

## ADT: Priority Queue

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- The following operations:
  - find element with highest priority
  - delete element with highest priority
  - insert element with assigned priority
- Many implementations possible:
  - Sorted list
  - Binary search tree
  - ...

## Binary Heap Implementation

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- Binary Heap data structure
  - Find –  $O(1)$
  - Delete and insert –  $O(\log n)$
- Enhance with “change priority”
  - Delete a given element
  - Change priority for a given element –need for some important applications
- Goal: Find the element in the heap in constant time! Maintain an additional array for all the entities of interest e.g. vertices in a graph, that contains their position in the (heap) priority queue (handle)

## Binary Heap

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### Operations

- BinaryHeap() creates a new, empty, binary heap.
- insert(k) adds a new item to the heap.
- find\_min() returns the item with the minimum key value, leaving item in the heap.
- del\_min() returns the item with the minimum key value, removing the item from the heap.
- is\_empty() returns true if the heap is empty, false otherwise.
- size() returns the number of items in the heap.
- Build\_heap(list) builds a new heap from a list of keys.

## Heaps and Heapsort

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- Definition A heap is a binary tree with keys at its nodes (one key per node) such that:
- It is essentially complete (note online text is slightly different), i.e., all its levels are full except possibly the last level, where only some rightmost keys may be missing – **shape property**.



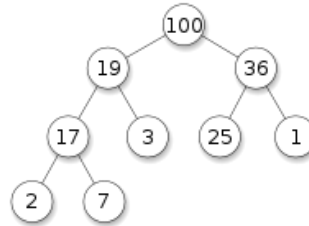
- The key at each node is  $\leq$  keys at its children (MinHeap)–  
**heap order (structure) property** ( $\geq$  MaxHeap)

## Illustration of the heap's definition

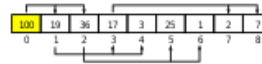
### Max Heap

Note: Heap's elements are ordered top down (along any path down from its root), but they are not ordered left to right

### Tree representation



### Array representation

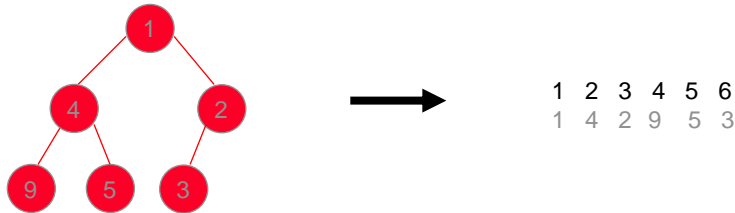


## Some Important Properties of a Heap (MaxHeap)

- Given  $n$ , there exists a unique binary tree with  $n$  nodes that is essentially complete, with  $h = \lfloor \log_2 n \rfloor$
- The root contains the largest key
- The subtree rooted at any node of a heap is also a heap
- A heap can be represented as an array

## Heap's Array Representation

- Store heap's elements in an array (whose elements indexed, for convenience, 1 to  $n$ ) in top-down left-to-right order
- Example:



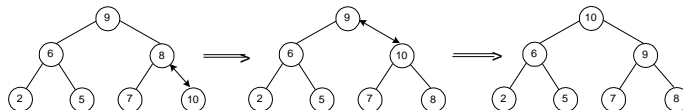
- Left child of node  $j$  is at  $2j$       Right child of node  $j$  is at  $2j+1$
- Parent of node  $j$  is at  $\lfloor j/2 \rfloor$
- Parental nodes are represented in the first  $\lfloor n/2 \rfloor$  locations

## Insertion of a New Element into a Heap

- Insert the new element at last position in heap.
- Compare it with its parent and, if it violates heap condition, exchange them (Drift up)
- Continue comparing the new element with nodes up the tree until the heap condition is satisfied

Example: Insert key 10

Efficiency:  $O(\log n)$



## Insertion into heap: perc\_up (drift\_up, sift\_up)

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- 1: Put new element into first open position, this maintains the structure property

```
def insert(self,k):
    self.heapList.append(k)
    self.currentSize = self.currentSize + 1
    self.percUp(self.currentSize)
```

- 2: Drift the element up until the heap property is restored

```
def percUp(self,i):
    while i // 2 > 0:
        if self.heapList[i] < self.heapList[i // 2]:
            tmp = self.heapList[i // 2]
            self.heapList[i // 2] = self.heapList[i]
            self.heapList[i] = tmp
        i = i // 2
```

## Top-down heap construction

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- Start with empty heap and repeatedly insert elements
- Performance  $O(n \log n)$

## Remove max from heap

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IDEA:

1. Store root (maximum element) for return
2. Swap last entry in heap with first entry in heap
3. Reduce heapsize by 1
4. *perc\_down* the new root (first element) until the heap property is restored

Note: There are only two basic “moves” in the heap.

