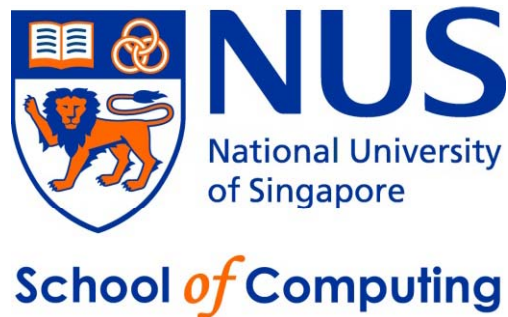


CS2020 – Data Structures and Algorithms Accelerated

Lecture 06 – Heaps of Fun

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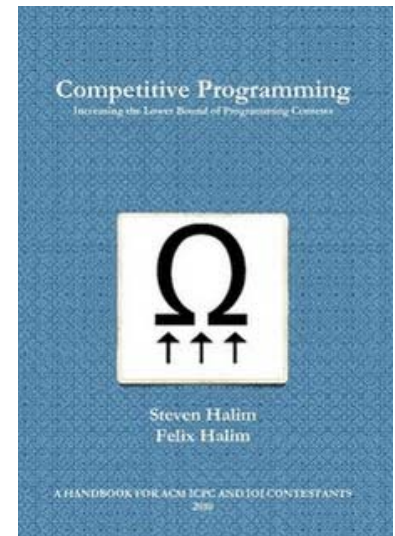
About Me (1)

- The 2nd lecturer of CS2020:
 - Dr Steven Halim
 - Call me as: Steven
- Website:
 - <http://www.comp.nus.edu.sg/~stevenha>
- How to reach me:
 - Email: stevenhalim@gmail.com
(+ Facebook 😊)
 - My office: COM2-03-37
 - Office Number: 6516-7361



About Me (2)

- I also teach:
 - CS3233 – Competitive Programming
- I co-author^ “Competitive Programming” book
 - <http://www.lulu.com/product/paperback/competitive-programming/12110025>
 - <https://sites.google.com/site/stevenhalim>
 - ~15 copies are available at 20 SGD/copy
- I am the coach for:
 - NUS ACM ICPC teams
 - International Collegiate Programming Contest
 - Singapore IOI team*
 - International Olympiad in Informatics



Special Note

- Steven will be lecturing mostly during the 2nd half of CS2020
- This is a “one off” lecture to cover a data structure that will be used again during the 2nd half of the class

Outline

- What are you going to learn in this lecture?
 - Motivation: Abstract Data Type: **PriorityQueue**
 - **Heap** data structure
 - **Heap sort**

Abstract Data Type: PriorityQueue

- Important Basic Operations:
 - Enqueue(x)
 - Put a new item x in the priority queue PQ (in some order)
 - $y \leftarrow \text{Dequeue}()$
 - Return an item y that has the **highest priority** (key) in the PQ
 - If there are more than one item with highest priority, return the one that is inserted first (FIFO)

Few Points To Remember


- Data Structure is...
 - A particular way of **storing** and **organizing data** in a computer so that it can be used efficiently
- Most data structure have **propert(ies)**
 - Each operation on that data structure has to **maintain** that **propert(ies)**

PriorityQueue Implementation (1)


- **Array-Based Implementation (Strategy 1)**
 - Property: the content of array is always in correct order
 - Enqueue(x)
 - Find the **correct insertion place**, $O(n)$
 - $y \leftarrow \text{Dequeue}()$
 - Return the **front-most item** which has the highest priority, $O(1)$

