

1 2 3 17 19 36 7 25 100

Objective

- Understand heap and its types (min-heap and max-heap)
- Implement heap data structure and operations
- Implement heap sort
- Use python heapq module
- Identify problems that can be efficiently solved using heap





Prerequisites

- Arrays
- Tree





Lecture Flow

- 1. Motivation problem
- 2. Definition
- 3. Types
- 4. Representation
- 5. Operations
- 6. Time and space complexity
- 7. Practice

- 8. Heap construction
- 9. Heap sort
- 10. Heapq module
- 11. Problem patterns
- 12. Real world applications
- 13. Common pitfalls
- 14. Quote of the day







Motivation Problem



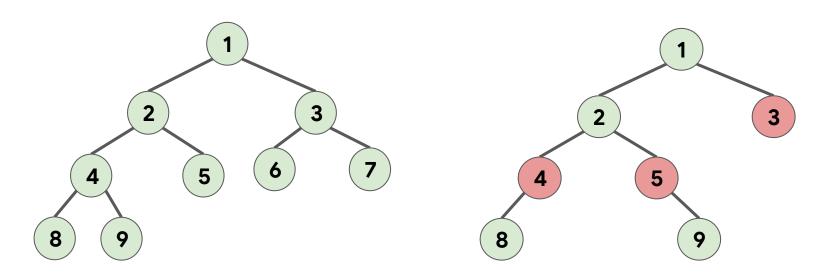
Definition







A binary heap is a complete binary tree that satisfies the heap order property.



Complete binary tree

Not complete binary tree







The heap order property is a key characteristic that defines the structure of a heap.

Max heap order property

Value of node >= Value of all its descendants

Min heap order property

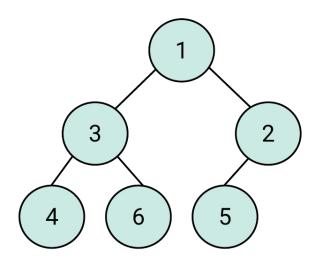
Value of node <= Value of all its descendants

For all nodes!

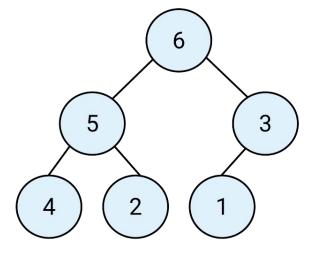


Types of Heaps





Min heap



Max heap



Representations







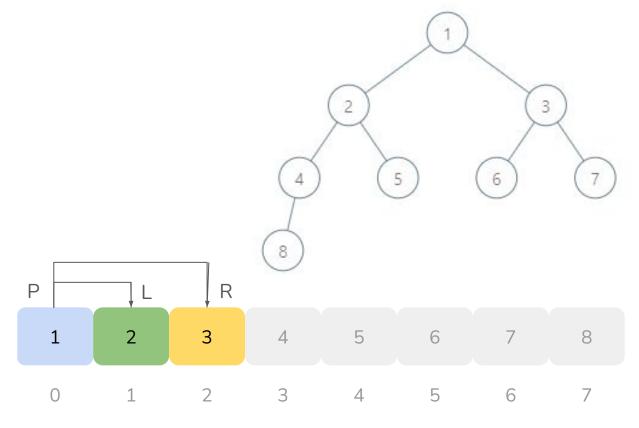
Binary heaps can be efficiently implemented using a binary tree structure, allowing for logarithmic time complexity for different heap operations.

Trees can be represented using arrays which makes implementations of heap operations easier.