- *perc\_down* – used in bottom up construction and removal
- *perc\_up* – used in insertion and top-down construction

## perc\_down

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```
def percDown(self, pos):
    while (pos * 2) <= self.currentSize:
        minchd = self.minChild(pos) // index of the min child
        if self.heapList[i] > self.heapList[minchd]: // if min child smaller
            tmp = self.heapList[i] // swap
            self.heapList[i] = self.heapList[minchd]
            self.heapList[minchd] = tmp
        pos = minchd
```

## min\_child

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```
def minChild(self,pos):
    if pos * 2 + 1 > self.currentSize:    // if no right child then return left child pos
        return pos * 2
    else:
        if self.heapList[pos * 2] < self.heapList[pos * 2 + 1]:
            return pos * 2
        else:
            return pos * 2 + 1
```

## Bottom-up heap construction

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1. Insert elements into array respecting the shape property.  
Tree is essentially complete.
2. Rearrange elements to enforce the heap order (structure) property

## Heap Construction (bottom-up)

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### High level pseudo-code

Initialize the array structure with keys in the order given  
(shape property)

Loop: node = rightmost parental node to root  
if node doesn't satisfy the heap condition:  
loop: exchange it with its smallest child until the heap  
condition holds ("perc\_down")

## Bottom-up heap construction

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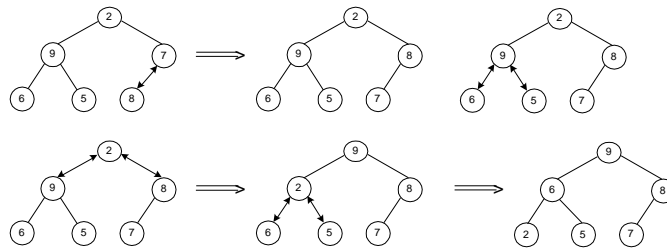
1. Insert elements into array respecting the shape property
2. Rearrange elements to enforce the heap order property

```
def buildHeap(self, alist):
    i = len(alist) // 2
    self.currentSize = len(alist)
    self.heapList = [0] + alist[:]
    while (i > 0):
        self.percDown(i)
        i = i - 1
```



## Example of Heap Construction

Construct a maxheap for the list 2, 9, 7, 6, 5, 8



## Which is better Top-down vs. Bottom-up heap construction?

- Bottom up is  $O(n)$ 
  - Most nodes are near the bottom of the tree, why
  - Almost all the nodes are leaves!
  - Driftdown drifts entries down, thus most nodes do not have far to travel
- Top down is  $O(n \log n)$ 
  - Half the nodes are near the bottom
  - Thus in the worst case, those  $n/2$  nodes may need to move to the top
  - Since the height of the tree is  $\log n$
  - That could require  $O(n \log n)$  comparisons

## Priority Queue

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- A priority queue is the ADT of a set of elements with numerical priorities and the following operations:
  - find element with highest priority
  - delete element with highest priority
  - insert element with assigned priority
- Heap is a very efficient way for implementing priority queues
- Applications determine what is a priority ordering!  
Sometimes want largest, sometimes smallest element to be found/deleted.
- Many implementation options : list, sorted list, ...

## Heapsort

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Construct a heap for a given list of  $n$  keys

Repeat operation of root removal  $n-1$  times:

- Exchange keys in the root and in the last (rightmost) leaf
- Decrease heap size by 1
- If necessary, swap new root with larger child until the heap condition holds (Drift Down)

## Example of Sorting by Heapsort

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Sort the list 2, 9, 7, 6, 5, 8 in ascending order using Max Heap

Stage 1 (bottom up heap construction)	Stage 2 (root/max removal)
2 9 7 6 5 8    7↔8 swap	9 6 8 2 5 7    swap 9,7
2 9 8 6 5 7    9↔6,5 no swap	7 6 8 2 5   9    ↓ size
2 9 8 6 5 7    2↔9,8 swap	8 6 7 2 5   9    swap 8,5
9 2 8 6 5 7    2↔6,5 swap	5 6 7 2   8 9    ↓ size
9 6 8 2 5 7    2 no children	7 6 5 2   8 9    ...
	2 6 5   7 8 9
	6 2 5   7 8 9
	5 2   6 7 8 9
	5 2   6 7 8 9
	2   5 6 7 8 9    sorted order