Index	0 (front)	1 (back)
Key	Aircraft X*	Aircraft Y*

Aircraft Z**

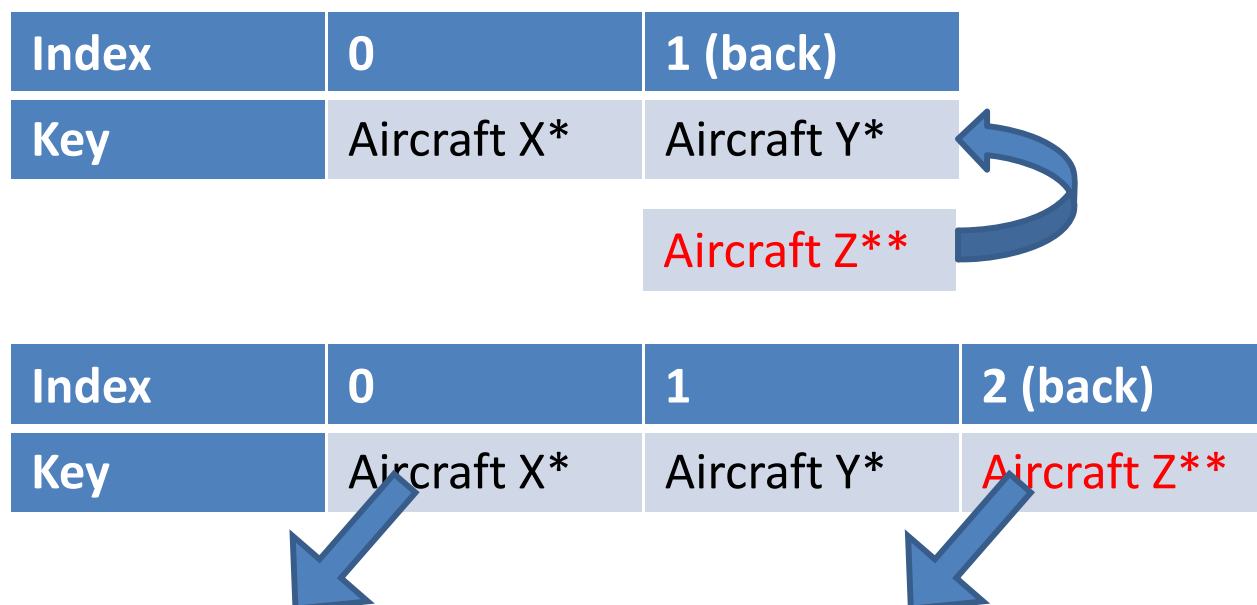


Index	0 (front)	1	2 (back)
Key	Aircraft Z**	Aircraft X*	Aircraft Y*



PriorityQueue Implementation (2)

- **Array-Based Implementation (Strategy 2)**
 - Property: dequeue() operation returns the correct item
 - Enqueue(x)
 - Put the new item at the **back of the queue**, $O(1)$
 - $y \leftarrow \text{Dequeue}()$
 - Scan the whole queue, return **first item with highest priority**, $O(n)$



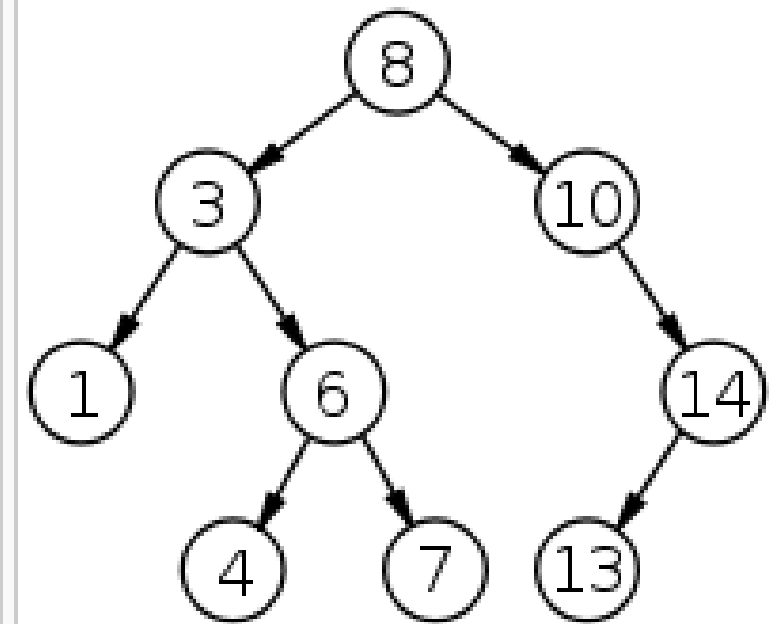
PriorityQueue Implementation (3)

Strategy	Enqueue	Dequeue
Array-Based PQ (1)	$O(N)$	$O(1)$
Array-Based PQ (2)	$O(1)$	$O(N)$
We can do better!	$O(?)$	$O(?)$

INTRODUCING HEAP DATA STRUCTURE

Quick Review

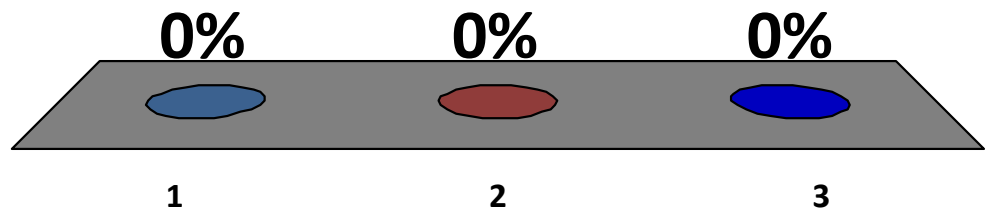
- Heap is similar to what you already know:
Binary Search Tree (BST, from previous two lectures)
 - Vertex/Node/Item
 - Edge
 - Root
 - Internal Nodes
 - Leaves
 - Binary Tree
 - Left/Right Sub-Tree
 - The **BST Property**...



