

## Lecture 8: Binary Heaps

### Priority Queue Interface

- Keep track of many items, quickly access/remove the most important
  - Example: router with limited bandwidth, must prioritize certain kinds of messages
  - Example: process scheduling in operating system kernels
  - Example: discrete-event simulation (when is next occurring event?)
  - Example: graph algorithms (later in the course)
- Order items by key = priority so **Set interface** (not Sequence interface)
- Optimized for a particular subset of Set operations:
 

<code>build(x)</code>	build priority queue from iterable $x$
<code>insert(x)</code>	add item $x$ to data structure
<code>delete_max()</code>	remove and return stored item with largest key
<code>find_max()</code>	return stored item with largest key
- (Usually optimized for max or min, not both)
- Focus on `insert` and `delete_max` operations: `build` can repeatedly `insert`; `find_max()` can `insert(delete_min())`

### Priority Queue Sort

- Any priority queue data structure translates into a sorting algorithm:
  - `build(A)`, e.g., insert items one by one in input order
  - Repeatedly `delete_min()` (or `delete_max()`) to determine (reverse) sorted order
- All the hard work happens inside the data structure
- Running time is  $T_{\text{build}} + n \cdot T_{\text{delete\_max}} \leq n \cdot T_{\text{insert}} + n \cdot T_{\text{delete\_max}}$
- Many sorting algorithms we've seen can be viewed as priority queue sort:

Priority Queue Data Structure	Operations $O(\cdot)$			Priority Queue Sort	
	<code>build(A)</code>	<code>insert(x)</code>	<code>delete_max()</code>	Time	In-place?
Dynamic Array	$n$	$1_{(a)}$	$n$	$n^2$	Y
Sorted Dynamic Array	$n \log n$	$n$	$1_{(a)}$	$n^2$	Y
Set AVL Tree	$n \log n$	$\log n$	$\log n$	$n \log n$	N
<b>Goal</b>	$n$	$\log n_{(a)}$	$\log n_{(a)}$	$n \log n$	Y

Selection Sort  
Insertion Sort  
AVL Sort  
Heap Sort

## Priority Queue: Set AVL Tree

- Set AVL trees support `insert(x)`, `find_min()`, `find_max()`, `delete_min()`, and `delete_max()` in  $O(\log n)$  time per operation
  - So priority queue sort runs in  $O(n \log n)$  time
    - This is (essentially) AVL sort from Lecture 7
  - Can speed up `find_min()` and `find_max()` to  $O(1)$  time via subtree augmentation
  - But this data structure is complicated and resulting sort is not in-place
  - Is there a simpler data structure for just priority queue, and in-place  $O(n \lg n)$  sort?  
YES, binary heap and heap sort
  - Essentially implement a Set data structure on top of a Sequence data structure (array), using what we learned about binary trees
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## Priority Queue: Array

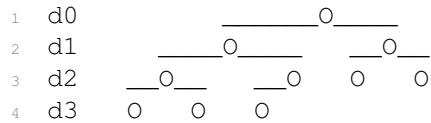
- Store elements in an **unordered** dynamic array
  - `insert(x)`: append  $x$  to end in amortized  $O(1)$  time
  - `delete_max()`: find max in  $O(n)$ , swap max to the end and remove
  - `insert` is quick, but `delete_max` is slow
  - Priority queue sort is selection sort! (plus some copying)
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## Priority Queue: Sorted Array

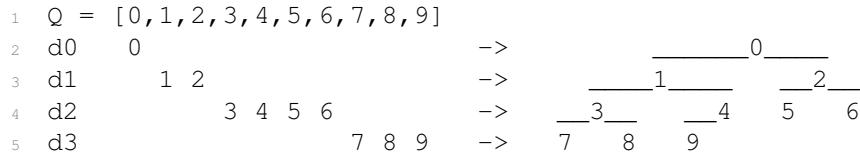
- Store elements in a **sorted** dynamic array
- `insert(x)`: append  $x$  to end, swap down to sorted position in  $O(n)$  time
- `delete_max()`: delete from end in  $O(1)$  amortized
- `delete_max` is quick, but `insert` is slow
- Priority queue sort is insertion sort! (plus some copying)
- Can we find a compromise between these two array priority queue extremes?

