CSE 5311: Design and Analysis of Algorithms

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Introduction

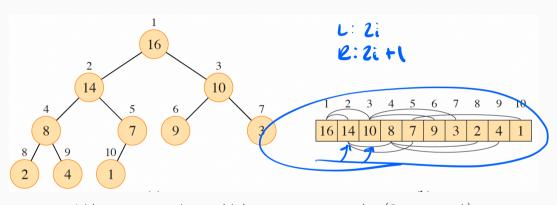
A binary heap can be represented as a binary tree, but is stored as an array.

- The root is the first element of the array.
- The left subnode for the element at index i is located at 2i and the right subnode is located at 2i + 1
- · This assumes a 1-based indexing.

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#### How would this change for 0-based indexing?

- 2i + 1 for the left.
- 2i + 2 for the right.
- The parent could be accessed via  $\lfloor \frac{i-1}{2} \rfloor$ .



A binary tree as a heap with its array representation (Cormen et al.).

Heaps come in two flavors: max-heaps and min-heaps.

They can be identified by satisfying a heap property.

- max-heap property:  $A[parent(i)] \ge A[i]$
- · min-heap property:  $A[parent(i)] \leq A[i]$

For sorting, a max-heap is used.

We will later study priority queues, where a **min-heap** is used.

# Maintaining the Heap Property

#### **Heap Property**

The heap should always satisfy the max-heap property.

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#### The heap should always satisfy the max-heap property.

- This relies on a procedure called max\_heapify.
- This assumes that the root element may violate the max-heap property, but...
- · Assumes subtrees rooted by its subnodes are valid max-heaps.
- Swap nodes down the tree until the misplaced element is in the correct position.

#### Heapify

```
def max_heapify(A, i, heap_size):
    l = left(i)
    r = right(i)
    largest = i
    if 1 < heap_size and A[1] > A[i]:
       largest = 1
    if r < heap size and A[r] > A[largest]:
        largest = r
    if largest != i:
        A[i], A[largest] = A[largest], A[i]
        max_heapify(A, largest, heap size)
```

Given that max\_heapify is a recursive function, we can analyze it with a recurrence.

- **Driving function:** the fix up that happens between the current node and its two subnodes:  $\Theta(1)$ .
- Recurrence: based on how many elements are in the subheap rooted at the current node.

#### Recurrence

- In the worst case of a binary tree, the last level of the tree is half full.
- The left subtree has height h + 1 compared to the right subtree's height of h.
- For a tree of size n, the left subtree has  $2^{h+2} 1$  nodes and the right subtree has  $2^{h+1} 1$  nodes.

The number of nodes in the tree is equal to  $1 + (2^{h+2} - 1) + (2^{h+1} - 1)$ .

$$n = 1 + 2^{h+2} - 1 + 2^{h+1} - 1$$

$$n = 2^{h+2} + 2^{h+1} - 1$$

$$n = 2^{h+1}(2+1) - 1$$

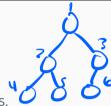
$$n = 3 \cdot 2^{h+1} - 1$$

- This implies that  $2^{h+1} = \frac{n+1}{3}$ .
- In the worst case, the left subtree would have  $2^{h+2} 1 = \frac{2(n+1)}{3} 1$  nodes which is bounded by  $\frac{2n}{3}$ .
- The recurrence for the worst case of max\_heapify is  $T(n) = T(\frac{2n}{3}) + O(1)$ .

Given an array of elements, how do we build the heap in the first place?

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Use a bottom-up approach from the leaves.



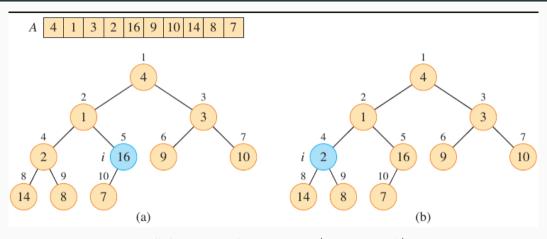
- The elements from  $\lfloor \frac{n}{2} \rfloor + 1$  to *n* are all leaves.
- This means that they are all 1-element heaps.
- Run max\_heapify on the remaining elements to build the heap.

```
def build_max_heap(A):
    heap_size = len(A)
    for i in range(len(A) // 2, -1, -1):
        max_heapify(A, i, heap_size)
```

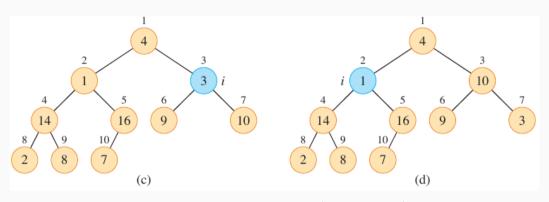
• Each node starting at  $\lfloor \frac{n}{2} \rfloor + 1$  is the root of a 1-element heap.

- Each node starting at  $\lfloor \frac{n}{2} \rfloor + 1$  is the root of a 1-element heap.
- The subnodes, which are to the right of node  $\lfloor \frac{n}{2} \rfloor$ , are roots of their own max-heaps.