A binary search tree of size 9 and depth 3, with root 8 and leaves 1, 4, 7 and 13

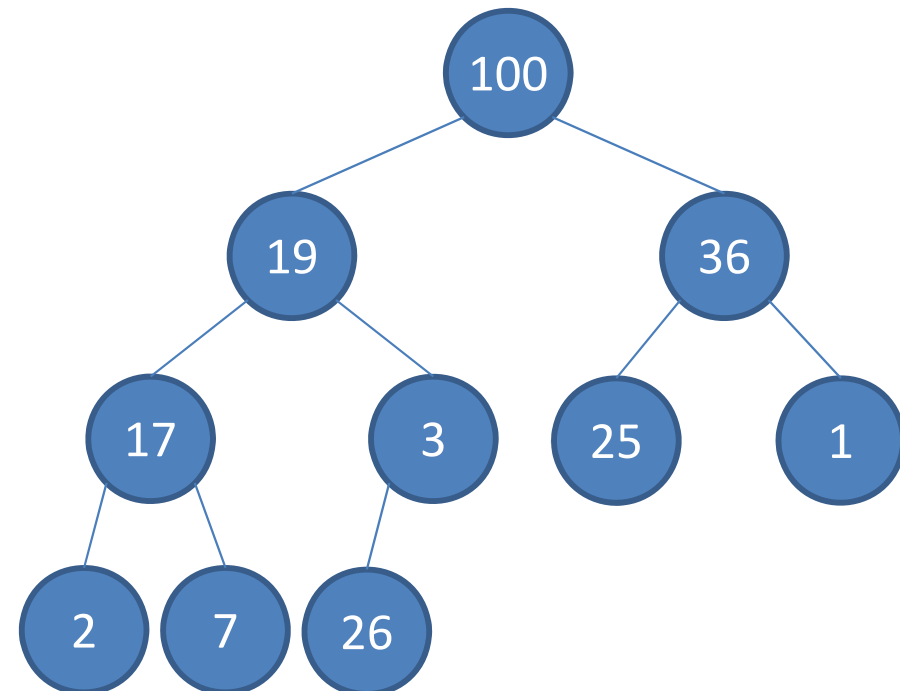
The BST Property Is...

1. $\text{key}[x] < \text{key}[\text{left}[x]] < \text{key}[\text{right}[x]]$
2. $\text{key}[\text{left}[x]] < \text{key}[x] < \text{key}[\text{right}[x]]$
3. $\text{key}[\text{right}[x]] < \text{key}[\text{left}[x]] < \text{key}[x]$



Complete Binary Tree

- Introducing few more concepts:
 - **Complete** Binary Tree
 - Binary tree in which every level, *except possibly the last*, is completely filled, and all nodes are as far left as possible
 - If you have a complete binary tree of N items, what will be **the height of it?**



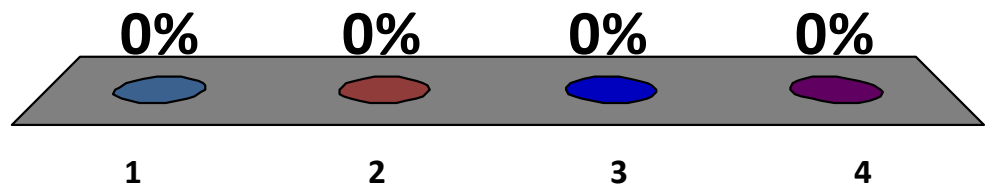
The Height of a Complete Binary Tree of N Items is...

