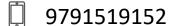


#### B.Bhuvaneswaran, AP (SG) / CSE



⊠ bhuvaneswaran@rajalakshmi.edu.in



## RAJALAKSHMI ENGINEERING COLLEGE

An AUTONOMOUS Institution
Affiliated to ANNA UNIVERSITY, Chennai

#### Heaps

- A heap is a data structure that is an implementation of the priority queue.
- Note that a priority queue is an abstract data structure.
- A heap is one of many ways to implement a priority queue.
- However, people often use the two terms interchangeably.

#### Heaps

- A heap is a container that stores elements, and supports the following operations:
  - Add an element in O(logn)
  - Remove the minimum element in O(logn)
  - Find the minimum element in O(1)

#### Note

- A heap can also find the max elements instead of the min elements.
- If a heap is configured to find/remove the min element, it's called a min heap.
- If it's configured to find/remove the max element, it's called a max heap.

#### Heaps

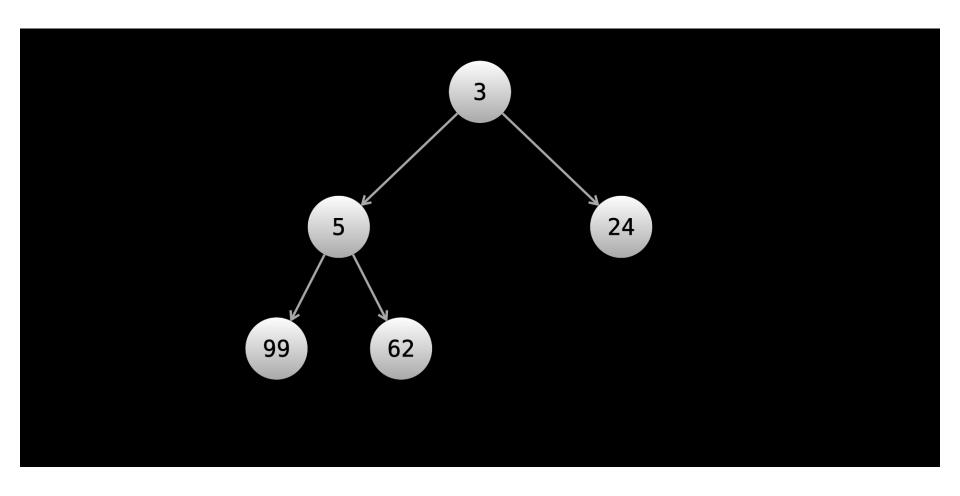
The ability to find the max/min element in constant time, while only needing logarithmic time to maintain this ability through changes makes a heap an extremely powerful data structure.

- Like a hash map, all major programming languages will have support for a heap, so you don't need to implement it yourself.
- In terms of solving algorithm problems, you only really care about the interface, not how it is implemented.
- But like with hash maps, it's still good to understand the implementation in case you are asked about it in an interview.

 There are multiple ways to implement a heap, although the most popular way is called a binary heap using an array.

- A binary heap implements a binary tree, but with only an array.
- The idea is that each element in the array is a node in the tree.
- The smallest element in the tree is the root, and the following property is maintained at every node:
  - if A is the parent of B, then A.val <= B.val.</li>
- Notice that this property directly implies that the root is the smallest element.
- Another constraint is that the tree must be a complete tree.

- The parent-child relationships are done using math with the indices.
- The first element at index 0 is the root, then the elements at indices 1 and 2 are the root's children, the elements at indices 3 and 4 are the children of the element at index 1 and the elements at indices 5 and 6 are the children of the element at index 2, and so on. If a node is at index i, then its children are at indices 2i + 1 and 2i + 2.
- When elements are added or removed, operations are done to maintain the aforementioned property of parent.val <= child.val.</li>
- The number of operations needed scales logarithmically with the number of elements in the heap, and the process is known as "bubbling up".



#### Note

- An existing array of elements can also be converted into a heap in linear time, although the process is complicated.
- Luckily, some major programming languages have built-in methods to do this.
- Remember: you shouldn't worry too much about how heaps are implemented. The important thing is that you understand the interface.

#### Note

- In many problems, using a heap can improve an algorithm's time complexity from  $O(n^2)$  to  $O(n \log n)$ , which is a massive improvement (for n = 1,000,000, this is 50,000 times faster).
- A heap is a great option whenever you need to find the maximum or minimum of something repeatedly.