## Array as a Complete Binary Tree

- **Idea:** interpret an array as a complete binary tree, with maximum  $2^i$  nodes at depth  $i$  except at the largest depth, where all nodes are **left-aligned**



- Equivalently, complete tree is filled densely in reading order: root to leaves, left to right
- Perspective: **bijection** between arrays and complete binary trees



- Height of complete tree perspective of array of  $n$  item is  $\lceil \lg n \rceil$ , so **balanced** binary tree
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## Implicit Complete Tree

- Complete binary tree structure can be **implicit** instead of storing pointers
- Root is at index 0
- Compute neighbors by index arithmetic:

$$\begin{aligned} \text{left}(i) &= 2i + 1 \\ \text{right}(i) &= 2i + 2 \\ \text{parent}(i) &= \left\lfloor \frac{i - 1}{2} \right\rfloor \end{aligned}$$

## Binary Heaps

- **Idea:** keep larger elements higher in tree, but only locally
  - **Max-Heap Property** at node  $i$ :  $Q[i] \geq Q[j]$  for  $j \in \{\text{left}(i), \text{right}(i)\}$
  - **Max-heap** is an array satisfying max-heap property at all nodes
  - **Claim:** In a max-heap, every node  $i$  satisfies  $Q[i] \geq Q[j]$  for **all nodes**  $j$  in  $\text{subtree}(i)$
  - Proof:
    - Induction on  $d = \text{depth}(j) - \text{depth}(i)$
    - Base case:  $d = 0$  implies  $i = j$  implies  $Q[i] \geq Q[j]$  (in fact, equal)
    - $\text{depth}(\text{parent}(j)) - \text{depth}(i) = d - 1 < d$ , so  $Q[i] \geq Q[\text{parent}(j)]$  by induction
    - $Q[\text{parent}(j)] \geq Q[j]$  by Max-Heap Property at  $\text{parent}(j)$  □
  - In particular, max item is at root of max-heap
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## Heap Insert

- Append new item  $x$  to end of array in  $O(1)$  amortized, making it next leaf  $i$  in reading order
- `max_heapify_up( $i$ )`: swap with parent until Max-Heap Property
  - Check whether  $Q[\text{parent}(i)] \geq Q[i]$  (part of Max-Heap Property at  $\text{parent}(i)$ )
  - If not, swap items  $Q[i]$  and  $Q[\text{parent}(i)]$ , and recursively `max_heapify_up( $\text{parent}(i)$ )`
- Correctness:
  - Max-Heap Property guarantees all nodes  $\geq$  descendants, except  $Q[i]$  might be  $>$  some of its ancestors (unless  $i$  is the root, so we're done)
  - If swap necessary, same guarantee is true with  $Q[\text{parent}(i)]$  instead of  $Q[i]$
- Running time: height of tree, so  $\Theta(\log n)!$

## Heap Delete Max

- Can only easily remove last element from dynamic array, but max key is in root of tree
  - So swap item at root node  $i = 0$  with last item at node  $n - 1$  in heap array
  - $\text{max\_heapify\_down}(i)$ : swap root with larger child until Max-Heap Property
    - Check whether  $Q[i] \geq Q[j]$  for  $j \in \{\text{left}(i), \text{right}(i)\}$  (Max-Heap Property at  $i$ )
    - If not, swap  $Q[i]$  with  $Q[j]$  for child  $j \in \{\text{left}(i), \text{right}(i)\}$  with maximum key, and recursively  $\text{max\_heapify\_down}(j)$
  - Correctness:
    - Max-Heap Property guarantees all nodes  $\geq$  descendants, except  $Q[i]$  might be  $<$  some descendants (unless  $i$  is a leaf, so we're done)
    - If swap is necessary, same guarantee is true with  $Q[j]$  instead of  $Q[i]$
  - Running time: height of tree, so  $\Theta(\log n)!$
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## Heap Sort

- Plugging max-heap into priority queue sort gives us a new sorting algorithm
  - Running time is  $O(n \log n)$  because each `insert` and `delete_max` takes  $O(\log n)$
  - But often include two improvements to this sorting algorithm:
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## In-place Priority Queue Sort

- Max-heap  $Q$  is a prefix of a larger array  $A$ , remember how many items  $|Q|$  belong to heap
- $|Q|$  is initially zero, eventually  $|A|$  (after inserts), then zero again (after deletes)
- `insert()` absorbs next item in array at index  $|Q|$  into heap
- `delete_max()` moves max item to end, then abandons it by decrementing  $|Q|$
- In-place priority queue sort with Array is exactly Selection Sort
- In-place priority queue sort with Sorted Array is exactly Insertion Sort
- In-place priority queue sort with binary Max Heap is **Heap Sort**