55

1. $O(\sqrt{N})$
2. $O(N)$
3. $O(\log N)$
4. $O(1)$

Now, memorize this answer, we will need that for all the time complexity analysis of heap operations

0



Storing Complete Binary Tree

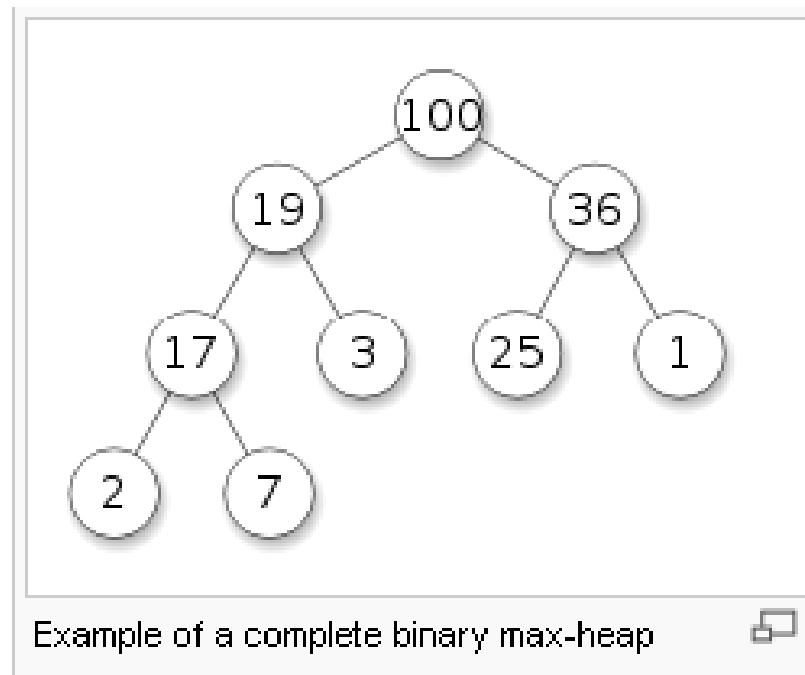
- As an 1-based compact array: $A[1..size(A)]$ $size(A)$

0	1	2	3	4	5	6	7	8	9	10	11
NIL	100	19	36	17	3	25	1	2	7	-	-

- Navigation operations:

- $Parent(i) = \text{floor}(i/2)$
 - Except for $i = 1$
- $Left(i) = 2*i$
- $Right(i) = 2*i + 1$
 - No left/right child when:
 - $Left(i) > \text{heapsize}$
 - $Right(i) > \text{heapsize}$

