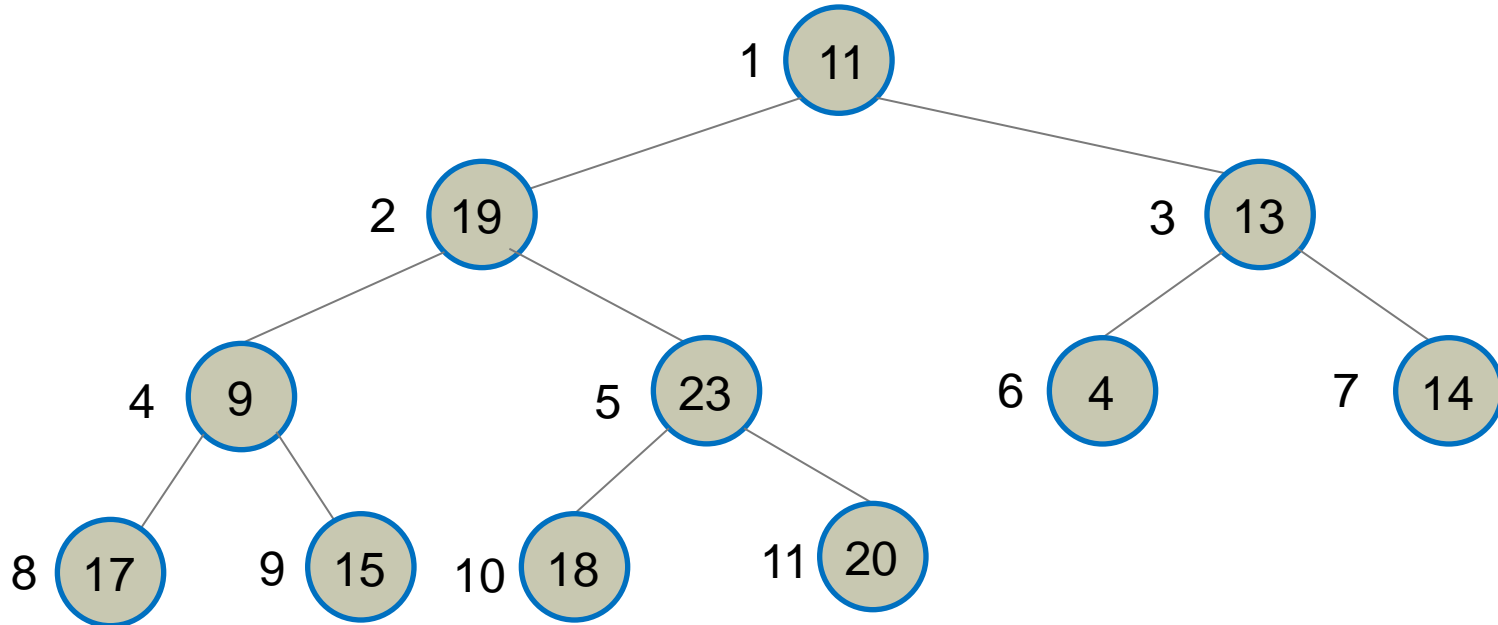
Time Complexity

$$\begin{aligned} & \log(1) + \log(2) + \log(3) + \dots + \log(N) \\ &= \log(1 \times 2 \times 3 \times \dots \times N) = \log(N!) \\ &= O(N \log N) \end{aligned}$$

11	19	13	9	23	4	14	17	15	18	20
1	2	3	4	5	6	7	8	9	10	11

Heapify in $O(N)$

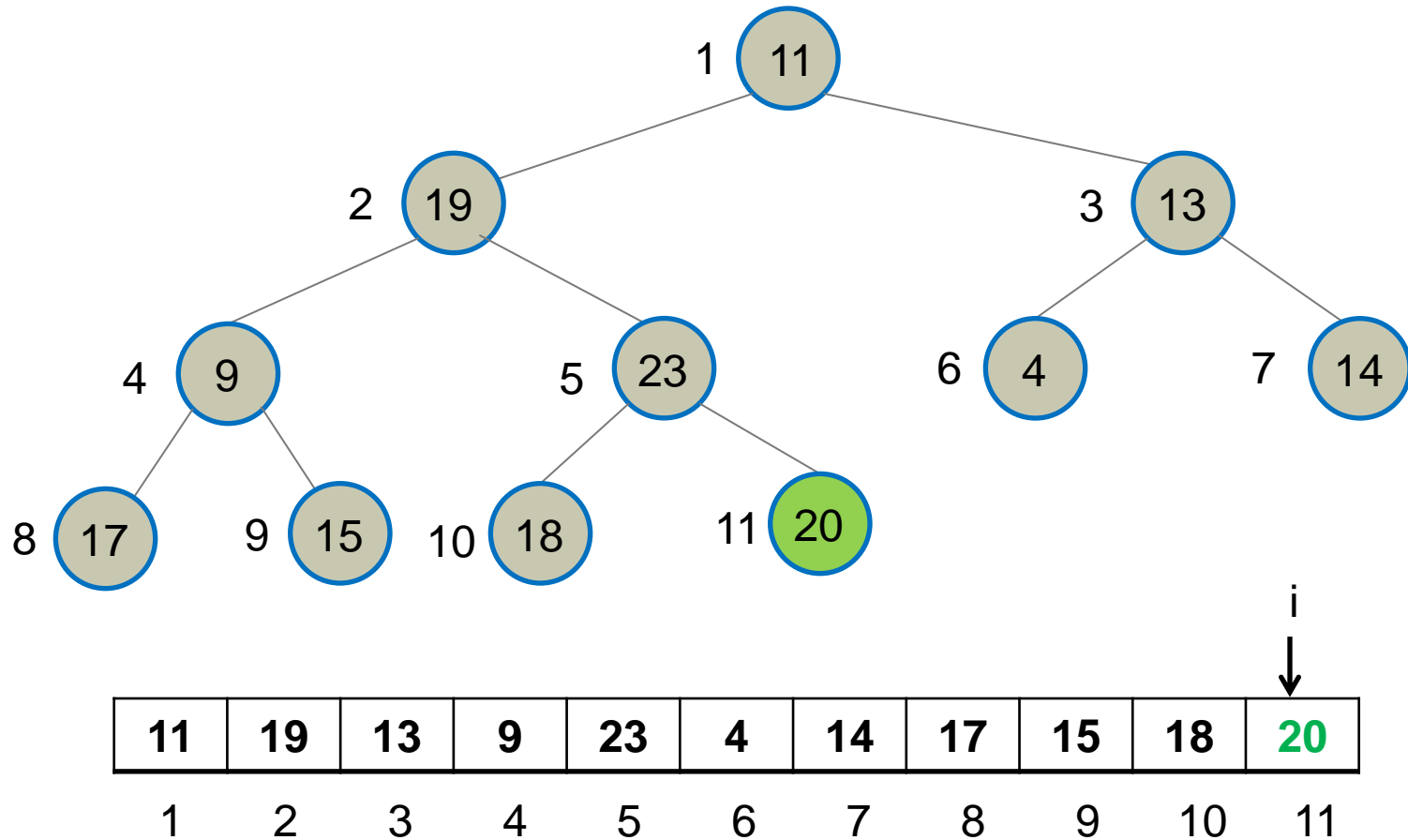
- **for** $i = N; i \geq 1; i--$
- downHeap array[i] in the heap rooted at array[i]



11	19	13	9	23	4	14	17	15	18	20
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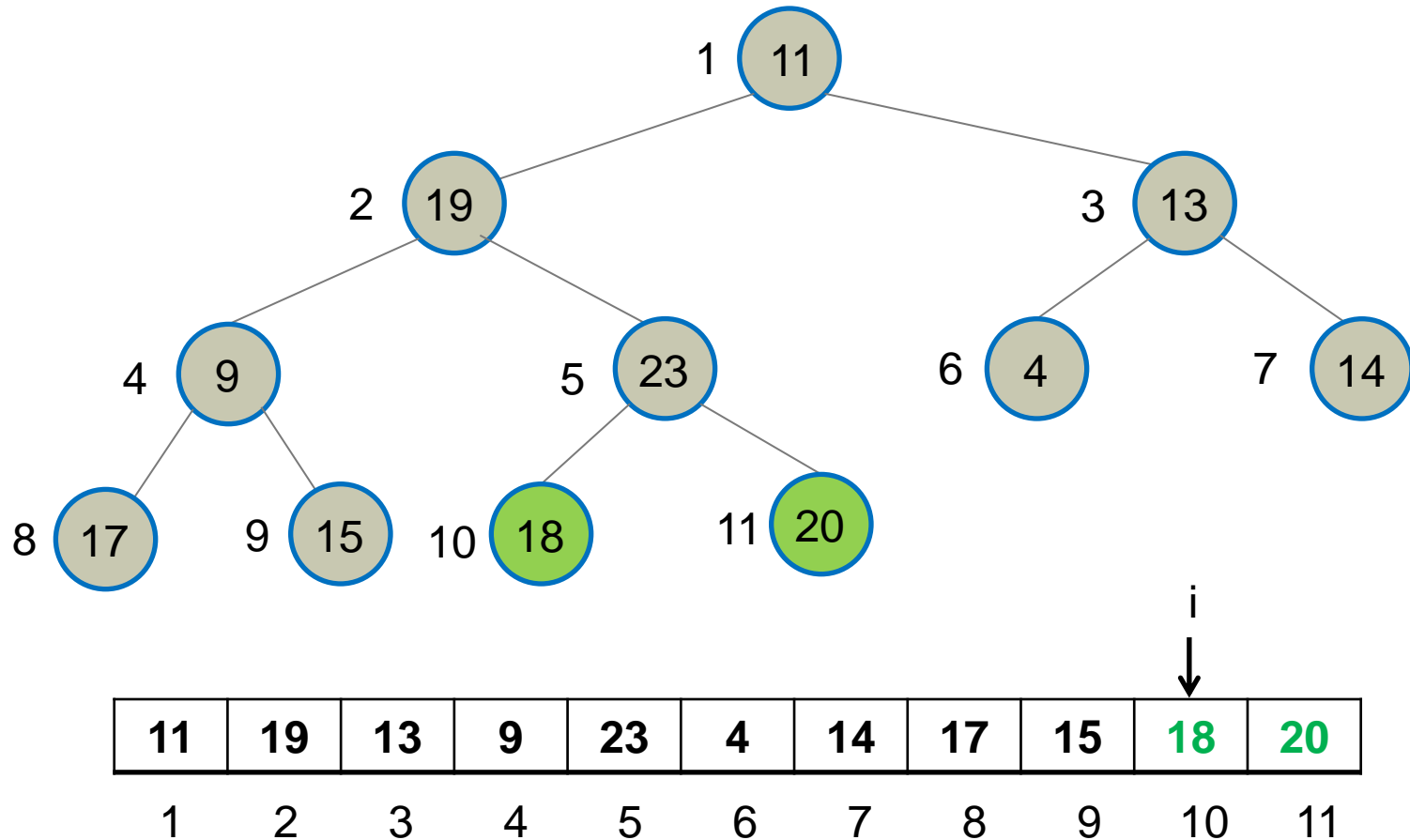
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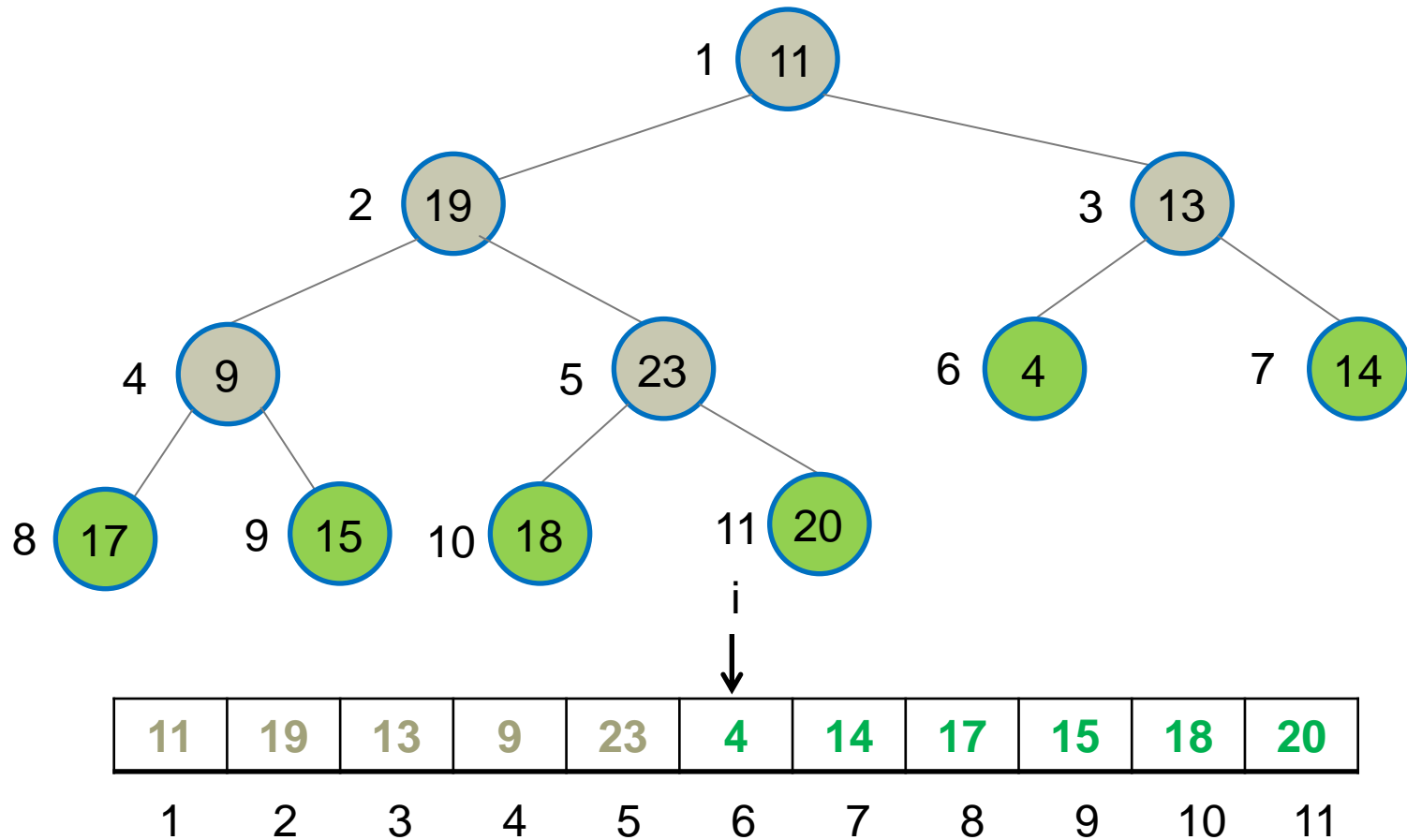