## Linear Build Heap

- Inserting  $n$  items into heap calls `max_heapify_up(i)` for  $i$  from 0 to  $n - 1$  (root down):

$$\text{worst-case swaps} \approx \sum_{i=0}^{n-1} \text{depth}(i) = \sum_{i=0}^{n-1} \lg i = \lg(n!) \geq (n/2) \lg(n/2) = \Omega(n \lg n)$$

- **Idea!** Treat full array as a complete binary tree from start, then `max_heapify_down(i)` for  $i$  from  $n - 1$  to 0 (leaves up):

$$\text{worst-case swaps} \approx \sum_{i=0}^{n-1} \text{height}(i) = \sum_{i=0}^{n-1} (\lg n - \lg i) = \lg \frac{n^n}{n!} = \Theta\left(\lg \frac{n^n}{\sqrt{n}(n/e)^n}\right) = O(n)$$

- So can build heap in  $O(n)$  time
  - (Doesn't speed up  $O(n \lg n)$  performance of heap sort)
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## Sequence AVL Tree Priority Queue

- Where else have we seen linear build time for an otherwise logarithmic data structure? Sequence AVL Tree!
  - Store items of priority queue in Sequence AVL Tree in **arbitrary order** (insertion order)
  - Maintain max (and/or min) augmentation:  
`node.max` = pointer to node in subtree of `node` with maximum key
    - This is a subtree property, so constant factor overhead to maintain
  - `find_min()` and `find_max()` in  $O(1)$  time
  - `delete_min()` and `delete_max()` in  $O(\log n)$  time
  - `build(A)` in  $O(n)$  time
  - Same bounds as binary heaps (and more)
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## Set vs. Multiset

- While our Set interface assumes no duplicate keys, we can use these Sets to implement Multisets that allow items with duplicate keys:
  - Each item in the Set is a Sequence (e.g., linked list) storing the Multiset items with the same key, which is the key of the Sequence
- In fact, without this reduction, binary heaps and AVL trees work directly for duplicate-key items (where e.g. `delete_max` deletes *some* item of maximum key), taking care to use  $\leq$  constraints (instead of  $<$  in Set AVL Trees)

TODAY: Heaps

- priority queue interface & sorting algorithm
- set AVL tree  $\rightarrow$  AVL sort
- array  $\rightarrow$  selection sort
- sorted array  $\rightarrow$  insertion sort
- binary heap  $\rightarrow$  heap sort  $\leftarrow$  in place &  $O(n \lg n)$  time

Priority queue interface: recall from Lecture 8

- (- build(X)): initialize to items in iterable X
- (- insert(x)): add item  $x$  (with key  $x.key$ )  $\xrightarrow{\text{e.g. insert}(x) \text{ for } x \in X}$
- (- delete-max()): delete & return item of max key
- (- find-max()): return max-key item  $\xrightarrow{\text{e.g. insert}(\text{delete}-\text{max}())}$   
(or ditto for min instead of max)
- $\rightarrow$  Subset of Set interface

Applications: packet routing, process scheduling,  
discrete-event simulation (when is next event?),  
graph algorithms [Lecture 13]

Priority queue  $\rightarrow$  sorting algorithm

① build(A) e.g.  $\text{insert}(x)$  for  $x$  in A

② delete-max()  $|A|$  times  $\rightarrow$  reverse sorted order

or delete-min()  $|A|$  times  $\rightarrow$  sorted order

$$\Rightarrow T_{\text{build}} + n \cdot T_{\text{delete-max}} \leq n \cdot T_{\text{insert}} + n \cdot T_{\text{delete-max}} \text{ time}$$

## Set AVL tree: [Lecture 7]

- insert() & find/delete-min/max() in  $O(\lg n)$ /op.
- priority queue sort  $\approx$  AVL sort
  - $O(n \lg n)$  time
- can speed find-min/max() to  $O(1)$  time via Subtree property augmentation
- Complicated, and not in-place sort

Array-based priority queues: can we use a Sequence data structure to implement Set?