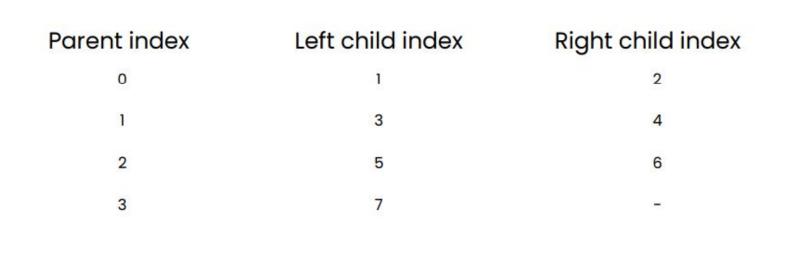


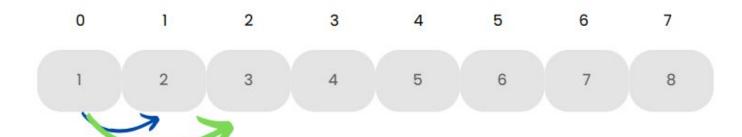
Representing Binary Tree Using Array



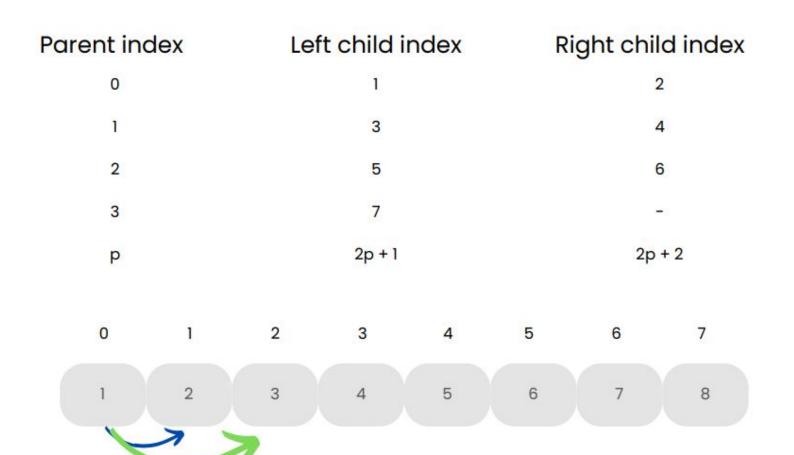












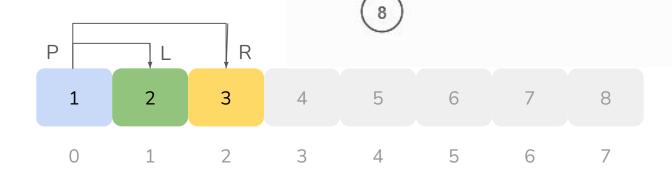


Representing Binary Tree Using Array

parent_idx = (child_idx - 1) // 2

left_child_idx = 2 * parent_idx + 1

right_child_idx = 2 * parent_idx + 2





Operations

In this lecture, we are going to focus on **min heaps**. However, the concepts discussed can seamlessly be extended to **max heaps** as well.



Insertion

To insert new value to min heap

- 1. Add it to the binary tree, however, we need to maintain the complete binary tree property, thus either place it
 - a. beyond the rightmost node at the bottom level of the tree, or
 - b. as the leftmost position of a new level, if the bottom level is already full (or if the heap is empty).
- After this action, the tree is complete, but it may violate the heap-order property.
- 3. Fulfill the heap order property. **How?**



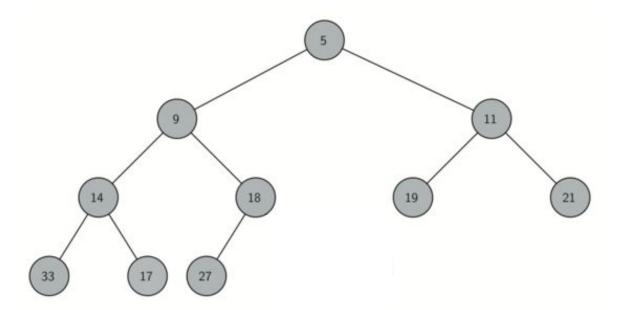
Up-heap Bubbling After Insertion

Compare the **newly** added value (Kn) with that of the **parent** (Kp)