Heapify in $O(N)$

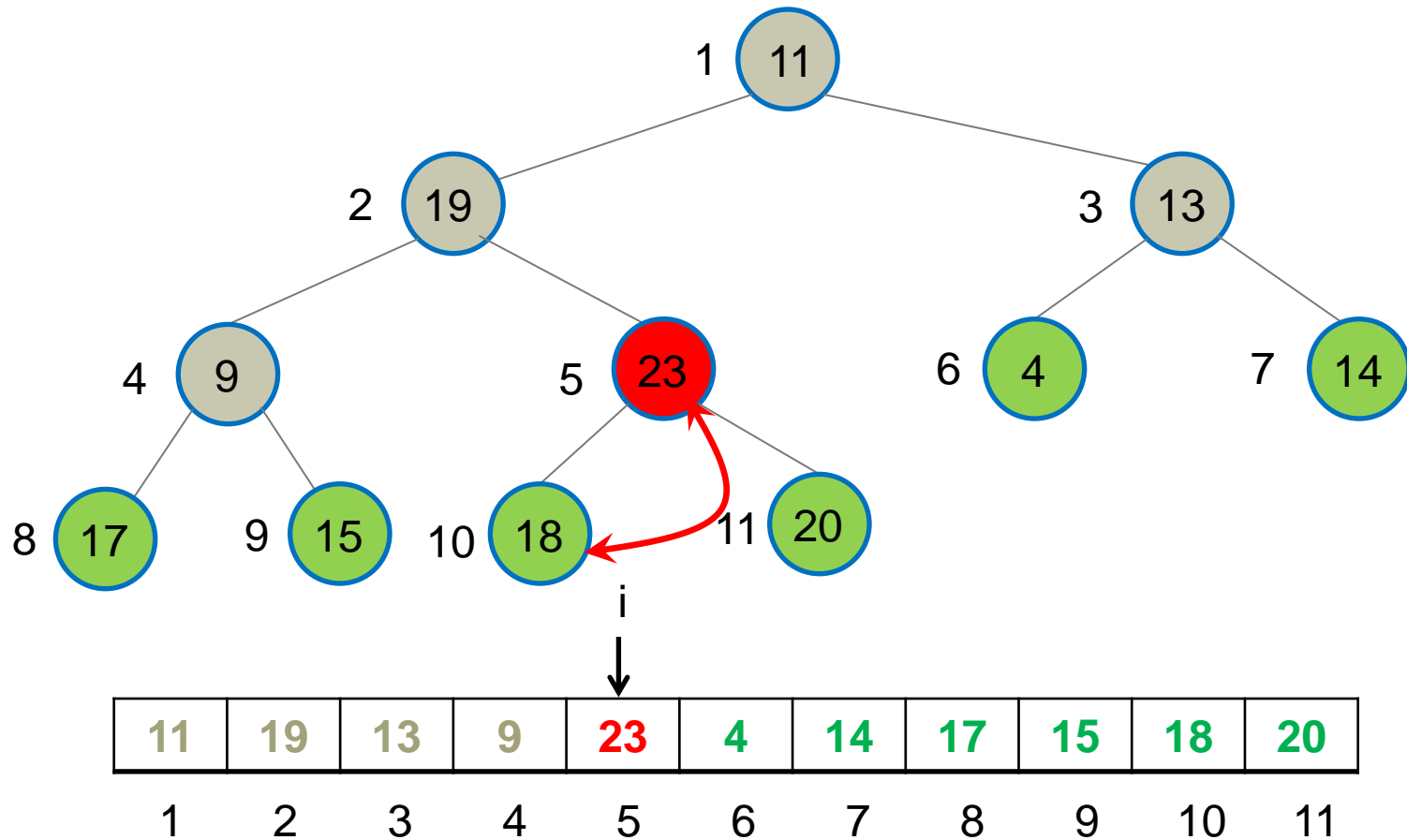
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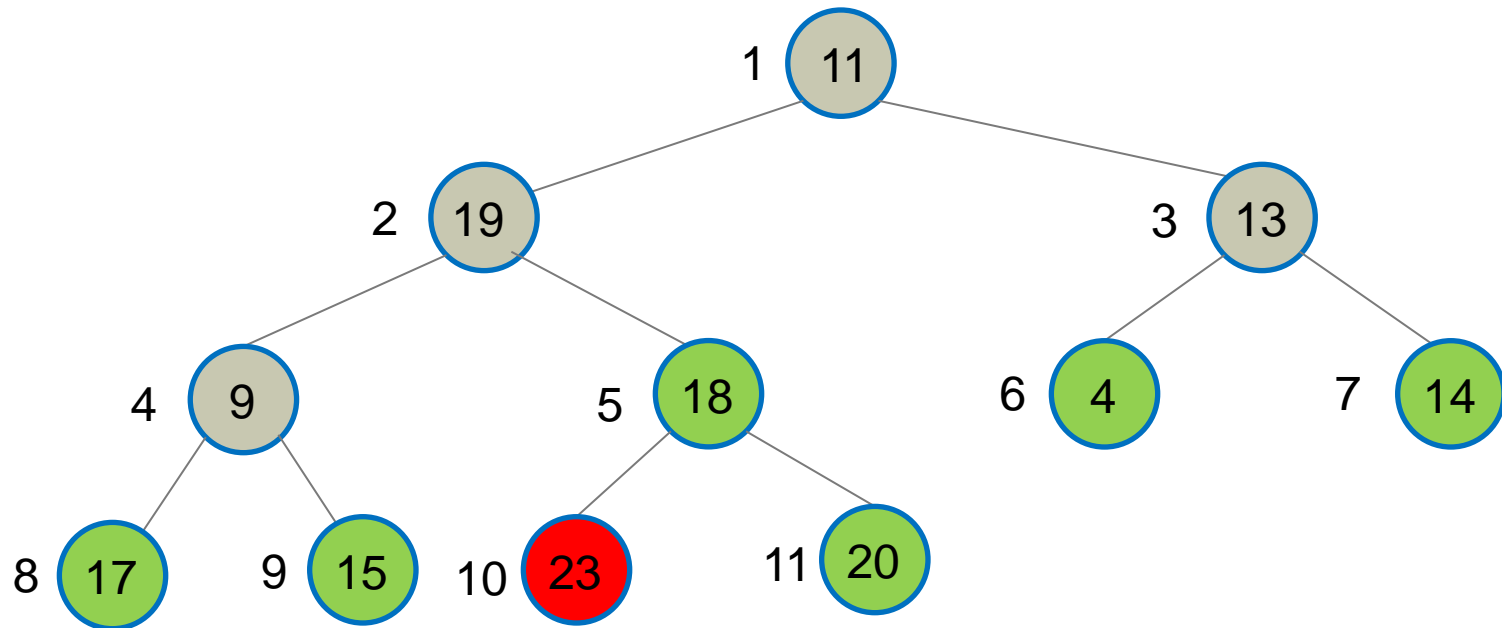
Heapify in $O(N)$



