

Binary Heaps in Dynamic Arrays

CSCI 2100 Teaching Team

Department of Computer Science and Engineering
Chinese University of Hong Kong

Outline

- 1 An array-based implementation of the binary heap.
- 2 A heap building algorithm with $O(n)$ time complexity.

Review: Priority Queue

A **priority queue** stores a set S of n integers and supports the following operations:

- **Insert(e)**: Adds a new integer to S .
- **Delete-min**: Removes and returns the smallest integer in S .

Review: Binary Heap

Let S be a set of n integers. A **binary heap** on S is a binary tree T satisfying:

- 1 T is complete.
- 2 Every node u in T corresponds to a distinct integer in S — the integer is called the **key** of u (and is stored at u).
- 3 If u is an internal node, the key of u is smaller than those of its child nodes.

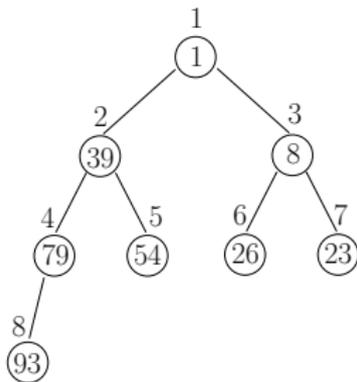
Storing a Complete Binary Tree Using an Array

Let T be any complete binary tree with n nodes. We can **linearize** the nodes in the following manner:

- Put the nodes at a higher level before those at a lower level.
- Within the same level, order the nodes from left to right.

Store the linearized node sequence in an array A of length n .

Example



Stored as

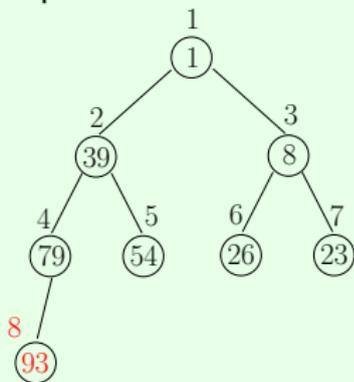
Index: 1 2 3 4 5 6 7 8

1	39	8	79	54	26	23	93
---	----	---	----	----	----	----	----

A

Property 1: The rightmost leaf node at the bottom level is stored at $A[n]$.

Example:



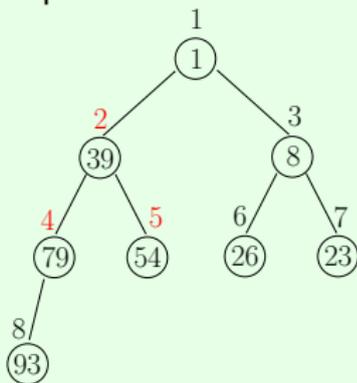
Index: 1 2 3 4 5 6 7 8

1	39	8	79	54	26	23	93
---	----	---	----	----	----	----	----

A

Property 2: Suppose that node u of T is stored at $A[i]$. Then, the left child of u is stored at $A[2i]$, and the right child at $A[2i+1]$.

Example:



Index: 1 2 3 4 5 6 7 8

1	39	8	79	54	26	23	93
---	----	---	----	----	----	----	----

A

Property 2 implies:

Property 3: Suppose that node u of T is stored at $A[i]$. Then, the parent of u is stored at $A[\lfloor i/2 \rfloor]$.

Now we are ready to implement the insertion and delete-min algorithms on the array representation of a binary heap.

Insertion Example

Insert 15 and swap-up.

Index: 1 2 3 4 5 6 7 8 9

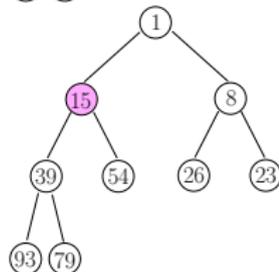
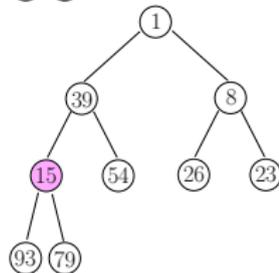
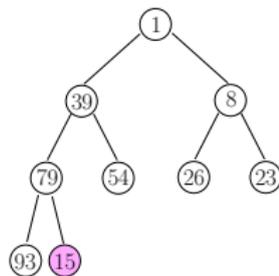
1	39	8	79	54	26	23	93	15
---	----	---	----	----	----	----	----	----

1 2 3 4 5 6 7 8 9

1	39	8	15	54	26	23	93	79
---	----	---	----	----	----	----	----	----

1 2 3 4 5 6 7 8 9

1	15	8	39	54	26	23	93	79
---	----	---	----	----	----	----	----	----



Delete-min Example

Replace 1 with 79 and **swap-down**.