$\text{heapsize} \leq \text{size}(A)$

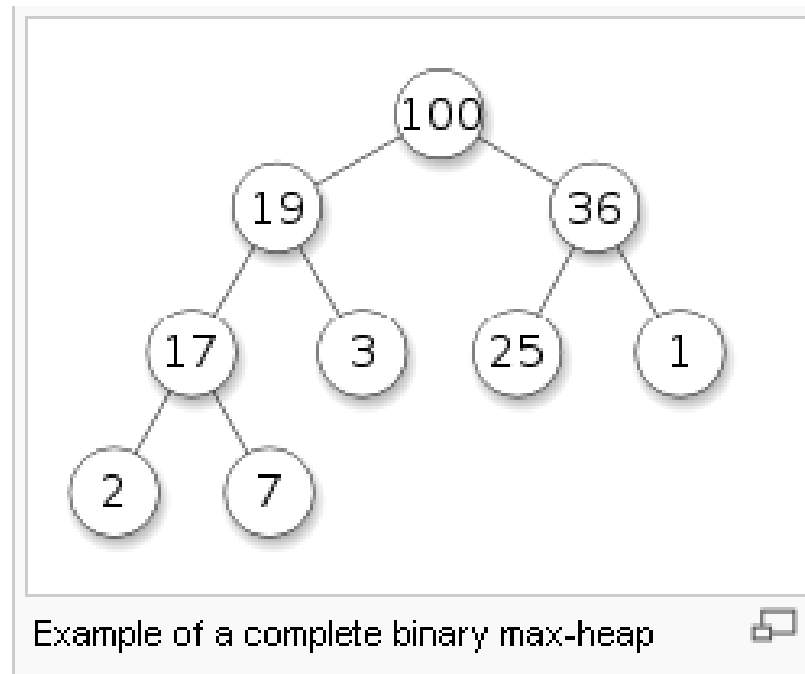


The Heap Property

– The **Heap property**

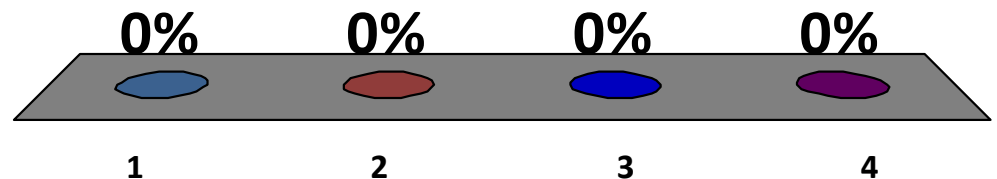
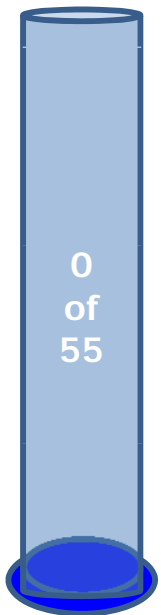
- $A[\text{parent}(i)] \geq A[x]$ (**max heap**)
- $A[\text{parent}(i)] \leq A[x]$ (**min heap**)

– Without loss of generality,
I will use “**max heap**”
for all examples
in this lecture



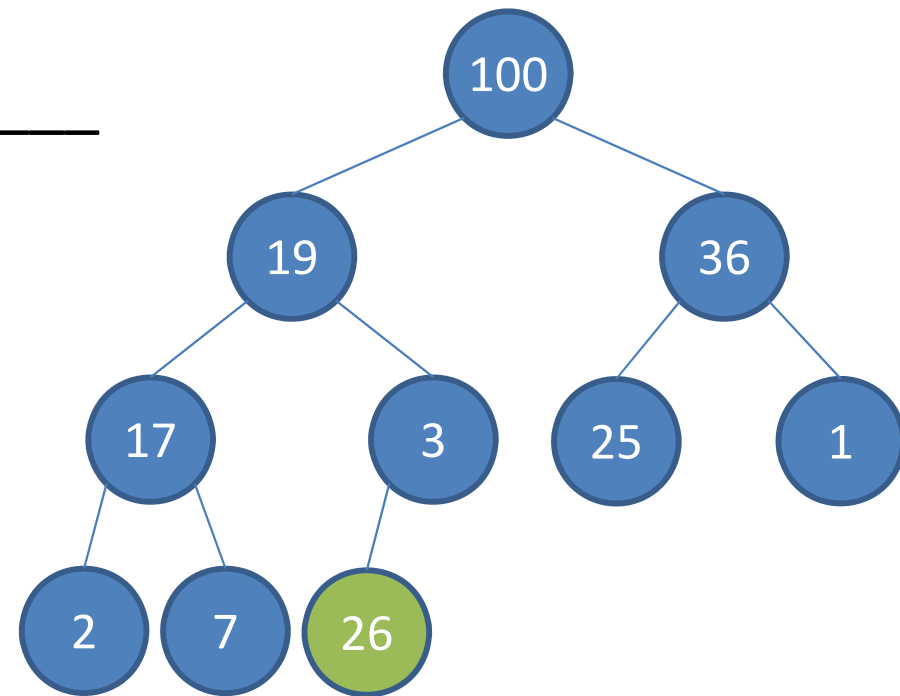
The largest element in a **max-heap** is stored at...

1. One of the leaves
2. One of the internal nodes
3. Can be anywhere in the heap
4. Must be at the root



Insertion to Existing Heap

- The most appropriate insertion point to an existing heap is the **bottom-most, right-most new leaf**
- Why?
 - _____
- But the Heap property can still be violated?
 - No problem, we use `shiftUp(i)` to fix the heap property



0	1	2	3	4	5	6	7	8	9	10	11
0	100	19	36	17	3	25	1	2	7		

Heap_Insert – Pseudo Code

```
Heap_Insert(key)
```

```
    heapsize = heapsize + 1; // extend  $O(1)$ 
```

```
    A[heapsize] = key // insert at the back  $O(1)$ 
```

```
    shiftUp(heapsize) // fix the heap property  
                      // in  $O(?)$ 
```

```
// Preliminary analysis:
```

```
// Heap_Insert(A, key) depends on shiftUp(A, i)
```