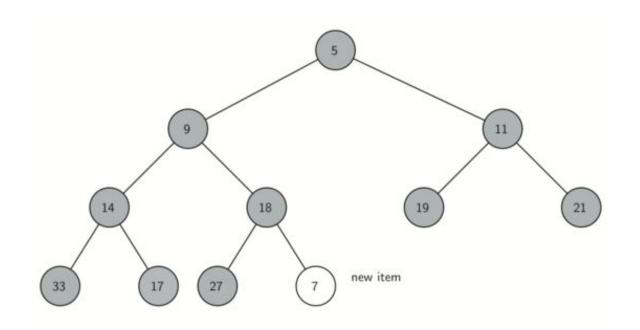
- If Kn >= Kp, the min heap order property is satisfied, the algorithm terminates
- 2. If Kn < Kp, Violation!
 - a. Swap child and parent, This swap causes the new item to **move up one level**. Hence the **up heap bubbling**.
 - Again the heap order property may be violated, so we repeat the process, going up in tree until no violation of the heap-order property occurs.



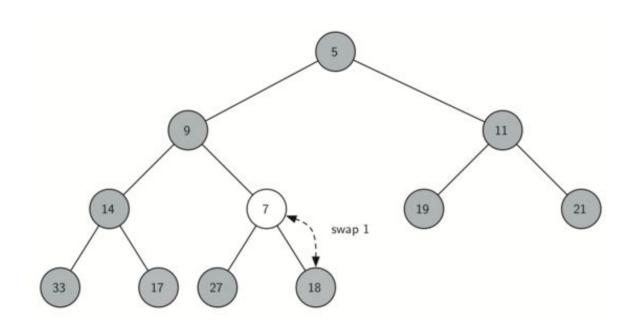
Let's insert 7 to the following heap



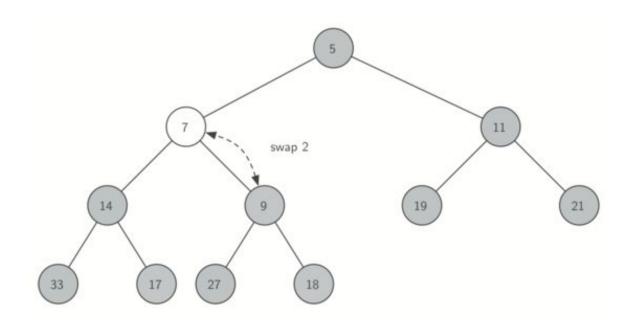














Time & Space Complexity of Insertion

Heapify up

- Time complexity ____
- Space complexity ____



Time & Space Complexity

Heapify up

- Time complexity O(log n)
- Space complexity O(1)

Why?



Can you implement heapify up (up-heap bubbling)?

Playground



Solution

```
def parent(j):
    return (j - 1) // 2
def swap(heap,i, j):
    heap[i], heap[j] = heap[j], heap[i]
def heappush(heap,value):
    heap.append(value)
    current = len(heap) - 1
    heapify up(heap, current)
def heapify up(heap, j):
    p = parent(j)
    if j > 0 and heap[j] < heap[p]:
         swap (heap, j, p)
         heapify up(heap, p)
```



Removing Smallest Element

The most straightforward element to remove from a tree is the **leaf**, while in an array, it corresponds to the **last element**.

Let's take advantage of that!

- 1. Swap the root of the heap (first element in our array representation) with the last element in the array.
- 2. Remove the last element
- The tree remains complete binary tree, however, it is very likely to violate min heap order property
- 4. Fulfill the min heap order property. **How?**



Down-heap Bubbling After Removal of Root Element

Let the **root** value be **(Kr)**, we can face the following two cases:

- 1. The root only has left child, let's denote the value of left child with **Kc**
- 2. The root has two children, let's denote the minimum value of the two by **Kc**i.e **Kc** = min(Kleft, Kright) ... min value of the left and right child



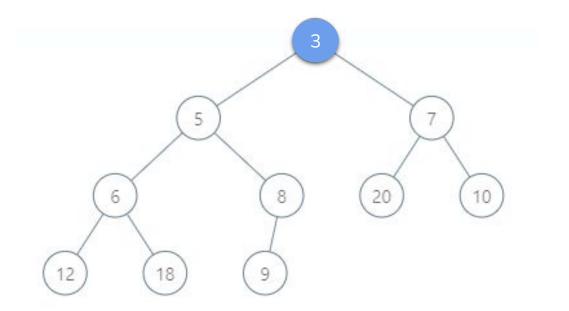
Continued...