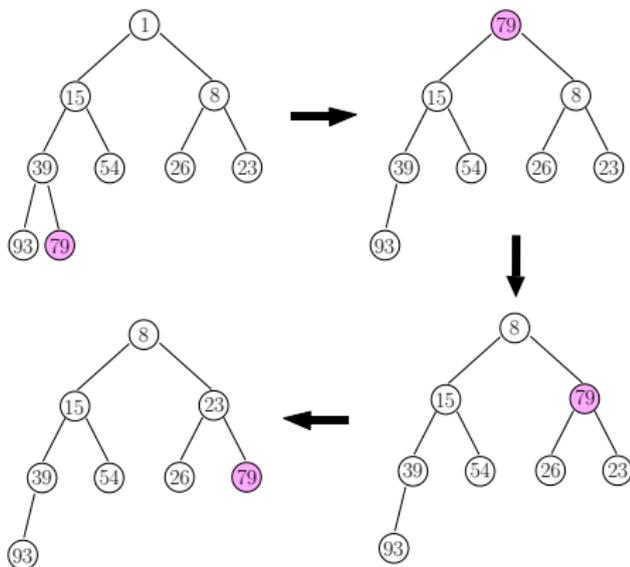
Index: 1 2 3 4 5 6 7 8 9

1	15	8	39	54	26	23	93	79
---	----	---	----	----	----	----	----	----

1	2	3	4	5	6	7	8
79	15	8	39	54	26	23	93

1	2	3	4	5	6	7	8
8	15	79	39	54	26	23	93

1	2	3	4	5	6	7	8
8	15	23	39	54	26	79	93



Performance Guarantees

Combining our analysis on (i) binary heaps and (ii) dynamic arrays, we obtain the following guarantees on a binary heap implemented with a dynamic array:

- Space consumption $O(n)$.
- Insertion: $O(\log n)$ time **amortized**.
- Delete-min: $O(\log n)$ time **amortized**.

Next, we will see a heap building algorithm that runs in $O(n)$ time.

Fixing a Messed-Up Root

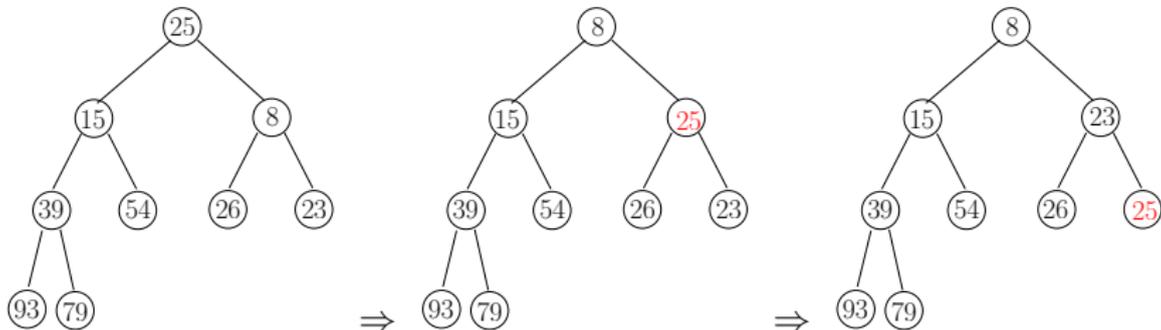
First, consider the following **root-fixing** problem. Suppose that we are given a complete binary tree T with root r such that

- the left subtree of r is a binary heap;
- the right subtree of r is a binary heap.

However, the key of r **may not** be smaller than the keys of its children. We need to fix the issue and makes T a binary heap.

This can be done in $O(\log n)$ time using the **swap-down** operation from the **delete-min** algorithm.

Example



Building a Heap

Given an array A that stores a set S of n integers, we can turn A into a binary heap on S using the following simple algorithm (which views A as a complete binary tree T).

- For each $i = \lfloor n/2 \rfloor$ **downto** 1
 - Apply swap-down to the subtree of T rooted at $A[i]$ to fix its root.

Think: Are the conditions of the root-fixing problem always satisfied?

Example

i

54	26	15	39	8	1	23	93
----	----	----	----	---	---	----	----

i

54	26	1	39	8	15	23	93
----	----	---	----	---	----	----	----

i

54	8	1	39	26	15	23	93
----	---	---	----	----	----	----	----

i

1	8	15	39	26	54	23	93
---	---	----	----	----	----	----	----

Running Time

Now let us analyze the time of the building algorithm. Suppose that T has height h . Without loss of generality, assume that all the levels of T are full – namely, $n = 2^h - 1$ (why no generality is lost?).

Observe:

- A node at Level $h - 1$ incurs $O(1)$ time in swap-down; 2^{h-1} such nodes.
- A node at Level $h - 2$ incurs $O(2)$ time in swap-down; 2^{h-2} such nodes.
- A node at Level $h - 3$ incurs $O(3)$ time in swap-down; 2^{h-3} such nodes.
- ...
- A node at Level $h - h$ incurs $O(h)$ time in swap-down; 2^0 such nodes.

Running Time

Hence, the total time is bounded by

$$\sum_{i=1}^h O(i \cdot 2^{h-i}) = O\left(\sum_{i=1}^h i \cdot 2^{h-i}\right)$$

We will prove that the right hand side is $O(n)$ in the next slide.

Running Time

Suppose that

$$x = 2^{h-1} + 2 \cdot 2^{h-2} + 3 \cdot 2^{h-3} + \dots + h \cdot 2^0 \quad (1)$$

$$\Rightarrow 2x = 2^h + 2 \cdot 2^{h-1} + 3 \cdot 2^{h-2} + \dots + h \cdot 2^1 \quad (2)$$

Subtracting (1) from (2) gives

$$\begin{aligned} x &= 2^h + 2^{h-1} + 2^{h-2} + \dots + 2^1 - h \\ &\leq 2^{h+1} \\ &= 2(n+1) = O(n). \end{aligned}$$