shiftUp – Pseudo Code

- Name is not unique:
shiftUp/bubbleUp/HeapIncreaseKey/etc

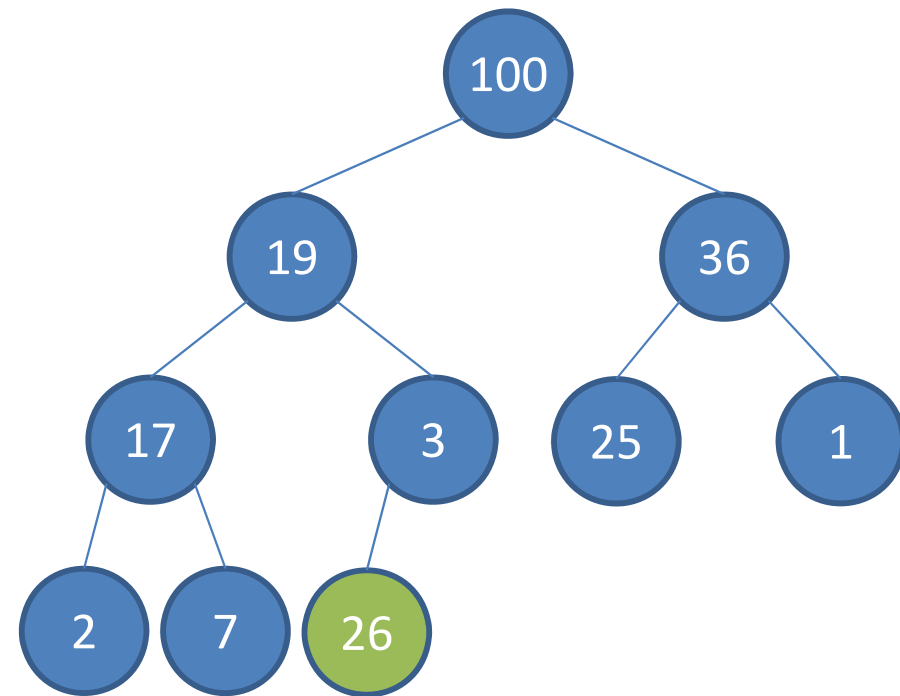
```
shiftUp(A, i)
  while i > 1 and A[parent(i)] < A[i]
    swap(A[i], A[parent(i)])
    i = parent(i)
```

“not root” (pointing to `i > 1`)

“violates max heap property” (pointing to `A[parent(i)] < A[i]`)

Animation (1)

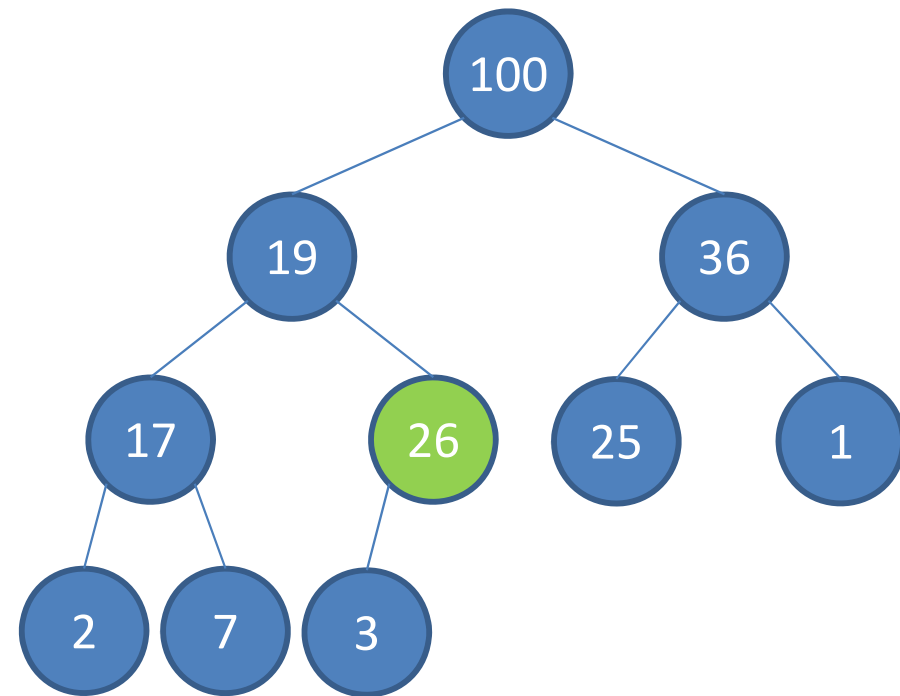
```
shiftUp(A, i)
  while i > 1 and A[parent(i)] < A[i]
    swap(A[i], A[parent(i)])
    i = parent(i)
```



0	1	2	3	4	5	6	7	8	9	10	11
0	100	19	36	17	3	25	1	2	7	26	

Animation (2)

```
shiftUp(A, i)
  while i > 1 and A[parent(i)] < A[i]
    swap(A[i], A[parent(i)])
    i = parent(i)
```