Then,

- 1. If $Kr \le Kc$, the heap order property is satisfied, the algorithm terminates
- 2. If Kr > Kc, Violation!
 - a. Swap the smaller child with the parent
 - b. Continue swapping down tree until no violation of the heap-order property occurs

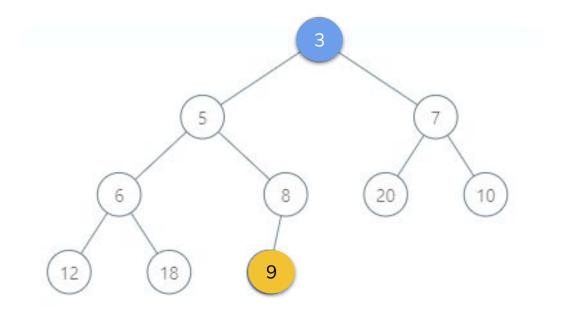


Let's simulate removal of **3** from the following heap



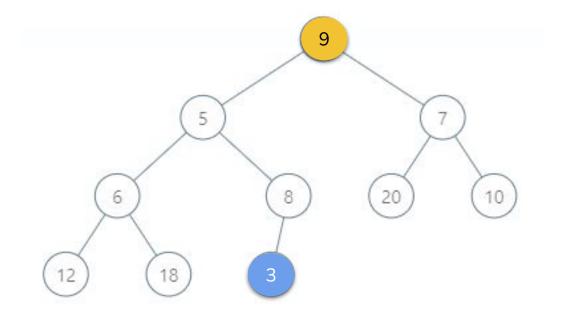


Swap root with the last element



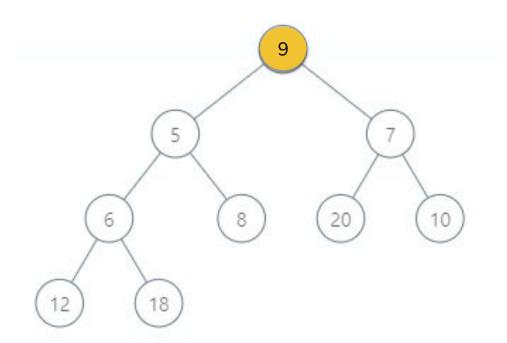


Remove the last element



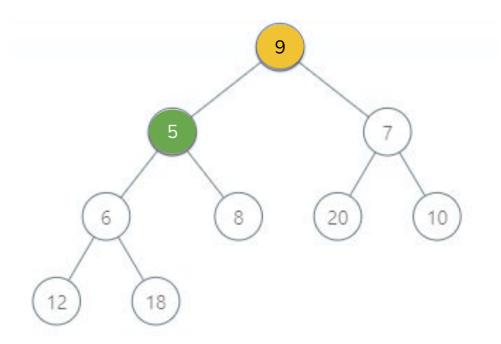


We have removed the minimum element, however, the heap order is violated.