### Unsorted array: (dynamic)

- insert(x): append  $x$
- delete-max(): scan, swap to end, pop

$\rightarrow$  priority queue sort  $\approx$  selection sort

$O(1)$  amortized  
 $O(n)$   $\ddot{\wedge}$



### Sorted array: (dynamic)

- insert(x): append  $x$ ,

swap down to sorted position

$\ddots$   
↓

$O(n)$   $\ddot{\wedge}$   
 $O(1)$  amortized

- delete-max(): pop

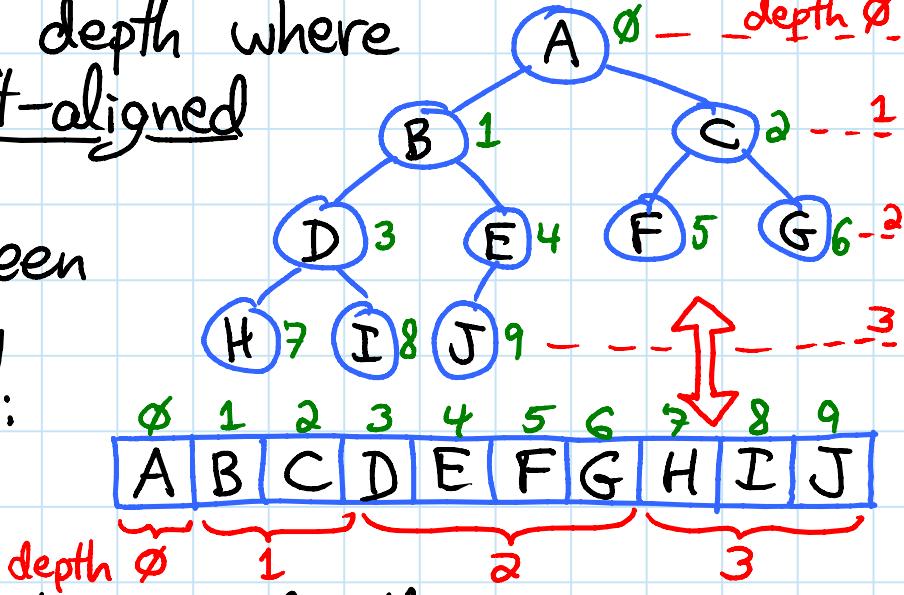
$\rightarrow$  priority queue sort  $\approx$  insertion sort

Can we find a compromise between these two array priority queue extremes?  
Using binary trees perhaps?

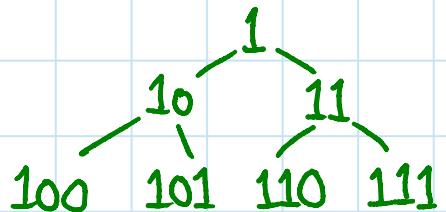
## Complete binary tree:

# nodes at depth  $i = 2^i$  (maximum)  
 except maximum depth where  
 nodes are left-aligned

- bijection between complete binary trees & arrays:



- complete binary tree implicitly represented by array - don't need to store pointers!
  - root at index  $\emptyset$
  - $\text{left}(i) = 2i + 1$
  - $\text{right}(i) = 2i + 2$
  - $\Rightarrow \text{parent}(i) = \lfloor \frac{i-1}{2} \rfloor$  i.e.  $(i-1) // 2$
  - one way to see:  
 look at binary rep.  
 of  $i+1$



$\rightarrow$  array of items  $\rightarrow$  associate node with index

Binary heap  $Q$ : every node  $i$  satisfies  
Max-Heap Property at node  $i$ :

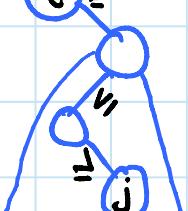
$$Q[i] \geq Q[j] \text{ for } j \in \{\text{left}(i), \text{right}(i)\}$$

$i$        $j$

$\hookrightarrow$  item stored in node  $i/j$

Lemma:  $\Rightarrow Q[i] \geq Q[j]$  for node  $j$  in  $\text{subtree}(i)$

Proof: by induction on  $d = \text{depth}(j) - \text{depth}(i)$

- base:  $d=0 \Rightarrow i=j \Rightarrow Q[i] = Q[j]$
  - Step:  $Q[i] \geq Q[\text{parent}(j)] \geq Q[j]$   $\square$   
 $\hookrightarrow$  by Max-Heap Property
- $i \geq$
- 
- $\hookrightarrow$  by induction

e.g. max at root  $\Rightarrow \text{find-max}() = Q[\emptyset]$

insert( $x$ ):

$\hookrightarrow$  Python append

- $Q.\text{insert\_last}(x)$
  - $i = |Q| - 1$  (index of  $x$ )
  - max-heapify-up( $i$ ):
    - if  $i=0$ : return
    - if  $Q[\text{parent}(i)] < Q[i]$ : violate Max-Heap P.
    - Swap  $Q[i] \leftrightarrow Q[\text{parent}(i)]$
    - recurse:  $\text{max-heapify-up}(\text{parent}(i))$
- $\Rightarrow O(\lg n)$  amortized

$O(1)$  amortized

$O(h) = O(\lg n)$

heap before  
 $\text{insert}(10)$



## delete\_max():

- swap  $Q[\emptyset] \leftrightarrow Q[|Q|-1]$

- $Q.\text{delete\_last}()$

- $i = \emptyset$

- max\_heapify\_down(i):

- if  $i$  is leaf: return if exists

- let  $j \in \{\text{left}(i), \text{right}(i)\}$  maximize  $Q[j]$

- if  $Q[i] < Q[j]$ : violate Max Heap Prop.