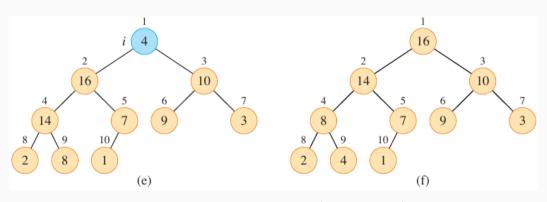
- Each node starting at  $\lfloor \frac{n}{2} \rfloor + 1$  is the root of a 1-element heap.
- The subnodes, which are to the right of node  $\lfloor \frac{n}{2} \rfloor$ , are roots of their own max-heaps.
- The procedure loops down to the first node until all sub-heaps have been max-heapified.



Building a heap from an array (Cormen et al.).



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## Analysis of Heapsort

The call to  ${\tt max\_heapify}$  is...

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The loop in  $build_max_heap runs...$ 

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The loop in build\_max\_heap runs... O(n) times.



#### def heapsort(A):

build\_max\_heap(A)

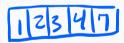
heap\_size = len(A)

for i in range(len(A) - 1, 0, -1):

A[O], A[i] = A[i], A[O]

heap\_size -= 1

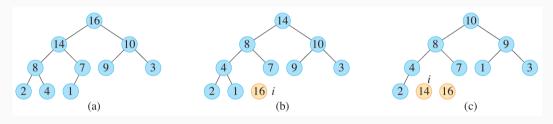
max\_heapify(A, 0, heap\_size)



$$\Theta(n | gn + n | gn) = \Theta(n | gn)$$

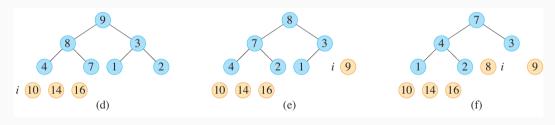
• Start by building a max-heap on the input array - O(n).

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- Take the root element out of the heap and run  $\max_{heapify}$  to maintain the max-heap property  $O(n \lg n)$ .



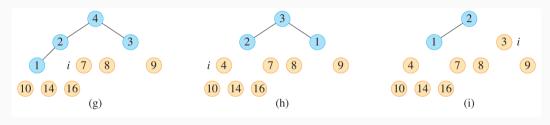
Example of Heapsort (Cormen et al.).

### Heapsort



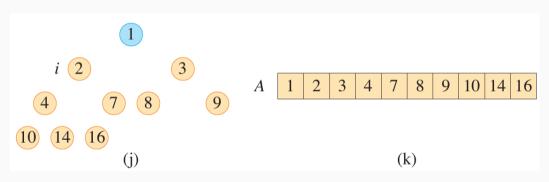
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What is it?

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- · A key-value data structure where each key has a priority.
- · The elements are processed as a queue.
- Implemented with either a min-heap or max-heap (depending on the use case).

### Operations

- Insert: Add a new element to the queue.
- Extract: Remove and return the element with the highest/lowest priority.
- Max/Min: Return the element with the highest/lowest priority without removing it.
- · Increase/Decrease Key: Change the priority of an element.

### **Priority Queue: Insert**

- 1. Add the new element to the end of the array.
- 2. Set the new element's priority to  $-\infty$  (for max-heap) or  $\infty$  (for min-heap).
- 3. Use the Increase/Decrease Key operation to set the correct priority.

### Priority Queue: Insert

```
def max_heap_insert(A, obj, n):
    if len(A) == n:
        raise ValueError("Heap overflow")
    key = float("-inf")
    obj.key = key
        A.append(obj)

# map obj to the last index -- dependent on the implementation
        max_heap_increase_key(A, obj, key)
```

### Priority Queue: Extract

- 1. Grab the root element (max or min).
- 2. Replace the root with the last element in the array.
- 3. Remove the last element.
- 4. Call max\_heapify or min\_heapify on the root to maintain the heap property.

## Priority Queue: Extract

```
mox-val = 10
def max heap maximum(A):
    if len(A) < 1:
        raise ValueError("Heap underflow")
    return A[0]
def max_heap_extract_max(A):
    max val = max heap maximum(A)
    A \lceil 0 \rceil = A \lceil -1 \rceil
    A.pop()
    max heapify(A, 0)
    return max val
```

### Priority Queue: Increase Key

- 1. Check if the new key is smaller than the current key.
- 2. Set the element's key to the new key.
- 3. While the element is not the root and its parent's key is less than the element's key:
  - 3.1 Swap the element with its parent.
  - 3.2 Move up to the parent's index.

### Priority Queue: Increase Key

```
def max_heap_increase_key(A, obj, key):
    if key ? obj.key:
        raise ValueError("New key is smaller than current key")
    obj.key = key
    i = A.index(obj) # gets the index of the object
    while i > 0 and A[parent(i)].key A[i].key:
        A[i], A[parent(i)] = A[parent(i)], A[i]
        i = parent(i)
```

#### **Class Exercise**

Implement a minimum priority queue using a min-heap.