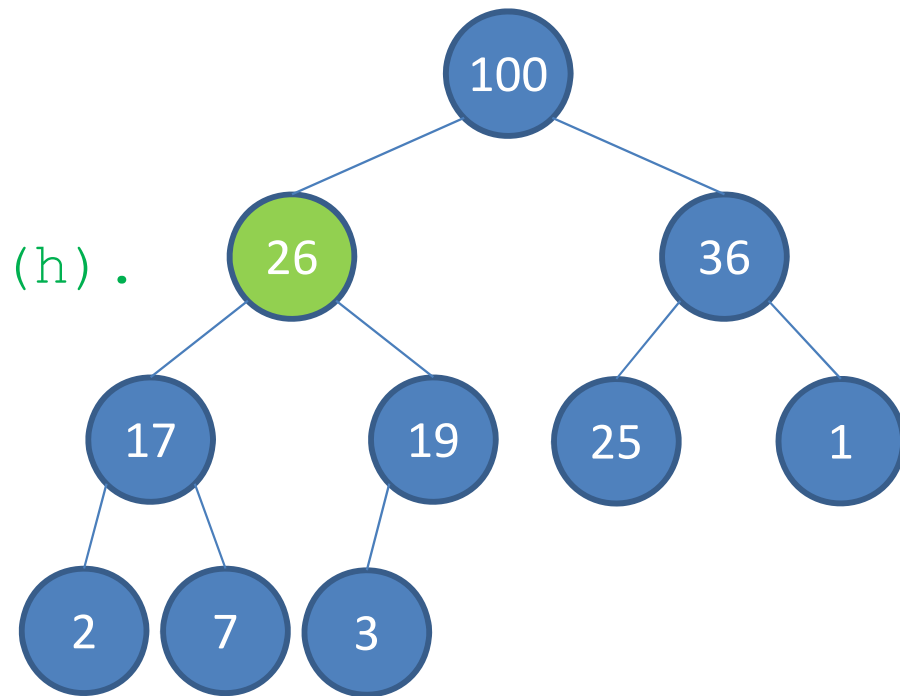


0	1	2	3	4	5	6	7	8	9	10	11
0	100	19	36	17	26	25	1	2	7	3	

Animation (3)

```
shiftUp(A, i)
  while i > 1 and A[parent(i)] < A[i] // see below
    swap(A[i], A[parent(i)]) // O(1)
    i = parent(i) // O(1)
```

```
// Analysis: The worst case is
// from deepest leaf to root O(h).
// In a complete binary tree,
// this h is just log N.
// Thus, shiftUp AND
// Heap_Insert runs in
// O(log N)
```



0	1	2	3	4	5	6	7	8	9	10	11
0	100	26	36	17	19	25	1	2	7	3	

Deleting Max Element

- The max element of a max heap is at **the root**
- But simply taking the root out from a max heap will disconnect the complete binary tree ☹
- We don't want that...
- So, which node is the best candidate to **replace** the root yet still maintain complete binary tree property?
- Again the _____ **existing leaf**
 - Which is again the last element in the compact array
- But the heap property can still be violated?
 - No problem, this time we call `shiftDown(1)`

Heap_ExtractMax - Pseudocode

```
Heap_ExtractMax()
```

```
    maxV  $\leftarrow$  A[1] // O(1)
```

```
    A[1]  $\leftarrow$  A[heapsize] // O(1)
```

```
    heapsize = heapsize - 1 // O(1)
```

```
    shiftDown(1) // O(?)
```

```
    return maxV
```

```
// Preliminary analysis:
```

```
// Heap_ExtractMax() depends on shiftDown()
```

ShiftDown – Pseudo Code

```
shiftDown(i)
  while i <= heapsize
    maxV ← A[i]; max_id = i;
    if Left(i) <= heapsize and maxV < A[Left(i)]
      maxV ← A[Left(i)]; max_id ← Left(i)
    if Right(i) <= heapsize and maxV < A[Right(i)]
      maxV ← A[Right(i)]; max_id ← Right(i)

    if (max_id != i)
      swap(A[i], A[max_id])
      i = max_id;
  else
    break;
```