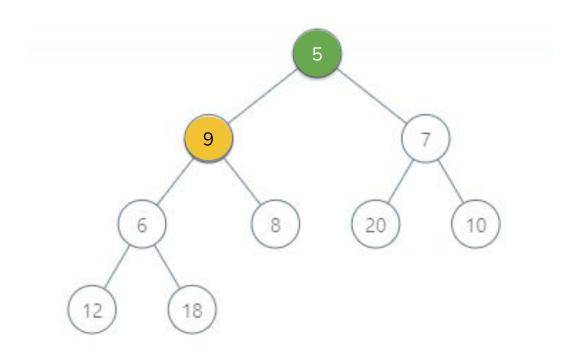




Our current node, the root, violates heap order. Thus, promote the smaller child

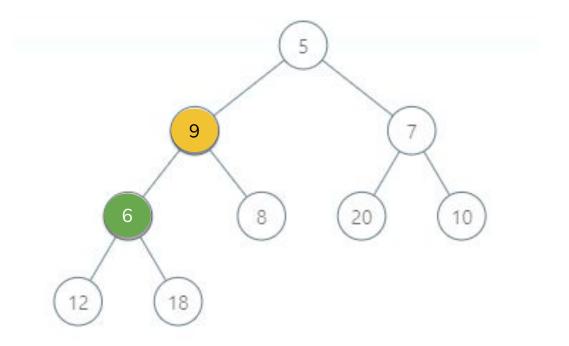






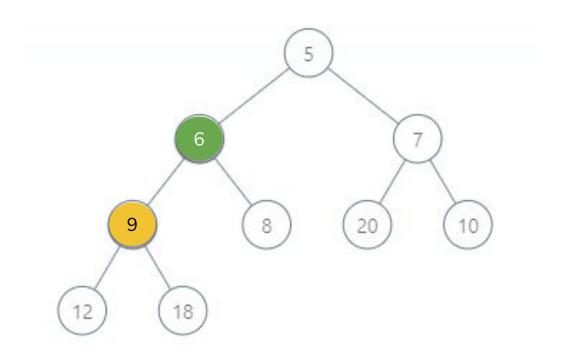


Our current node, still violates heap order. Thus, promote the smaller child





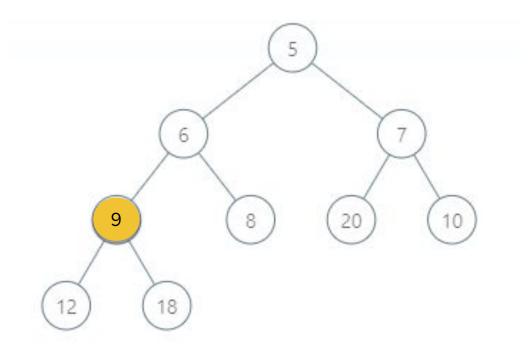
Removal & Down-heap Bubbling Simulation





Removal & Down-heap Bubbling Simulation

Our current node, does not violate heap order. We are done!





Time & Space Complexity

Heapify down

- Time complexity ____
- Space complexity ____



Time & Space Complexity

Heapify down

- Time complexity O(log n)
- Space complexity O(1)

Why?



Can you implement heapify down?

Playground



Solution

```
def heapify down(heap, n, current idx):
    small child idx = current idx
    left idx = 2 * current idx + 1
    right idx = 2 * current idx + 2
    if left idx < n and heap[left idx] < heap[small child idx]:</pre>
         small child idx = left idx
    if right idx < n and heap[right idx] < heap[small child idx]:</pre>
         small child idx = right idx
    if small child idx != current idx:
         swap(heap, current idx, small child idx)
        heapify down(heap, n, small child idx)
```



Heap Construction



Method 1

Start with empty heap and insert one element at a time

Time Complexity

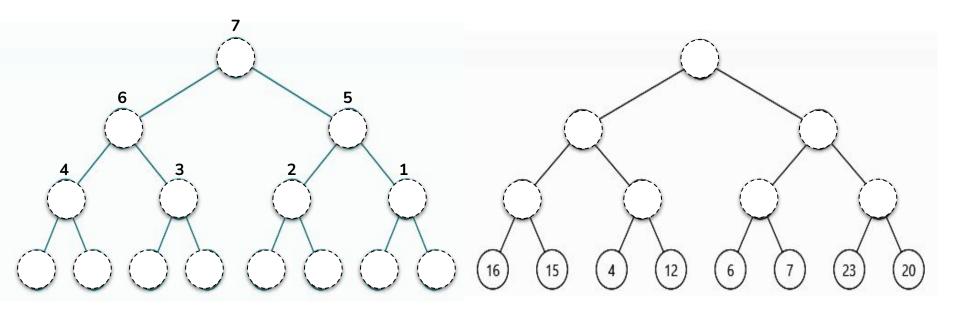
Time complexity - O(n log n)