Heapify in $O(N)$

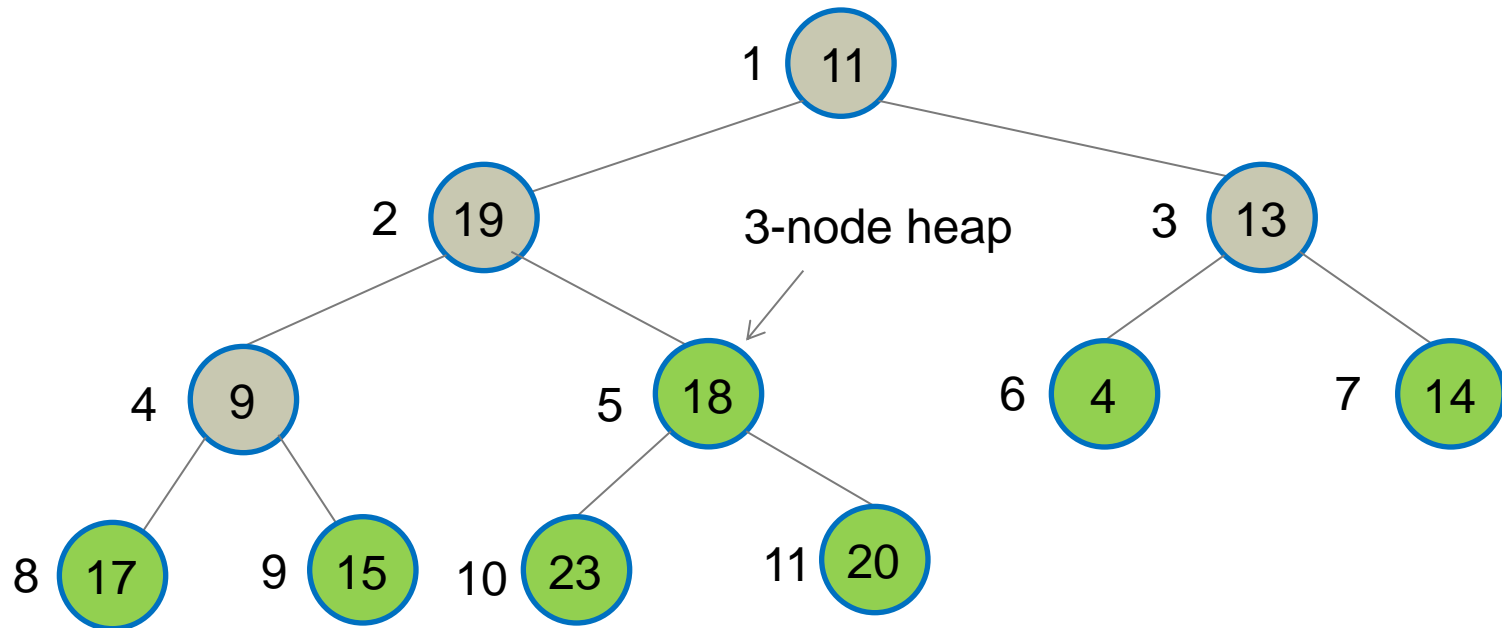


Heapify in $O(N)$



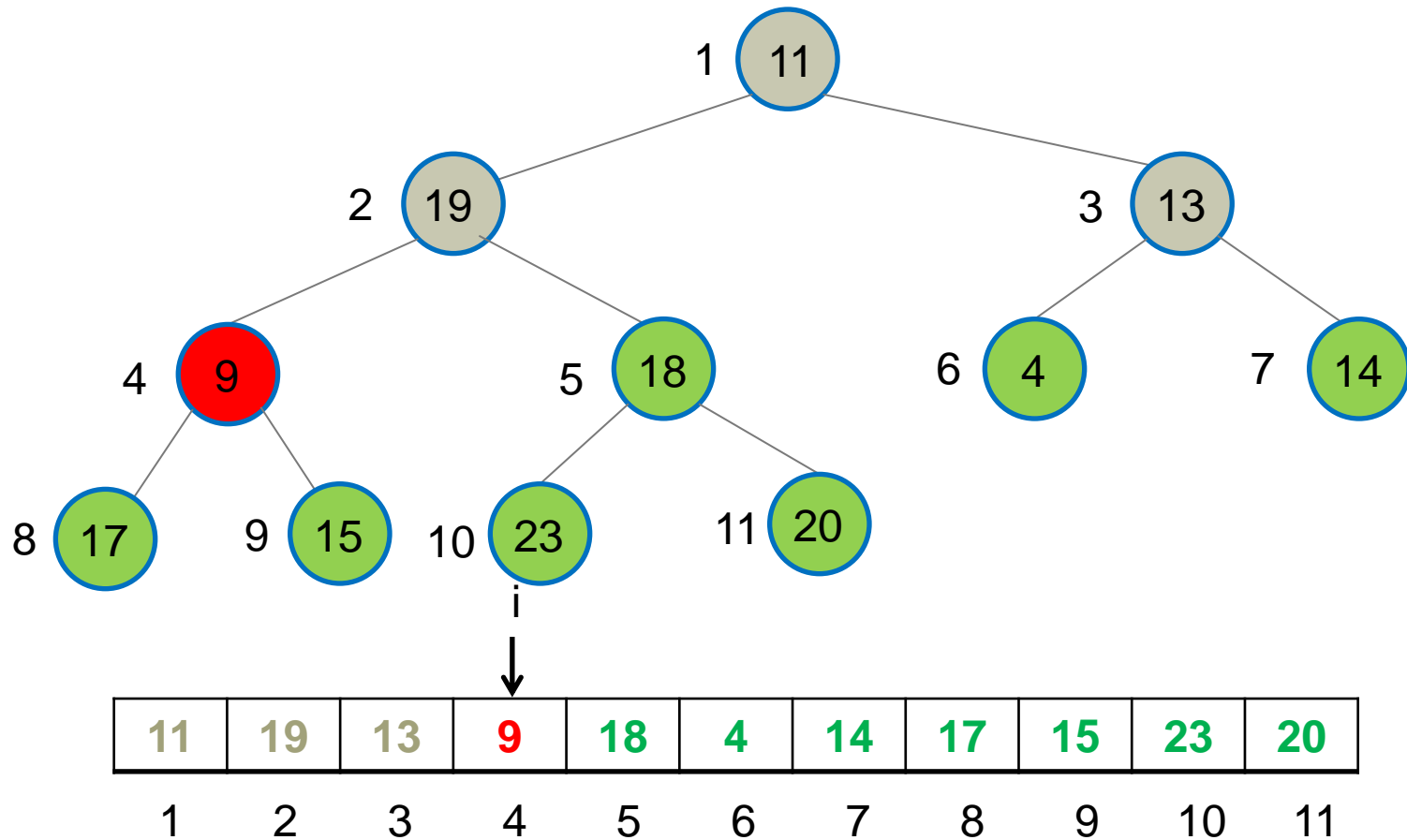
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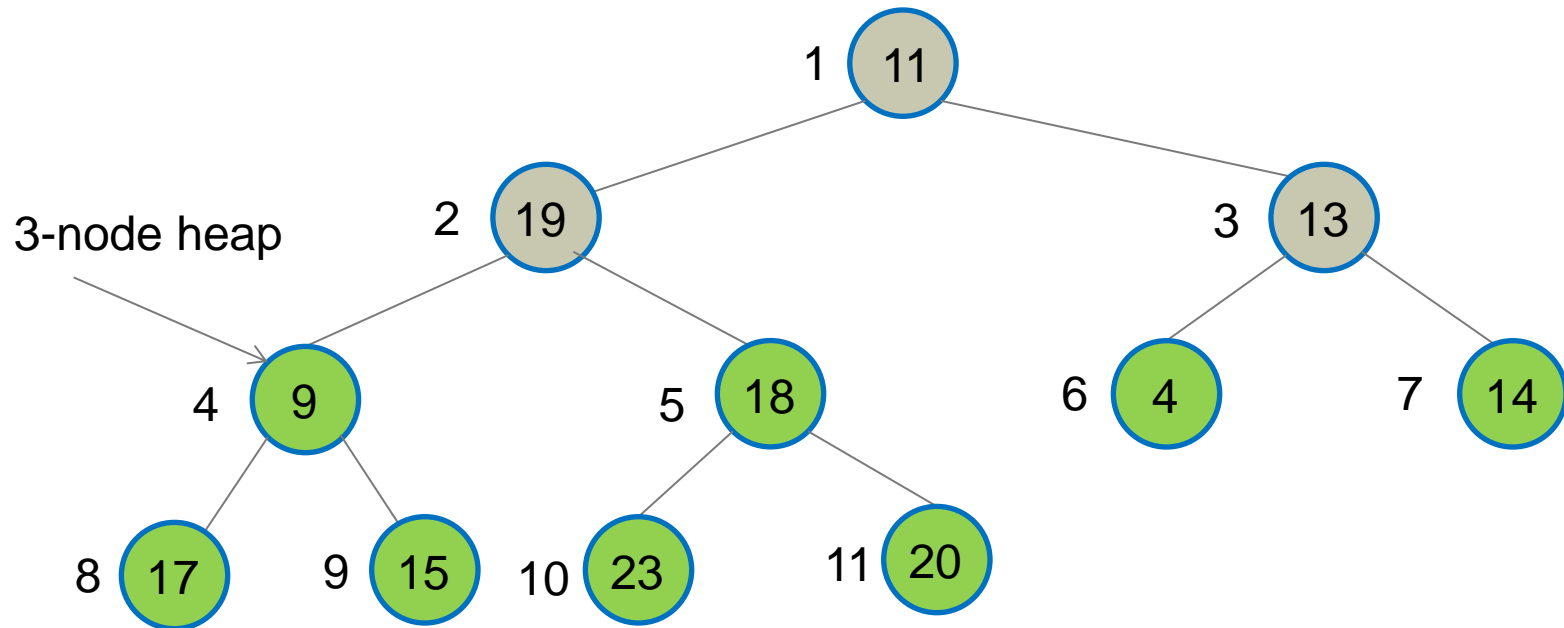


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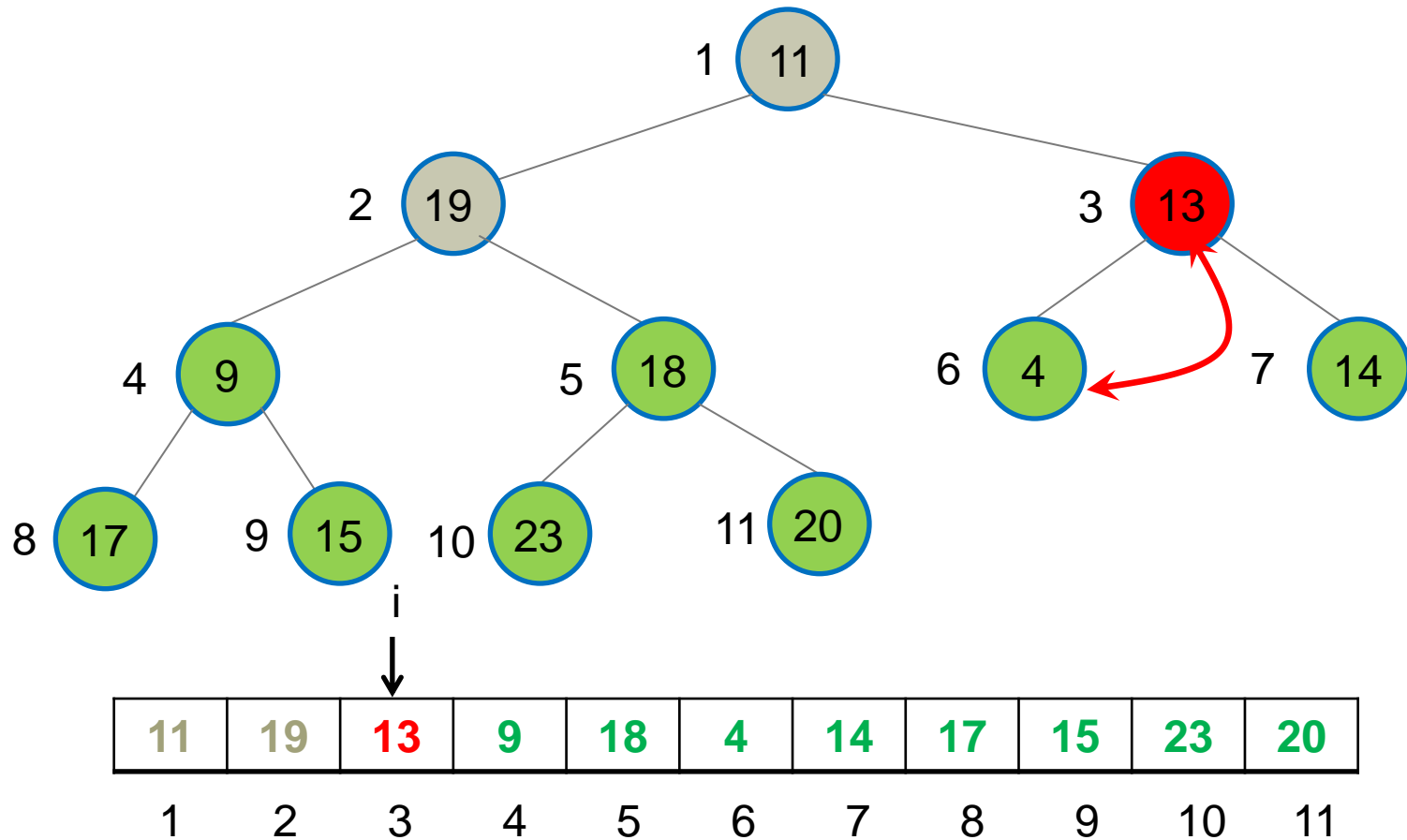


