

## 14 Array-Based Heaps

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### Outline

#### 1 Complete Binary Trees

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#### 3 Analysis

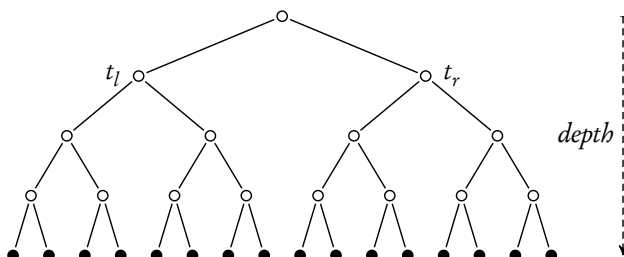
👁 Textbook §8.3.2, 9.3.3 – 9.3.7.

#### Complete Binary Trees

### Full Binary Trees

A full binary tree is

- either *empty*, or
- a binary tree whose two subtrees —  $t_l, t_r$  — are also full binary trees of the same size.



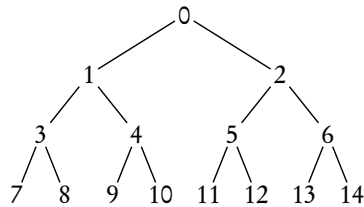
A full binary tree of depth  $d$  has size  $2^{d+1} - 1$ .



## Numbering Nodes in a Full Binary Tree

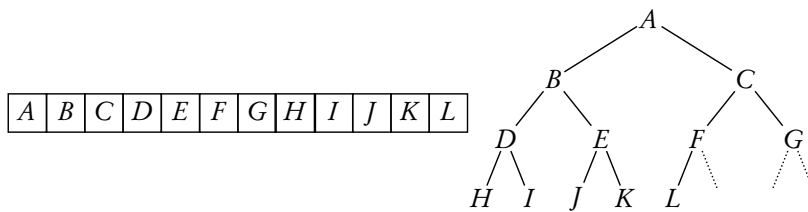
We number the nodes in a full binary tree from top to bottom, left to right.

- Given that the root is numbered 0, the left most node of depth  $d$  is numbered  $2^d - 1$ .
- The left child and right child of a node numbered  $i$  are numbered  $2i + 1$  and  $2i + 2$ .
- The parent of a node numbered  $i$  is numbered  $\left\lfloor \frac{i-1}{2} \right\rfloor$ .



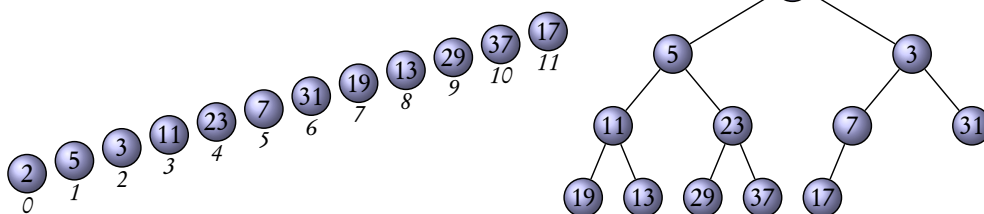
## Array-Based Complete Binary Trees

- We may store the nodes of a full binary tree in an array-based list, each taking a position according to their numbers, that is, the node numbered  $i$  is stored as element  $a[i]$  in list  $a$ .
- On the other hand, an array of elements can be structured as a full binary tree, if its size is  $2^{d+1} - 1$ . For an array of arbitrary size  $n$ , by removing those nodes with numbers greater than  $n - 1$  from the full binary tree, we have a *complete binary tree*.



## Array-Based Heaps

If a complete binary tree also has the heap property, then such a heap can be stored in an array-based list  $a$ . Obviously, the root  $a[0]$  contains the minimum element



```
1 class ArrayHeap:
2     def __init__(self):
3         self.a = []
```

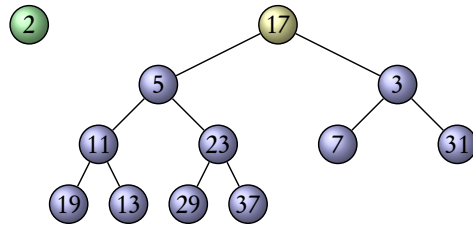
```
4     def __bool__(self):
5         return bool(self.a)
```



## Recovering the Heap Property by Sifting Down

If we want to remove the root, we need to relocate a node in the tree to the root, and we must recover the heap property.

- We can only detach the bottom-right most node  $x$ , in order to maintain the complete binary tree. This is the last element in the array-based list.
- We put  $x$  to the root, and sift it down to a proper location where the children are no less, maintaining the heap property.
- We must choose the least node among  $x$  and its two children at each step. This is in fact a rotation along some path.



## Sifting Down

The function `sift_down` takes a starting vacant position  $i$  and the element  $x$  to sift down, finds the sifting path and moves the elements along the path, finally puts  $x$  at the end position.

```

1 def sift_down(a, i, x):
2     n = len(a)
3     j = 2*i+1 # index of left child
4     if j < n: # at least a child exists
5         if j+1 < n and not a[j] <= a[j+1]: # right child exists and is smaller
6             j += 1 # index of right child
7         if not x <= a[j]: # x must be put down further
8             a[i] = a[j]
9             return sift_down(a, j, x)
10    a[i] = x

```



## The `pop_min` Method

We use `sift_down` to help recover the heap property after the deletion of the root in the `pop_min` method.

```

1 def pop_min(self):
2     x, last = self.a[0], self.a[-1]
3
4     del self.a[-1]
5     if self.a:
6         sift_down(self.a, 0, last)
7
8     return x

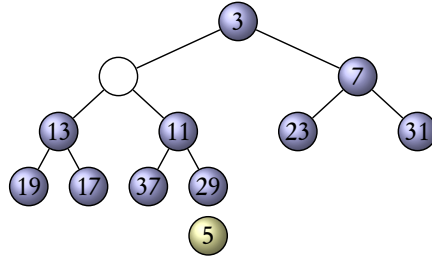
```



## The *push* Method and Sifting Up

We can only append an element  $x$  to the end of an array-based list — the bottom-right most of a heap — efficiently, we need to relocate it to recover the heap property.