Method 2 - Bottom up construction

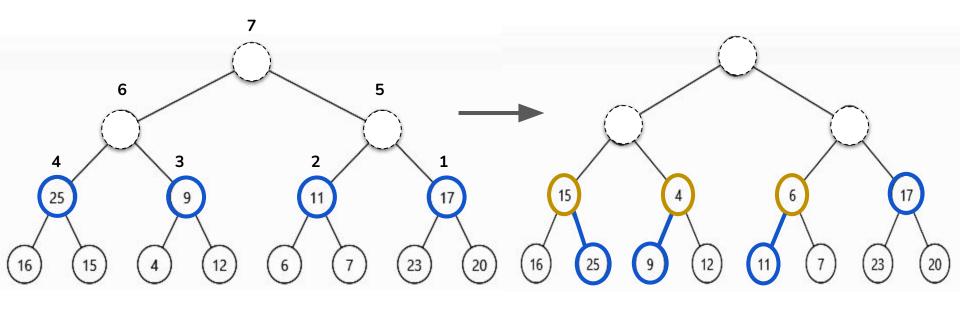
- We start with the last non-leaf node.
- We then apply the heapify_down algorithm to this node,
- We then move to the previous non-leaf node and apply the heapify_down algorithm to it as well.
- We continue this process, moving up the tree in reverse order, until we reach the root node.



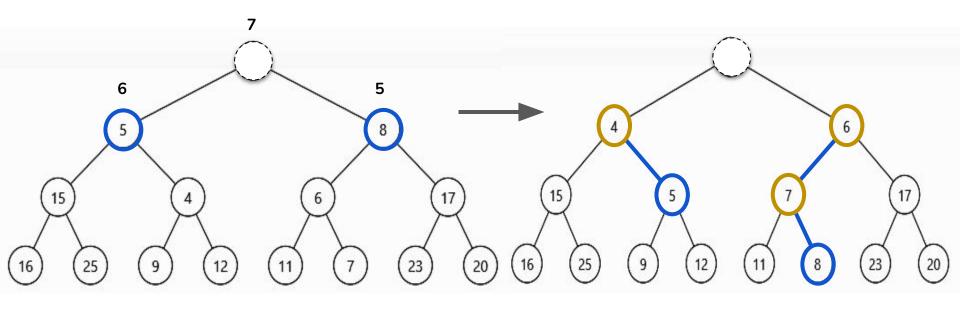


The sequence of numbers from 1 to 7 denotes the order in which we should invoke the heapify_down operation, i.e starting from the last non-leaf node.

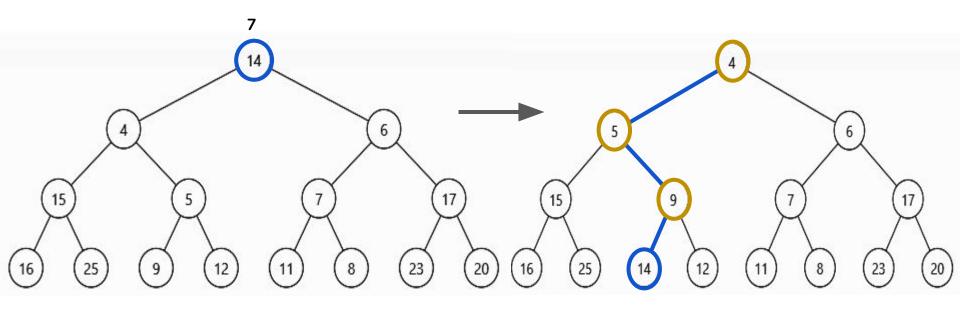














Time & Space Complexity

- Time complexity ____
- Space complexity ____



Time & Space Complexity

- Time complexity O(n)
- Space complexity O(1)

Why?

Time complexity proof



Can you implement heapify?

```
def heapify(array):
    # your code goes here
```



Solution

```
def heapify(array):
    n = len(arr)

for i in range(n // 2 - 1, -1, -1):
    heapify_down(arr, n, i)
```



Heap Sort



What is Heap Sort?

Heap sort is a **comparison-based** sorting technique based on binary heap data structure.

Steps:

- 1. We first find the maximum element and place it at the end.
- 2. Repeat the same process for the remaining elements.

Implement heap sort



Can you implement heap sort in place?