Heapify in $O(N)$

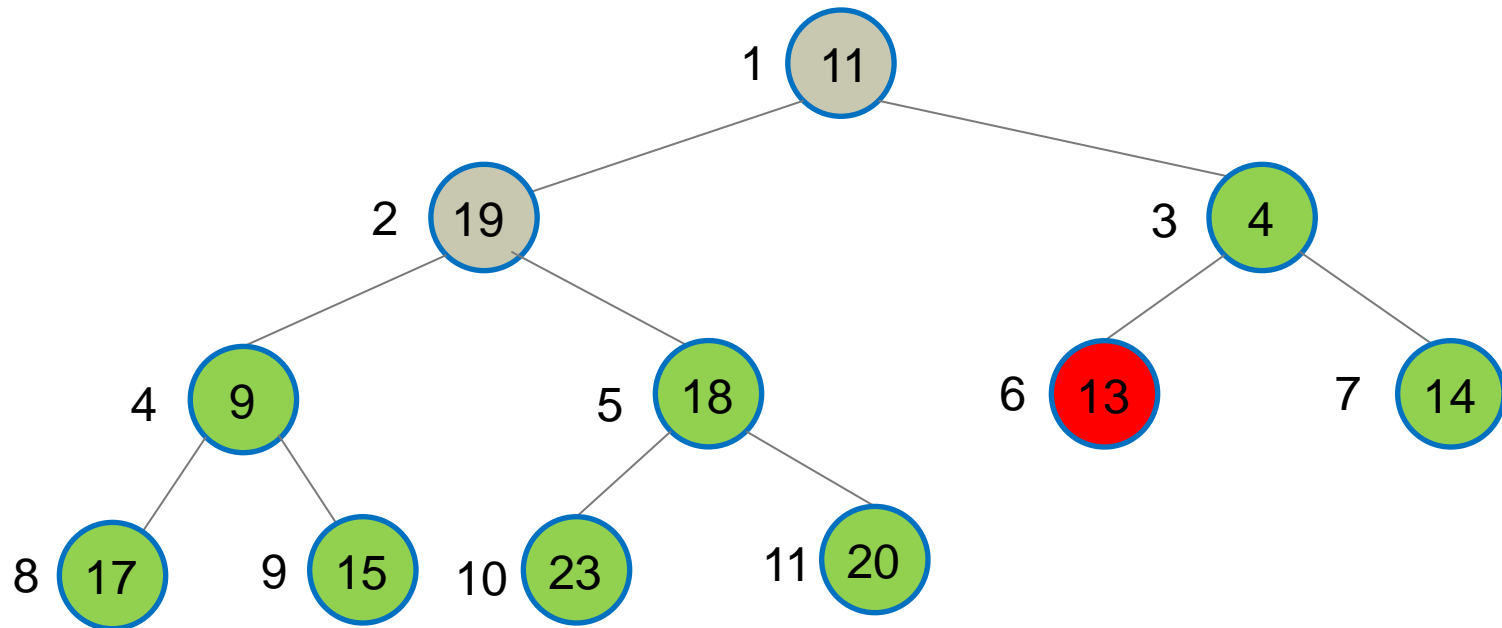


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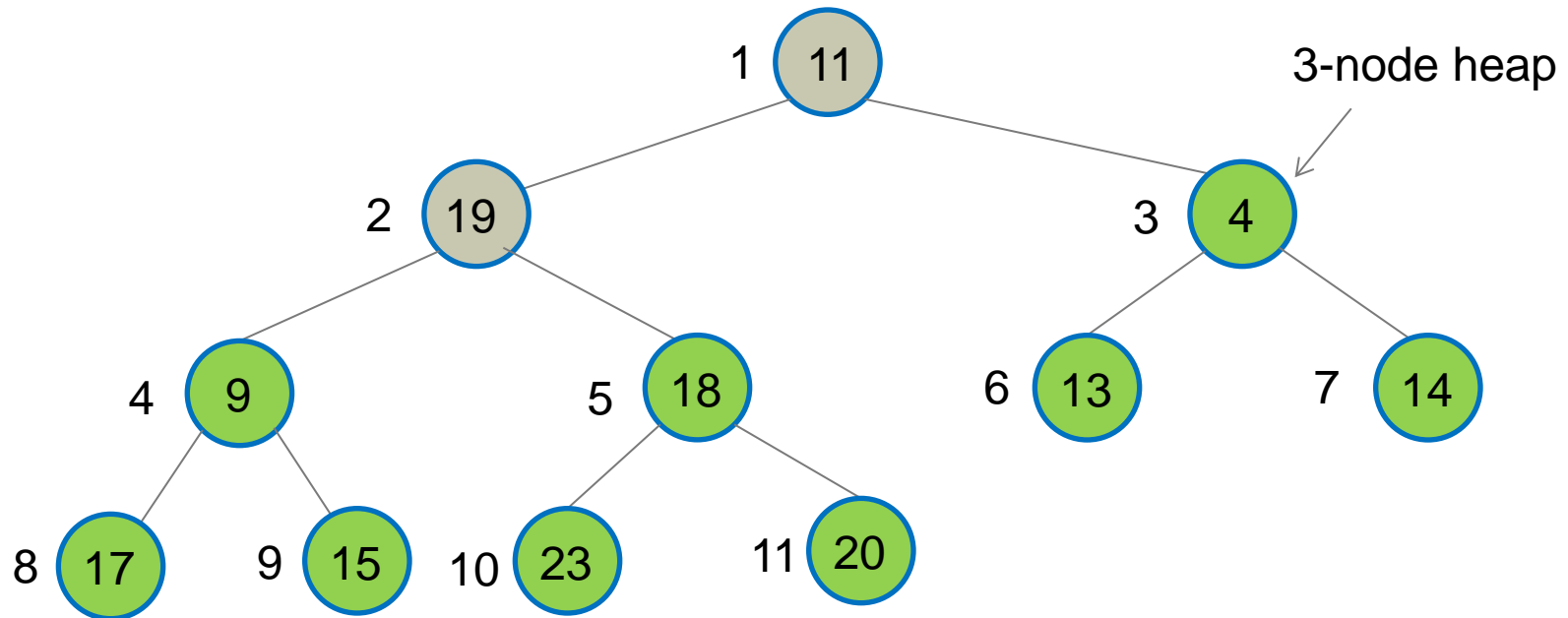


Heapify in $O(N)$



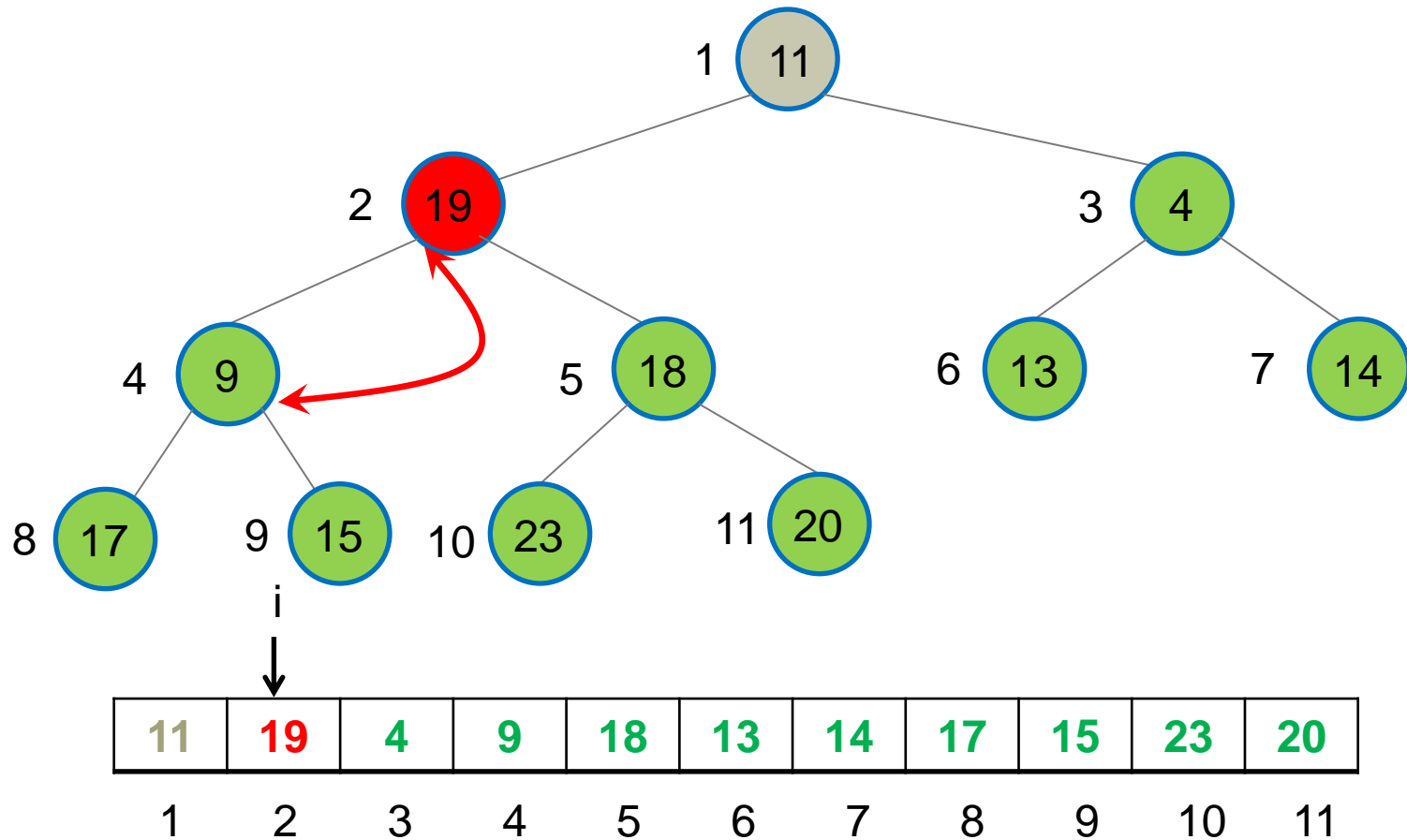
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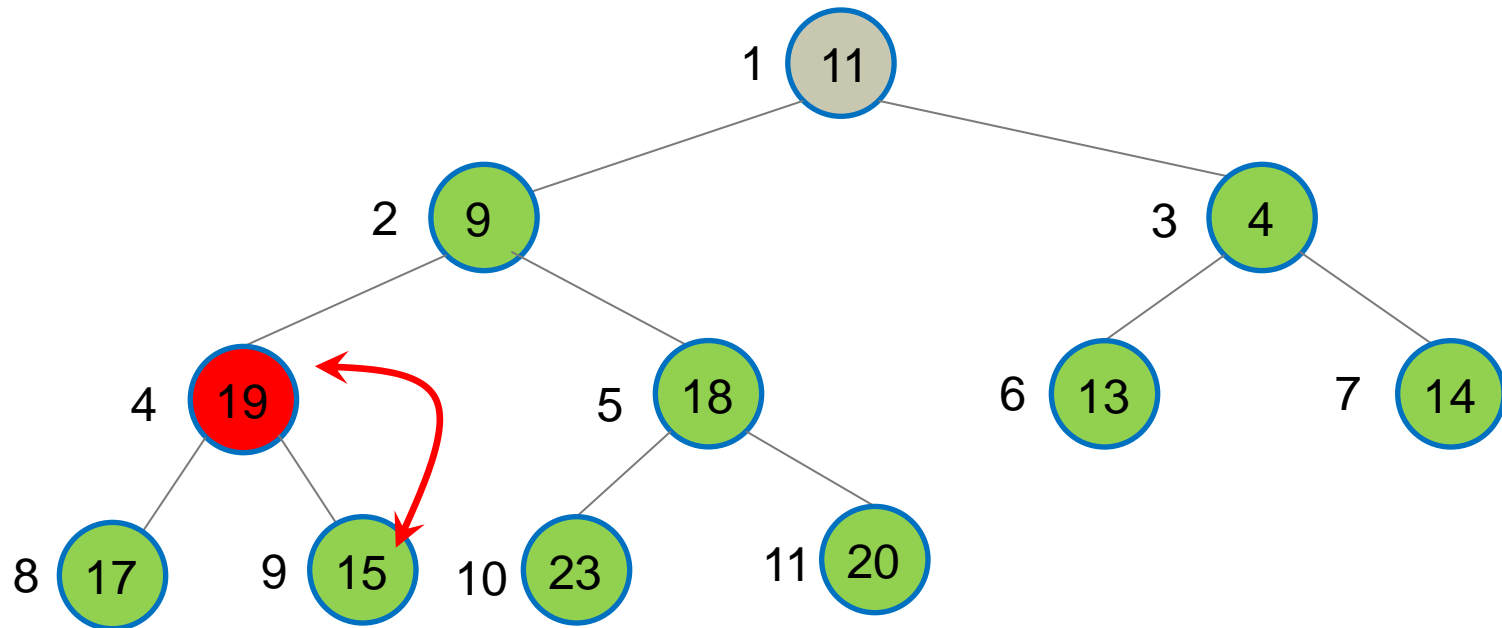