### Interface guide

- Note: some languages implement a min heap by default, while some implement a max heap by default.
- If you're dealing with numbers and you want to deal with the opposite type of heap that your language implements, an easy way to do this is to multiply all numbers by -1

#### Interface guide

```
// In Java, we will use the PriorityQueue interface and the
// PriorityQueue implementation. By default, this implements
// a min heap
PriorityQueue<Integer> heap = new PriorityQueue<>();
// Add to heap
heap.add(1);
heap.add(2);
heap.add(3);
```

## Interface guide

```
// Check minimum element
heap.peek(); // 1
// Pop minimum element
heap.remove(); // 1
// Get size
heap.size(); // 2
// Bonus: if you want a max heap instead, you can pass
// Comparator.reverseOrder() to the constructor:
PriorityQueue<Integer> maxHeap = new
PriorityQueue<>(Comparator.reverseOrder());
```

## Heap Examples

 A heap is an amazing tool whenever you need to repeatedly find the maximum or minimum element.

## Last Stone Weight

- You are given an array of integers stones where stones[i] is the weight of the i<sup>th</sup> stone.
- On each turn, we choose the heaviest two stones and smash them together. Suppose the heaviest two stones have weights x and y with x <= y.</li>
- If x == y, then both stones are destroyed.
- If x != y, then x is destroyed and y loses x weight.
- Return the weight of the last remaining stone, or 0 if there are no stones left.

## Example

- Input:
  - stones = [2, 7, 4, 1, 8, 1]
- Output:
  - 1

#### Minimum Operations to Halve Array Sum

- You are given an array nums of positive integers.
- In one operation, you can choose any number from nums and reduce it to exactly half the number.
- Return the minimum number of operations to reduce the sum of nums by at least half.

## Example

- Input:
  - nums = [5, 19, 8, 1]
- Output:
  - 3

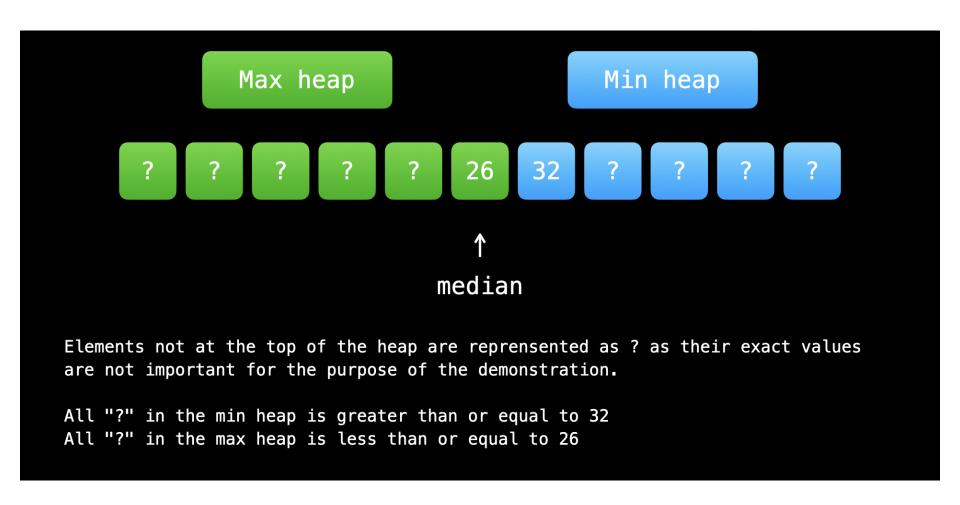
### Two heaps

- Using multiple heaps is uncommon and the problems that require it are generally on the harder side.
- If a problem involves finding a median, it's a good thing to think about.

#### Find Median from Data Stream

- The median is the middle value in an ordered integer list.
- If the size of the list is even, the median is the average of the two middle values. Implement the MedianFinder class:
- MedianFinder() initializes the MedianFinder object.
- void addNum(int num) adds the integer num to the data structure.
- double findMedian() returns the median of all elements so far.

#### Example



#### Code

```
class MedianFinder {
  private PriorityQueue<Integer> minHeap = new PriorityQueue<>();
  private PriorityQueue<Integer> maxHeap = new PriorityQueue<>(Comparator.reverseOrder());;
  public void addNum(int num) {
    maxHeap.add(num);
    minHeap.add(maxHeap.remove());
    if (minHeap.size() > maxHeap.size()) {
      maxHeap.add(minHeap.remove());
  public double findMedian() {
    if (maxHeap.size() > minHeap.size()) {
      return maxHeap.peek();
    return (minHeap.peek() + maxHeap.peek()) / 2.0;
```

# Queries?

# Thank You...!