Let's sort the following array,

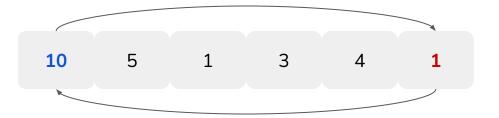
[1, 4, 10, 3, 5, 1]

Heapify (max)

[10, 5, 1, 3, 4, 1]



Swapping max element (root) with last element





But don't pop it from the list

1 5 1 3 4 **10**



heapify_down on the root element but our heap is now just the first n - 1
elements



And that is the use of **n** in our **heapify_down** implementation

```
def heapify down(heap, n, current idx):
```

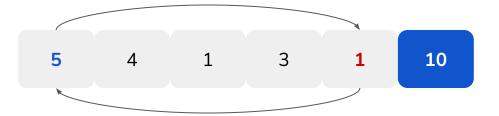


After heapifying the $\bf n$ - $\bf 1$ elements

5 4 1 3 1 10

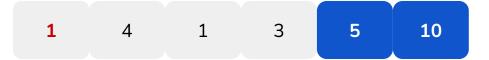


Swap max element (root) with last element





heapify_down on the root element but our heap now has n - 2 elements



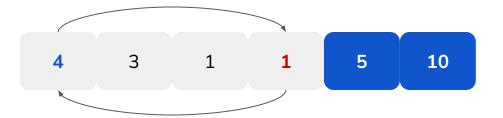


After heapifying the \mathbf{n} - $\mathbf{2}$ elements

4 3 1 1 5 10



Swap





Then heapify the n - 3 elements

1 3 1 **4 5 10**







Swap





Then heapify **n - 4** elements and swap





Then heapify **n** - **5** elements and swap





We are done!





Time & Space Complexity

- Time complexity O(n log n)
- Space complexity **O(1)**





Advantages of heap sort

- Efficiency O(n log n)
- Memory usage O(1)
- Simplicity
- Can be used in hybrid algorithms like the IntroSort.
- Sorting a nearly sorted (or K sorted) array.
- Finding K-largest (or smallest) elements in an array.



Disadvantage of heap sort

- Unstable
- Scalability not very efficient when working with highly complex data.





The Python **heapq module** offers an implementation of the operations we covered for a **min heap**, and some more functionalities.

- heapify(iterable): this function transforms the iterable into a heap in-place
- heappush(heap, item): this function pushes the item onto the heap while maintaining the heap invariant.
- heappop(heap): This function pops and returns the smallest item from the heap while maintaining the heap invariant.



- heapreplace(heap, item): this function pops and returns the smallest item from the heap, and then pushes the new item onto the heap. It is more efficient than calling heappop() followed by heappush()
- heappushpop(heap, item): this function combines the functionality of heappush() and heappop(), pushing the item onto the heap and then popping and returning the smallest item. It can be more efficient than heappush() followed by heappop()



- nlargest(n, iterable): this function returns the n-largest elements from the iterable.
- nsmallest(n, iterable): this function returns the n-smallest elements from the iterable.



Problem Patterns

- A. Top K Pattern
- B. Merge K Sorted Pattern
- C. Two Heap Pattern



Top K Pattern

A common problem that involves finding the k largest or smallest elements from a collection of n elements



Merge K Sorted Pattern

A common problem that involves merging k sorted arrays or lists into a single sorted array or list.



Two Heaps Pattern

Solving problems that require

- Quick access to the smallest and largest elements
- Maintaining a fixed-size window of elements with efficient insertion and removal operations.



Real World Applications

- Heap sort
- Priority Queue
 - Process scheduling in operating systems
- Graph algorithms
 - Shortest path
 - Minimum spanning tree



Common Pitfalls

Popping from empty heap

```
heapq.heappop(heap)

if heap:
heapq.heappop(heap)

Bad

Good
```

Appending to a full heap (fixed size)



Check that we haven't

Practice Questions

<u>Last Stone Weight</u> <u>Top K Frequent Words</u>

Kth Largest Element in a Stream Furthest Building You Can Reach

Kth Largest Element in an Array Find Median from Data Stream

Kth Smallest Element in a Sorted Matrix Minimum Cost to Hire K Workers

<u>Single-Threaded CPU</u> <u>Merge k Sorted Lists</u>

<u>Find K Pairs with Smallest Sums</u>
<u>The Skyline Problem</u>

<u>Top K Frequent Elements</u> <u>Heap Operations</u>