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Heapify in $O(N)$

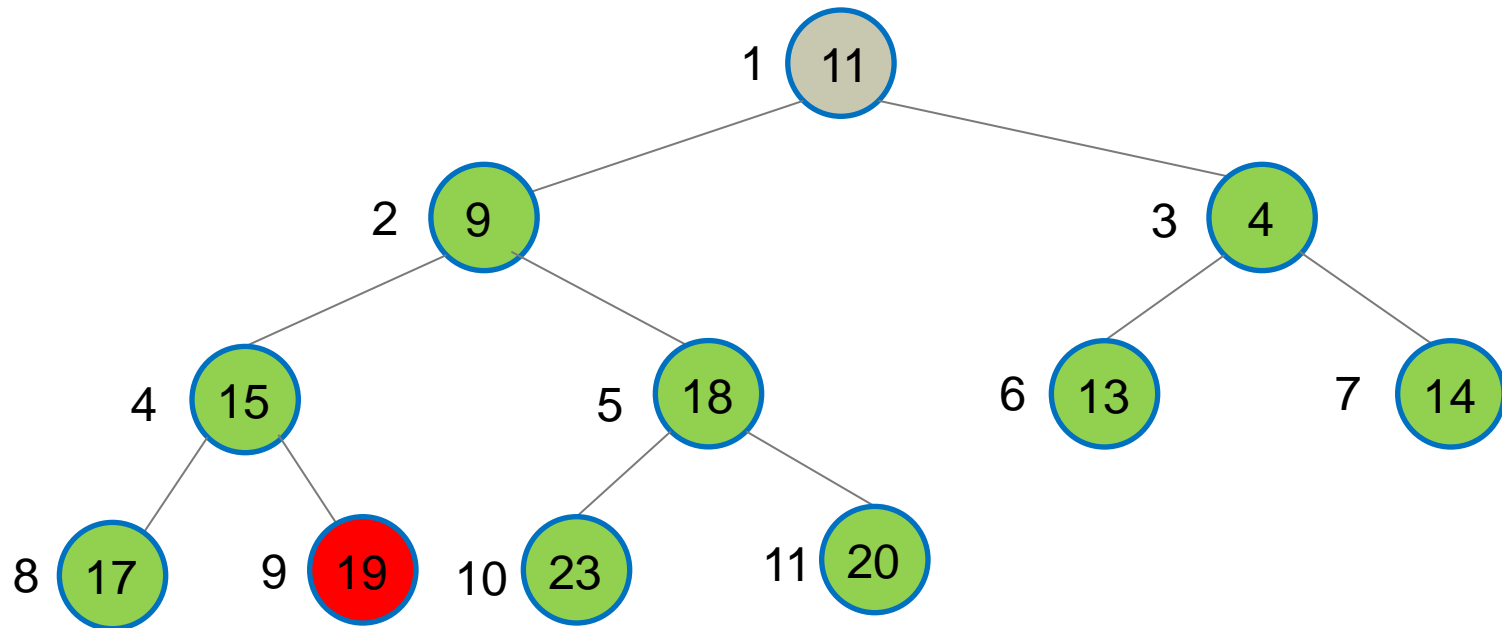


Heapify in $O(N)$



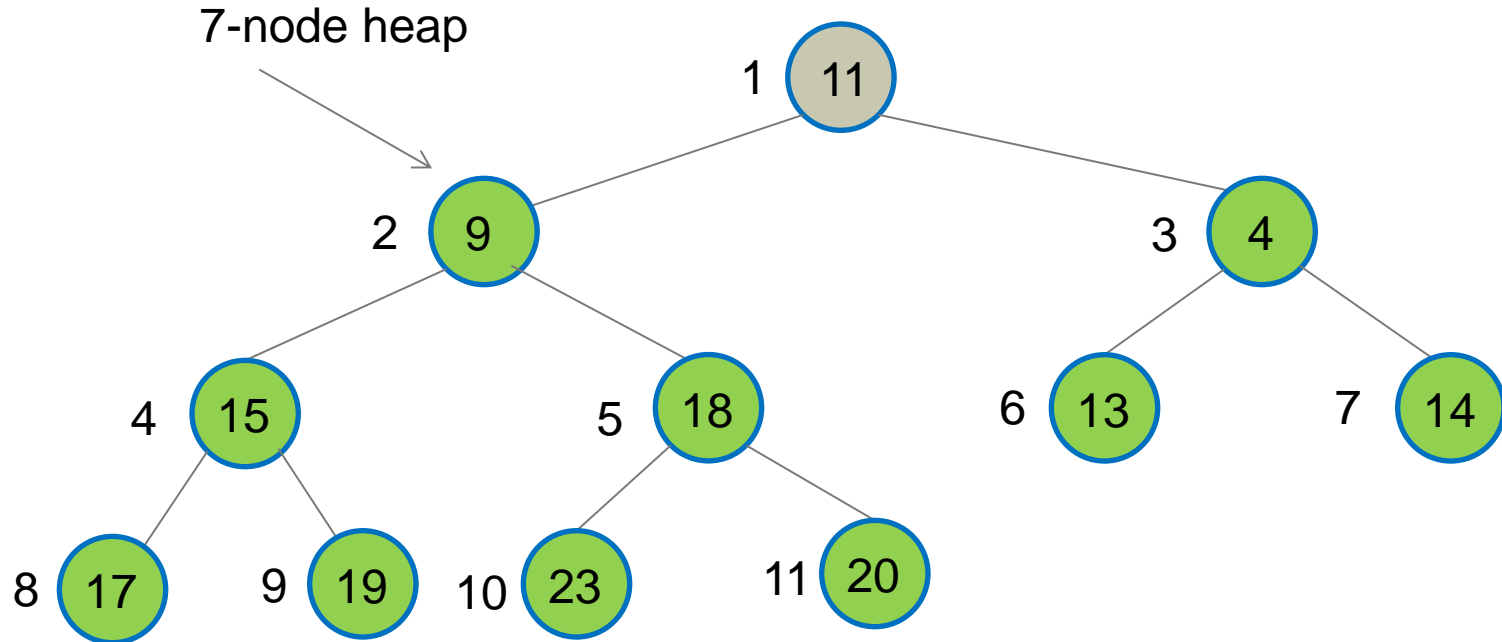
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Heapify in $O(N)$



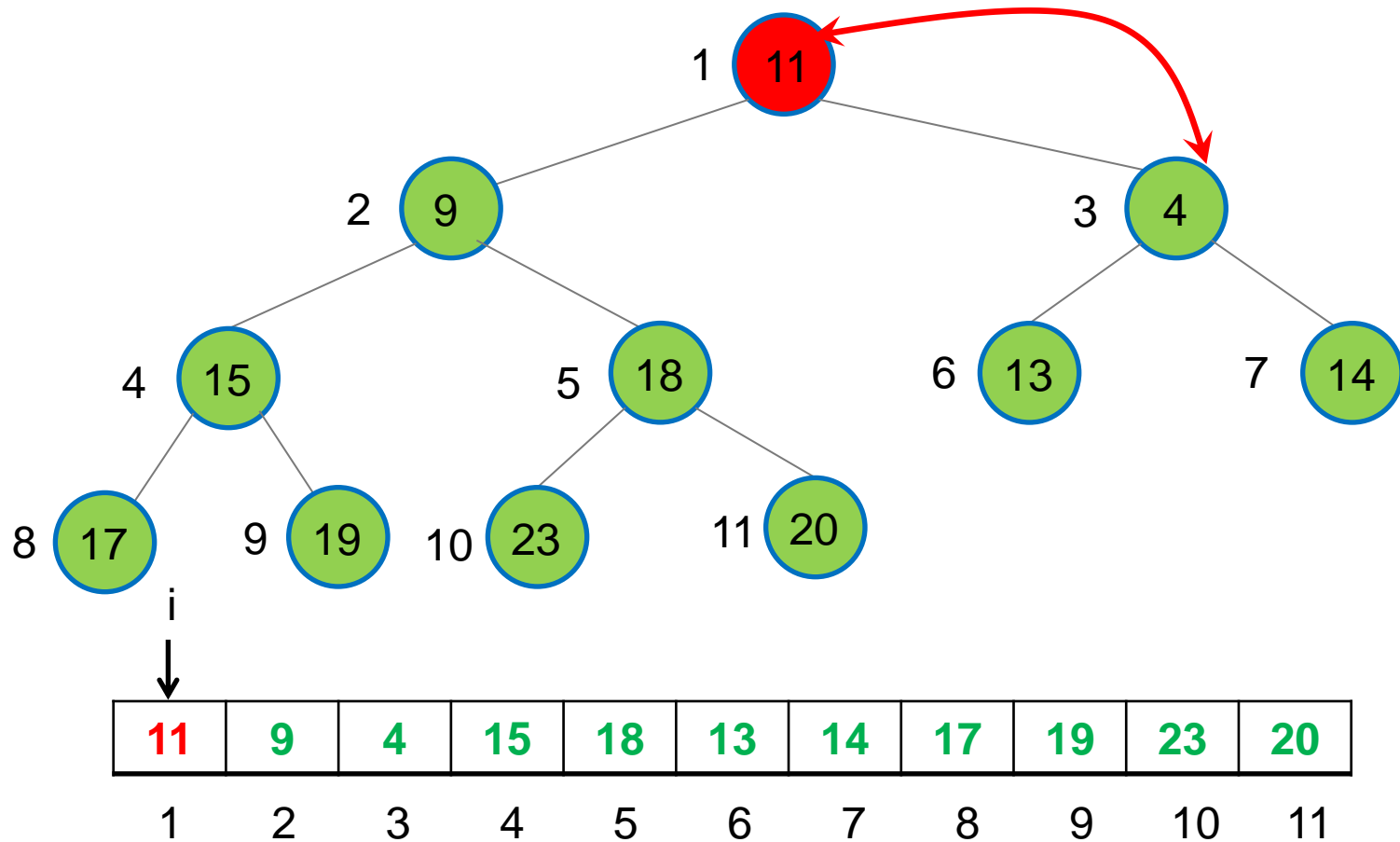
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Heapify in $O(N)$