- swap  $Q[i] \leftrightarrow Q[j]$

- recurse:  $\text{max\_heapify\_down}(j)$

$\Rightarrow O(\lg n)$  amortized

heap before  
 $\emptyset$  swapped to root



Heap sort  $\approx$  priority queue sort with binary heap

## In-place priority queue sort:

- $Q$  is a prefix of input array  $A$

- $|Q|$  starts  $\emptyset$ , eventually  $|A|$ , then  $\emptyset$  again

- $\text{insert}()$  increments  $|Q|$  instead of  $\text{insert\_last}()$

- $\text{delete\_min}()$  decrements  $|Q|$  instead of  $\text{delete\_last}()$

- with array  $\rightarrow$  selection sort

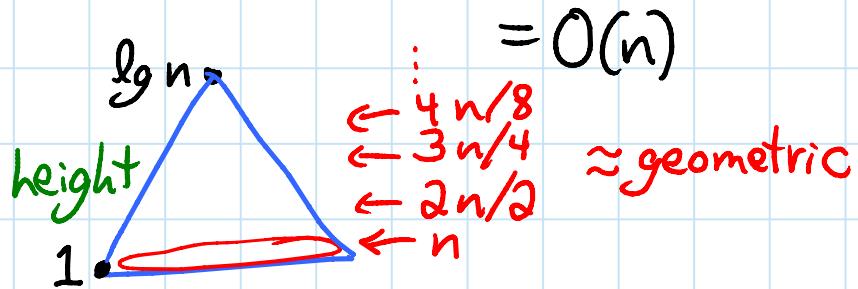
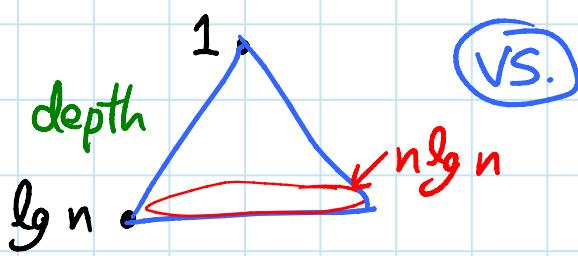
- with sorted array  $\rightarrow$  insertion sort

- with binary heap  $\rightarrow$  heap sort

$\Rightarrow$  In-place  $O(n \lg n)$  sorting algorithm!

## Build heap:

- $n$  inserts calls `max-heapify-up(i)` for  $i=0, \dots, n-1$   
 $\Rightarrow \sum_{i=0}^{n-1} \text{depth}(i) = \sum_{i=0}^{n-1} \lg i = \lg n! \geq \frac{n}{2} \lg \frac{n}{2} = \Omega(n \lg n)$
- instead: `max-heapify-down(i)` for  $i=n-1, \dots, 0$   
 $\Rightarrow \sum_{i=0}^{n-1} \text{height}(i) = \sum_{i=0}^{n-1} (\lg n - \lg i) = \lg \frac{n^n}{n!} \approx \lg \frac{n^n}{\sqrt{n} (n/e)^n}$



## Sequence AVL tree priority queue: an alternative

- Sequence AVL tree also builds in  $O(n)$  time
- use arbitrary traversal order (insertion order)
- augment to store pointer to max-key node within every node's subtree
  - Subtree property:  
 $\text{node.max} = \text{node}$  or  $\text{node.left.max}$   
 or  $\text{node.right.max}$

$\Rightarrow$  can maintain

- `delete_max()`: `subtree_delete(root.max)`
- $\Rightarrow$  `insert(x)` & `delete_min/max()` in  $O(\lg n)$
- & `find_min/max()` in  $O(1)$  & build in  $O(n)$
- (Same as binary heaps, plus more)
- (but more complicated & not in-place sort)

## Set vs. multiset:

- so far assumed keys are unique in Set
- priority queue & sorting naturally have duplicate keys (priorities)
- can build Multiset from Set:
  - Set item = Sequence → e.g. linked list of Multiset items of same key  
key of Sequence =
- in fact binary heap & set AVL tree directly work with duplicate keys (using  $\leq$  constraint instead of  $<$  in AVL)
  - `delete-max()` = delete some max-key item