- Such an append leaves no hole in the heap, maintaining the complete binary tree.
- We *sift up*  $x$  to a proper location where the parent is no greater, maintaining the heap property.



## The *sift\_up* Function and the *push* Method — Code

The *sift\_up* method rotates  $x$  with the ancestors greater than  $x$ . We don't need to check with the size of the heap, for the sifting-up goes towards the root, whose index is 0.

```

1 def sift_up(a, i, x):
2     if i > 0:
3         j = (i-1)//2 # index of parent
4         if not a[j] <= x:
5             a[i] = a[j]
6             return sift_up(a, j, x)
7     a[i] = x

```

```

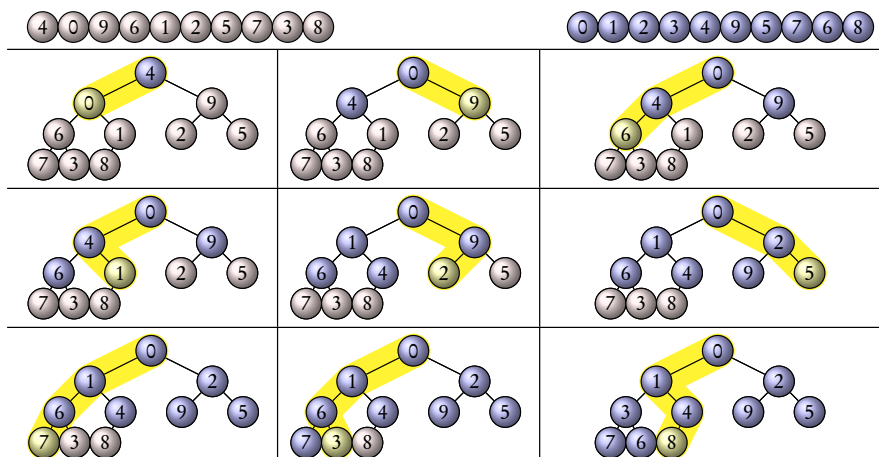
1 def push(self, x):
2     self.a.append(None)
3     sift_up(self.a, len(self.a)-1, x)

```

We insert an element by appending it to the heap and sifting it up.



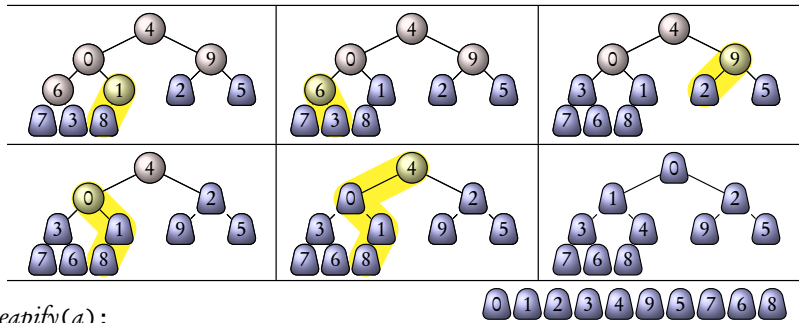
## Building a Heap — Heapify-ing — by Insertion





## Heapifying by Merging

We may even build the heap by sifting down, starting from the bottom up to the top. Sifting a node down can be regarded as merging the node with its two sub-heaps into a bigger heap.

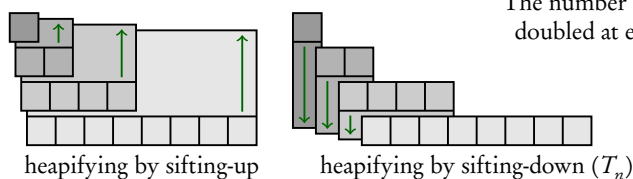


```
def heapify(a):
    for i in range((len(a)-2)//2, -1, -1):
        sift_down(a, i, a[i])
```



## Which Is Better — Up or Down?

- When heapifying by sifting-up, the deeper levels get the larger multipliers;
- When heapifying by sifting-down, the deeper levels get the smaller multipliers.



$$T_n = \frac{n}{4} + \frac{2n}{8} + \frac{3n}{16} + \frac{4n}{32} + \dots$$

$$= \left( \frac{n}{4} + \frac{n}{8} + \frac{n}{16} + \frac{n}{32} + \dots \right) + \left( \frac{n}{8} + \frac{2n}{16} + \frac{3n}{32} + \dots \right) = \frac{n}{2} + \frac{T_n}{2}.$$



## Analysis

For a heap of  $n$  elements, we only need a fix amount of auxiliary space for sifting up and sifting down.

- $\mathcal{O}(1)$  auxiliary space.

We count the number of element comparisons.

- A sifting up moves along a path from bottom to top, in each step, there is only one comparison with the parent.
- A sifting down moves along a path from top to bottom, in each step, there are two comparisons, one between the children, one with the selected child.

Since the maximum depth of a complete binary tree is  $d$  when the number of nodes is between  $2^d$  and  $2^{d+1} - 1$ , the *push* and *pop\_min* of a heap of size  $n$  all take only  $\mathcal{O}(\log n)$  time. For sifting-down heapification, the number of moves is at most

$$\frac{n}{4} + \frac{2n}{8} + \frac{3n}{16} + \dots = n \in \mathcal{O}(n).$$