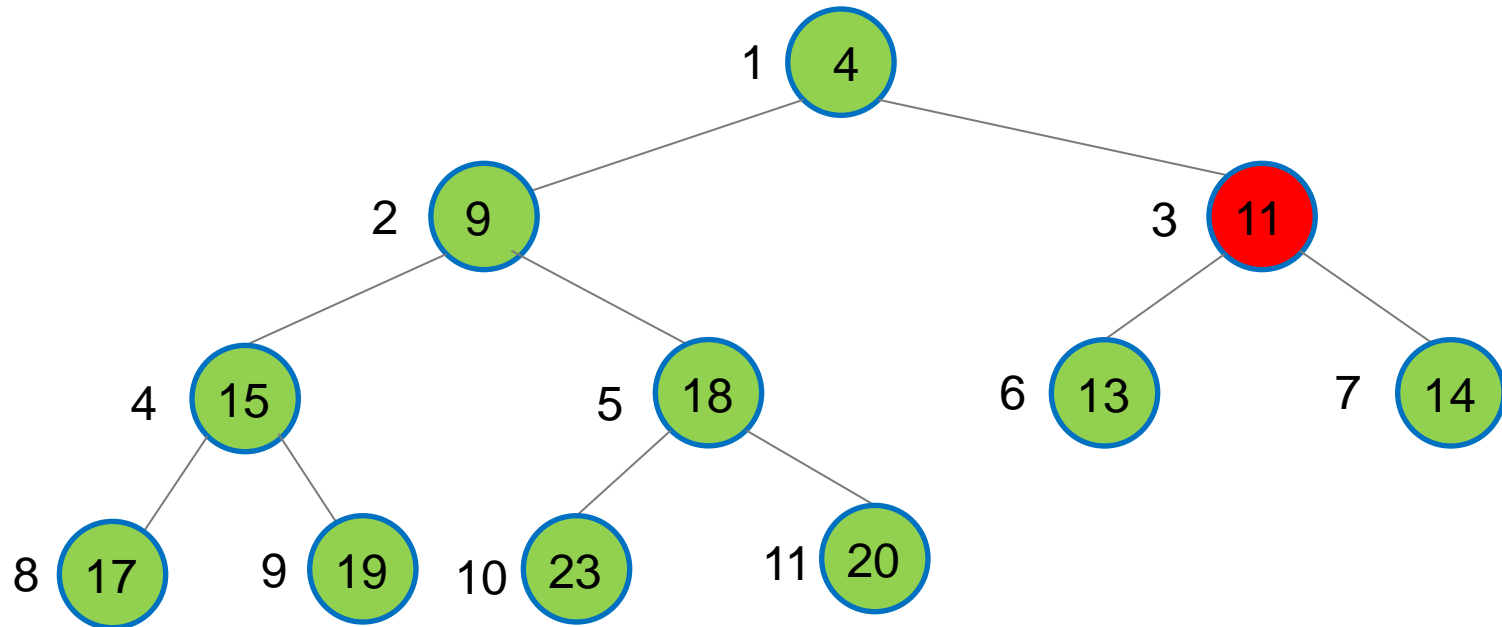


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1	2	3	4	5	6	7	8	9	10	11

Heapify in $O(N)$

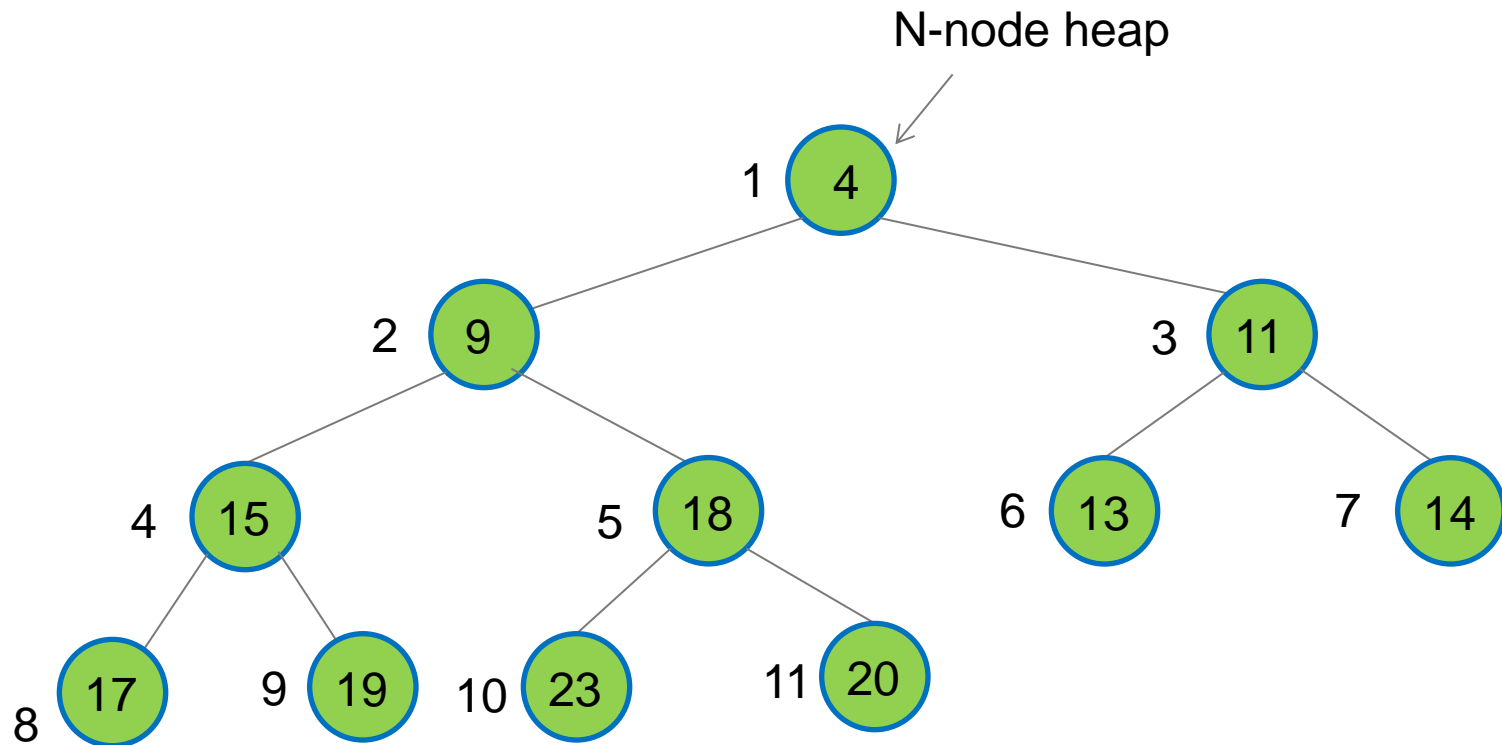


Heapify in $O(N)$



4	9	11	15	18	13	14	17	19	23	20
1	2	3	4	5	6	7	8	9	10	11

Heapify in $O(N)$



4	9	11	15	18	13	14	17	19	23	20
1	2	3	4	5	6	7	8	9	10	11

Heapify in $O(N)$

Complexity Analysis

Swaps for level 1 nodes (i.e., leaf nodes):	0	number of nodes: $N/2$
Swaps for level 2 nodes:	1	number of nodes: $N/4$
Swaps for level 3 nodes:	2	number of nodes: $N/8$
Swaps for level 4 nodes:	3	number of nodes: $N/16$
...		
Swap for root node:	h	number of nodes: 1

Total cost: $1 \cdot N/4 + 2 \cdot N/8 + 3 \cdot N/16 + \dots + h \cdot 1$

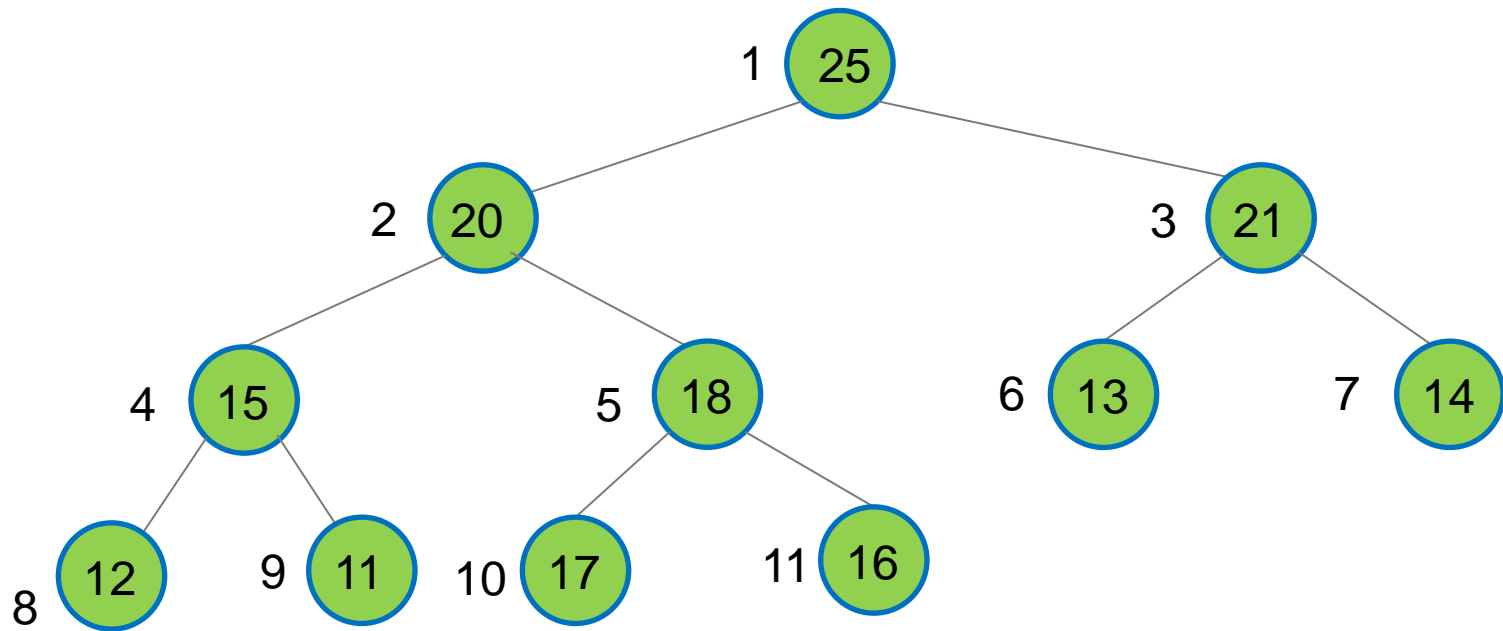
$$= N(1/4 + 2/8 + 3/16 + \dots + h/N)$$

$\leq N(1/4 + 2/8 + 3/16 + \dots)$ // $(1/4 + 2/8 + 3/16 + \dots)$ converges to a constant)

$$= O(N)$$

Min-Heap and Max-Heap

- The examples we saw earlier prefer smaller elements (i.e., smaller element is at the top of the heap). Such heap is called a min-heap
- A max-heap prefers larger elements and can be implemented similarly



Heap Sort

- Heapify the input array A in a min-heap
- Initialize an empty array B of size N
- For $i = 1$ to N
 - Remove the top element from the heap (e.g., A[1]) and put it at B[i]
- Copy array B to array A

Time complexity:

Space complexity:

Is the above sorting algorithm stable?

In-place Algorithm: An algorithm is called in-place if it uses only constant space (i.e., $O(1)$) additional to the space used by input.

Is the above version of Heap Sort In-place?

The above version of Heap Sort is not in-place as it requires an additional $O(N)$ space for array B.

In-Place Heap Sort

- Heapify the input array A using a max-heap

- For $i = 1$ to N

→ Remove biggest element from the heap (i.e., $A[1]$)

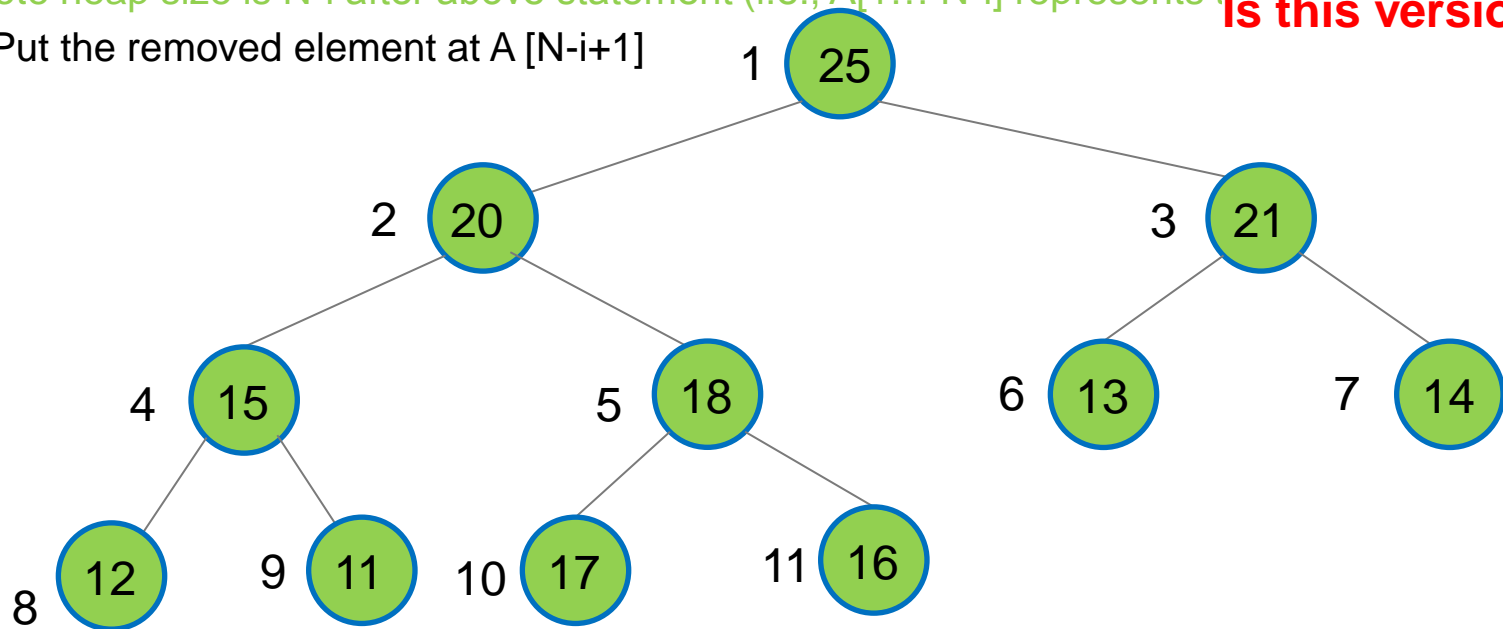
// Note heap size is $N-i$ after above statement (i.e., $A[1 \dots N-i]$ represents the heap)

→ Put the removed element at $A[N-i+1]$

Time complexity?

Space Complexity?

Is this version stable?



20	18	16	15	17	13	14	12	11	21	25
1	2	3	4	5	6	7	8	9	10	11

Divide and Conquer Sorting

Divide and Conquer Paradigm

- **Divide** the problem into smaller sub-problems
- **Conquer** (solve) each sub-problem
- **Combine** the results

Divide and Conquer Sorting Algorithms

- Merge Sort
- Quick Sort

Merge Sort

// Merge Function: Merging two sorted arrays

i = 1, **j** = 1;

while **i** <= a.length **and** **j** <= b.length

if a [**i**] < b [**j**]

 answer.append(a[**i**])

i++;

else

 answer.append(b[**j**])

j++;

while **i** <= a.length

 answer.append(a[**i**])

i++;

while **j** <= b.length

 answer.append(b[**j**]);

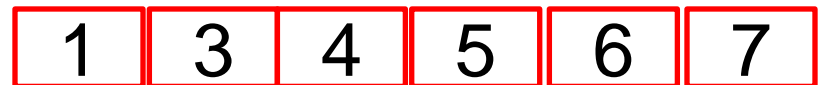
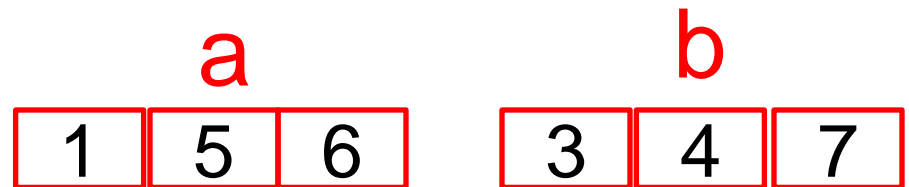
j++;

Time Complexity:

= $O(|a| + |b|)$

Is this In-place?

Is this version of merging stable?



answer

Merge Sort

// Merge Function: Merging two sorted arrays

i = 1, j = 1;

while i <= a.length **and** j <= b.length

if a [i] <= b [j]

answer.append(a[i])

i++;

else

answer.append(b[j])

j++;

while i <= a.length

answer.append(a[i])

i++;

while j <= b.length

answer.append(b[j]);

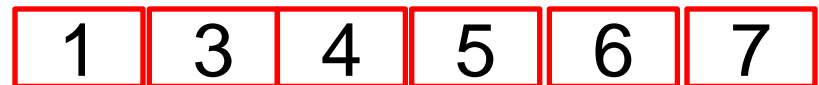
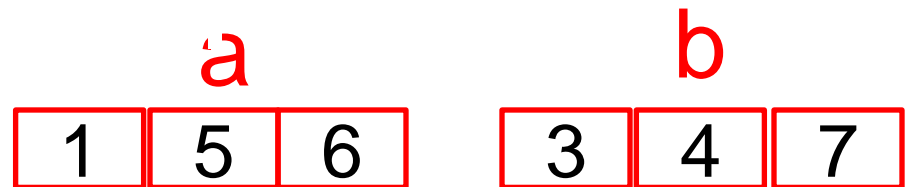
j++;

Time Complexity:

= $O(|a| + |b|)$

Note: Merge Function (hence Merge Sort) is not in-place

Is this version of merging stable?



answer

Merge Sort

Divide Step

- If array size is 1
 - return
- Else
 - divid

Conquer

- Recursi

Combine

- Combin

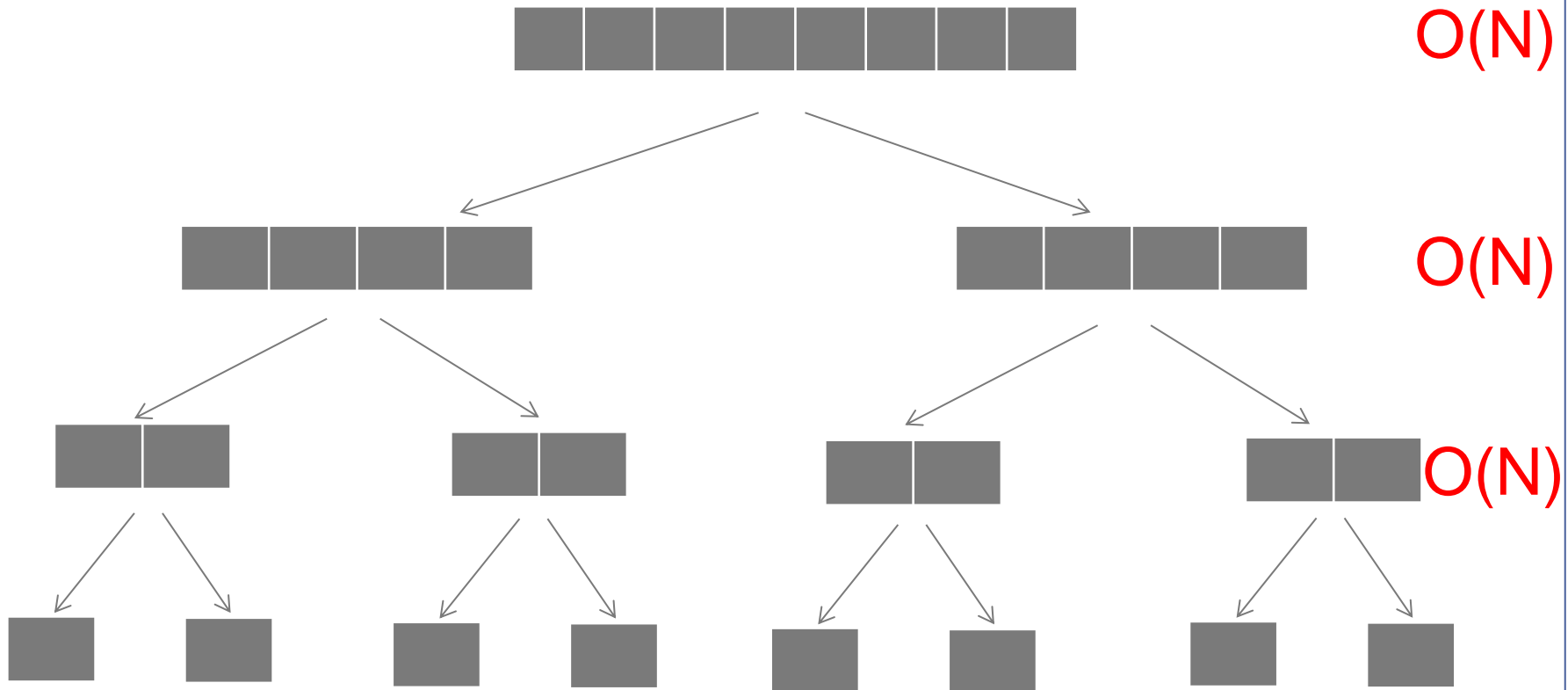
6 5 3 1 8 7 2 4

Source: [wikimedia.org](https://commons.wikimedia.org/wiki/File:MergeSortDiagram.svg)

Is Merge Sort stable?

Is Merge Sort in-place?

Complexity of Merge Sort



Height: $O(\log N)$

Worst-case complexity: $O(N \log N)$

Best-case time complexity?
Average-case time complexity?

Quicksort

Partitioning

- Choose a pivot p
- Partition the array in two sub-arrays w.r.t. p
 - LEFT \leftarrow elements smaller than or equal to p
 - RIGHT \leftarrow elements greater than p
- QuickSort(LEFT)
- QuickSort(RIGHT)



In sorted position

Pivot **X**

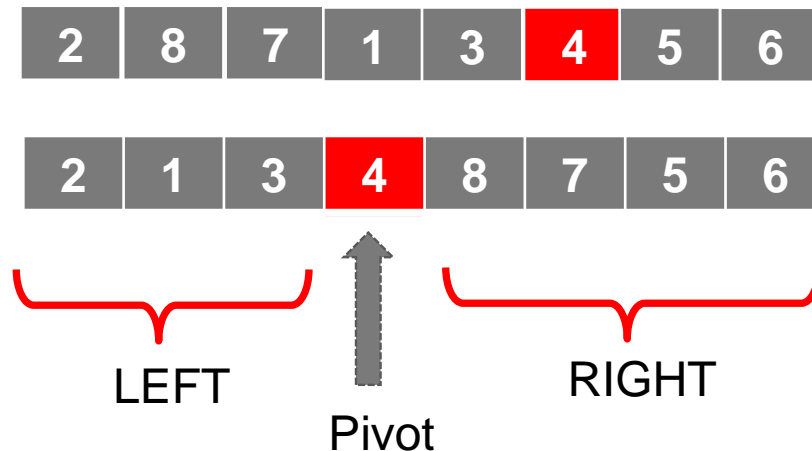
In Sorted position **X**

Others **X**

Partitioning: An out-of-place version

- Initialize two lists LEFT and RIGHT
- For each element e (except pivot)
 - If $e \leq \text{pivot}$
 - ✦ Insert e in LEFT
 - If $e > \text{pivot}$
 - ✦ Insert e in RIGHT
- Copy {LEFT, pivot, RIGHT} to the array

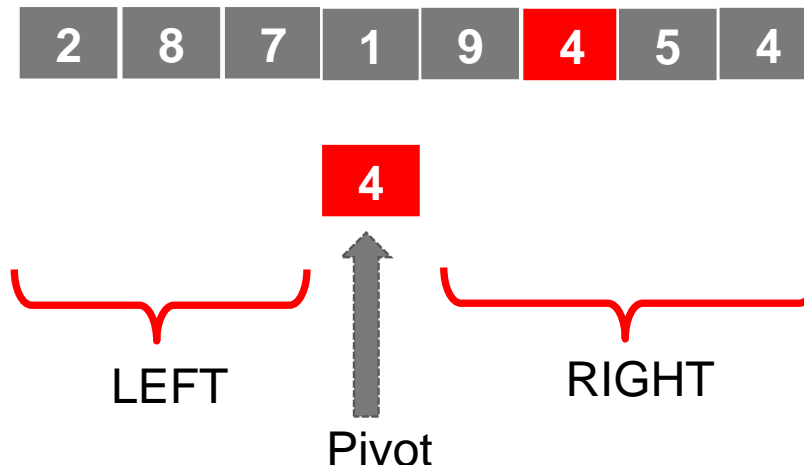
This is clearly not in-place.
Will this result in stable sorting?



Partitioning: An out-of-place version

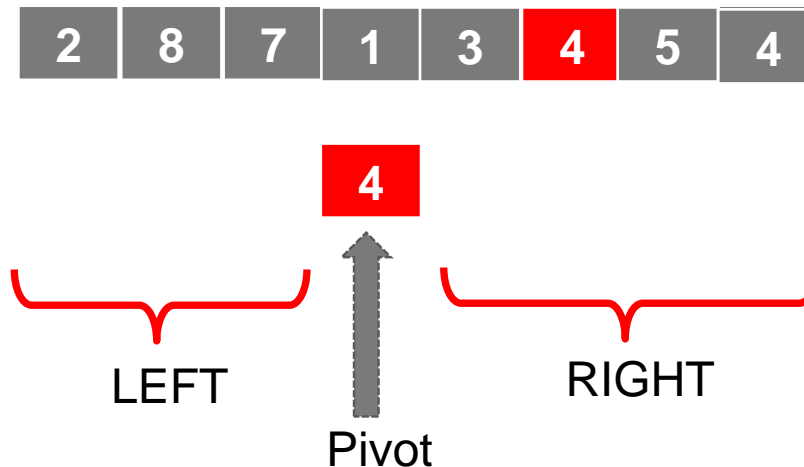
- Initialize two lists LEFT and RIGHT
- For each element e (except pivot)
 - If $e \leq \text{pivot}$
 - ✦ Insert e in LEFT
 - If $e > \text{pivot}$
 - ✦ Insert e in RIGHT
- Copy {LEFT, pivot, RIGHT} to the array

This version is unstable but it can be made stable!



Partitioning: A stable version

- Initialize two lists LEFT and RIGHT
- For each element e (except pivot)
 - If $e \leq \text{pivot}$
 - ✦ If $e == \text{pivot}$ and $e.\text{index} > \text{pivot.index}$
 - Insert e in RIGHT
 - ✦ Else
 - Insert e in LEFT
 - If $e > \text{pivot}$
 - ✦ Insert e in RIGHT
- Copy {LEFT, pivot, RIGHT} to the array



In-Place Partitioning

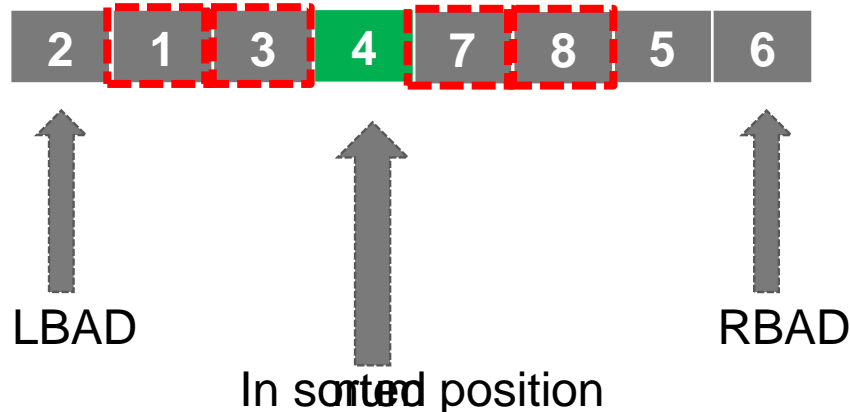
- $\text{num} \leftarrow$ the number of elements smaller than or equal to pivot $O(N)$

- Swap pivot with element at num

- Repeat until no bad element is found

- Find a bad element (LBAD) on the L.H.S. of pivot
- Find a bad element (RBAD) on the R.H.S. of pivot
- Swap LBAD and RBAD

$O(N)$



This partitioning algorithm is in-place but results in unstable sorting.

In-Place Partitioning (Improved)

- Swap pivot with the right most element
- LBAD points to the left most element
- RBAD points to the second right most element
- Repeat until LBAD and RBAD point to the same element
 - Move LBAD towards right until it points to an element $e > \text{pivot}$
 - Move RBAD towards left until it points to an element $e \leq \text{pivot}$
 - Swap elements pointed by LBAD and RBAD
- Swap pivot with the element pointed by LBAD/RBAD



↑
LBAD

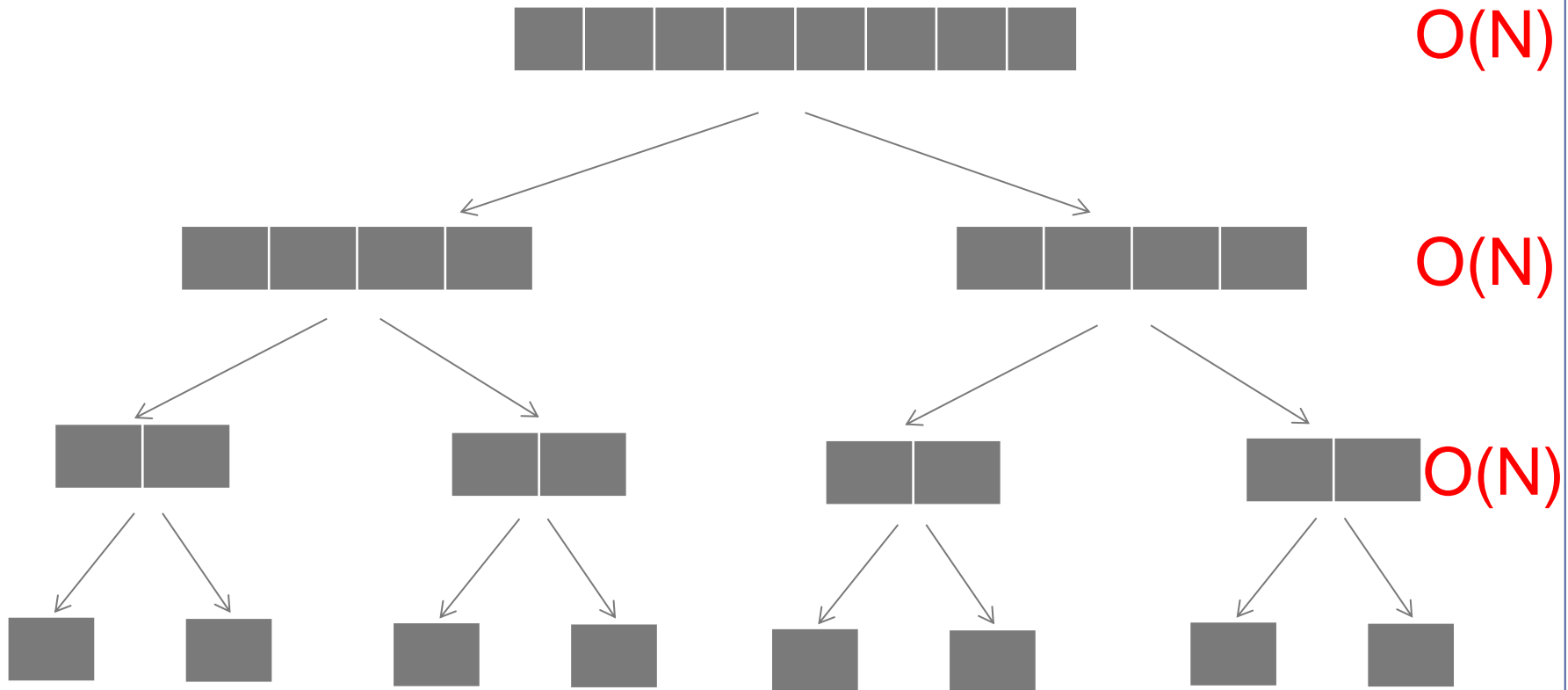


In sorted position

↑
RBAD

This partitioning is in-place
but unstable.

Best-case complexity of Quicksort

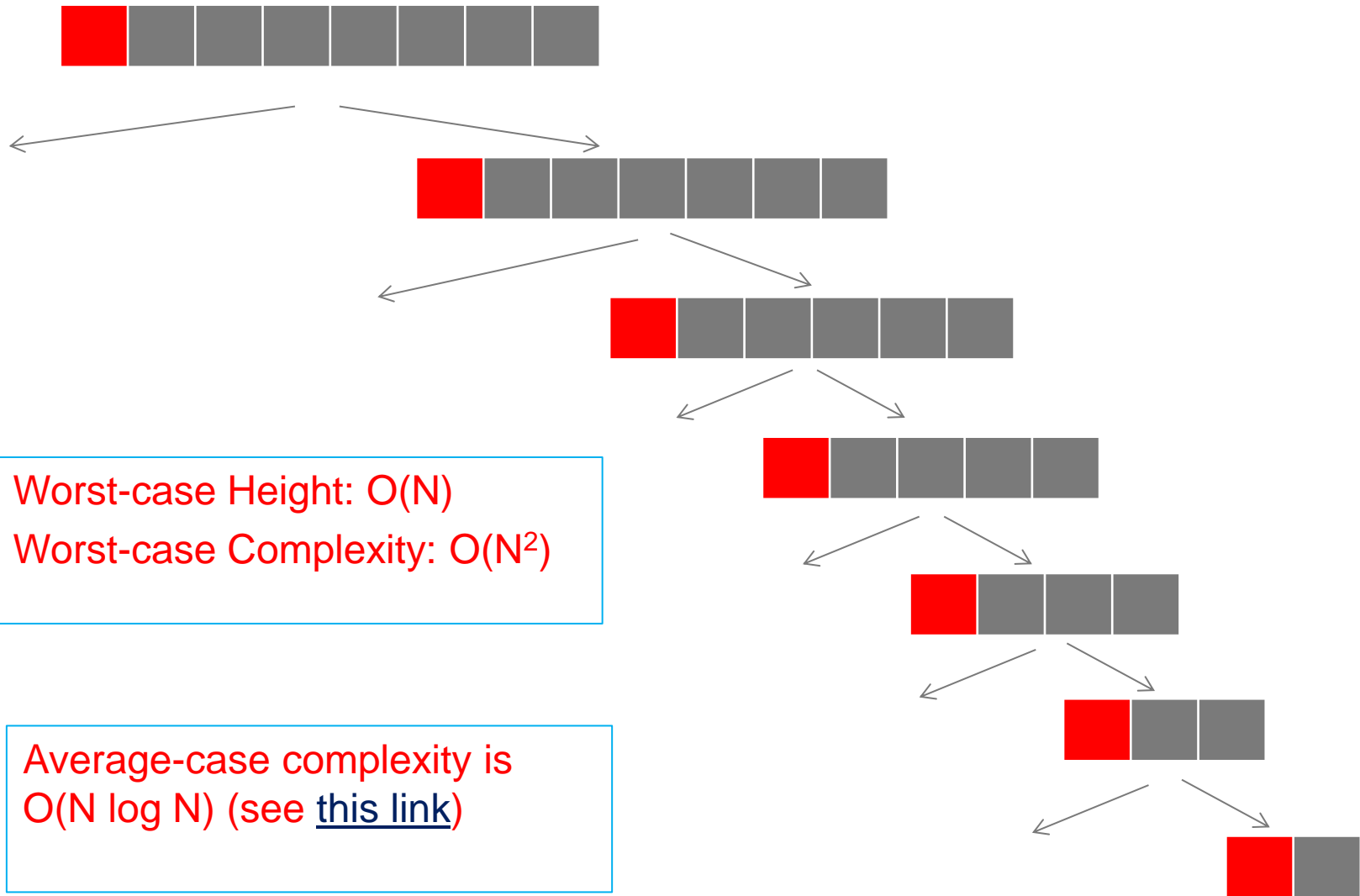


Best-case Height: $O(\log N)$
Best-case complexity: $O(N \log N)$

Important: Quicksort is not in-place even when in-place partitioning is used. Why?

Requires $O(\log N)$ space for recursion

Worst-case Complexity of Quicksort



Summary of complexities

	Best	Worst	Average	Stable?	In-place?
Selection Sort	$O(N^2)$	$O(N^2)$	$O(N^2)$	No	Yes
Insertion Sort	$O(N)$	$O(N^2)$	$O(N^2)$	Yes	Yes
Heap Sort	$O(N \log N)$	$O(N \log N)$	$O(N \log N)$	No	Yes
Merge Sort	$O(N \log N)$	$O(N \log N)$	$O(N \log N)$	Yes	No
Quick Sort	$O(N \log N)$	$O(N^2)$	$O(N \log N)$	Depends	No

Is it possible to develop a sorting algorithm with worst-case time complexity better than $O(N \log N)$?

Lower Bound Complexity

- Lower bound complexity for a **problem** is the lowest possible complexity **any** algorithm (known or unknown) can achieve to solve the problem
- What is the lower bound complexity of finding the minimum element in an array of N elements
 - **Ans: $O(N)$**
 - Since the algorithm we saw earlier has $O(N)$ complexity, it is optimal
- What is the lower bound complexity for sorting?
 - For comparison-based algorithm, lower bound complexity is $O(N \log N)$.
 - Read <https://www.cs.cmu.edu/~avrim/451f11/lectures/lect0913.pdf> to see why the lower bound is $O(N \log N)$

See you next week 😊