Again, name is not unique:
shiftDown/bubbleDown/Heapify/etc

Animation (1)

```
Heap_ExtractMax()
```

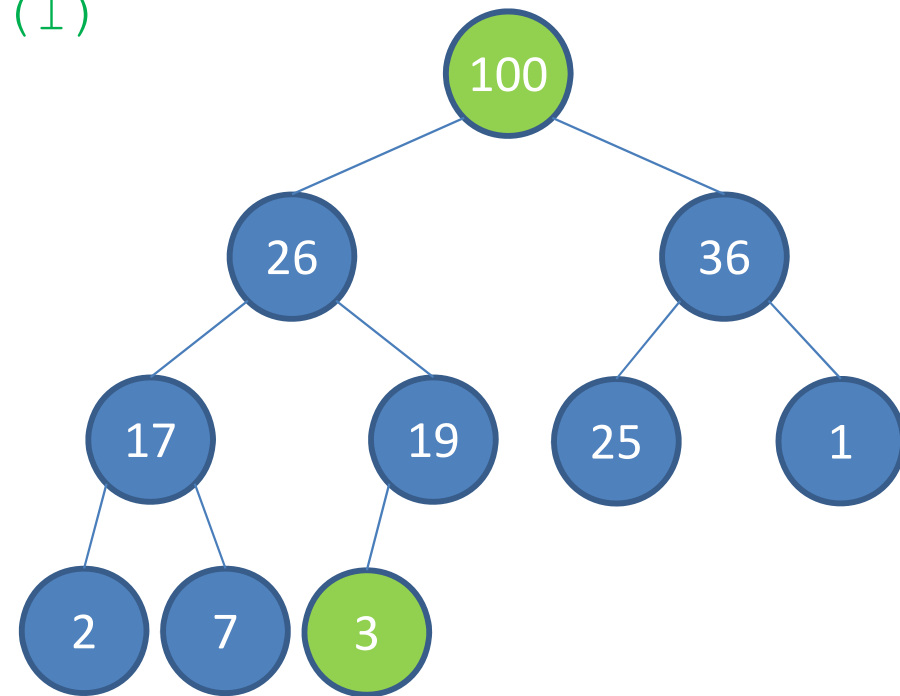
```
maxV ← A[1] // O(1)
```

```
A[1] ← A[heapsize] // O(1)
```

```
heapsize = heapsize - 1 // O(1)
```

```
shiftDown(1) // O(?)
```

```
return maxV
```

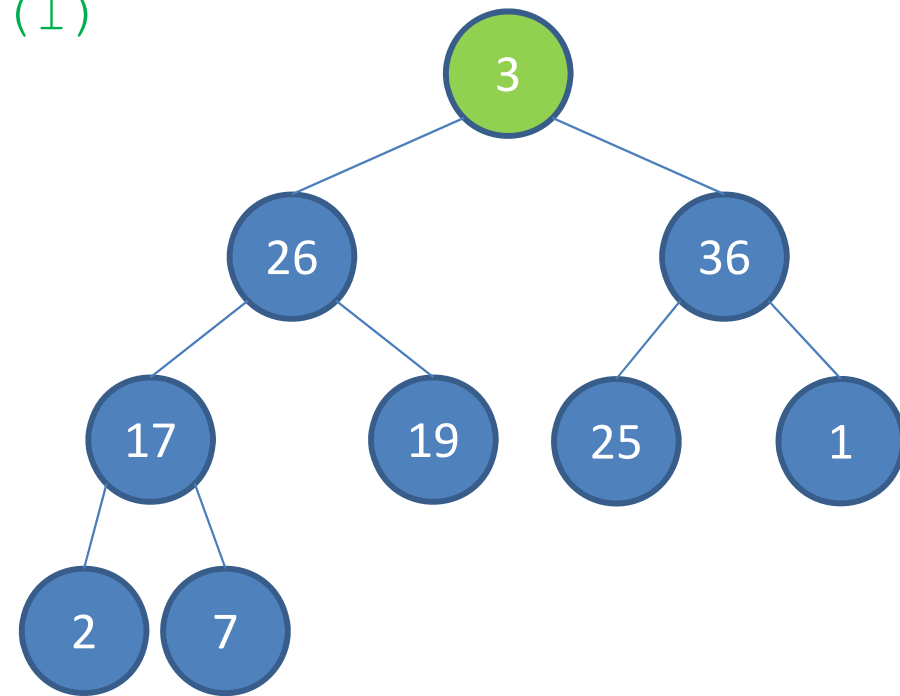


0	1	2	3	4	5	6	7	8	9	[10]	11
0	100	26	36	17	19	25	1	2	7	3	

Animation (2)

```
Heap_ExtractMax()  
  maxV ← A[1] // O(1)  
  A[1] ← A[heapsize] // O(1)  
  heapsize = heapsize - 1 // O(1)  
  shiftDown(1) // O(?)  
  return maxV
```

100 is stored at maxV
and returned later after
shiftDown(1) is done

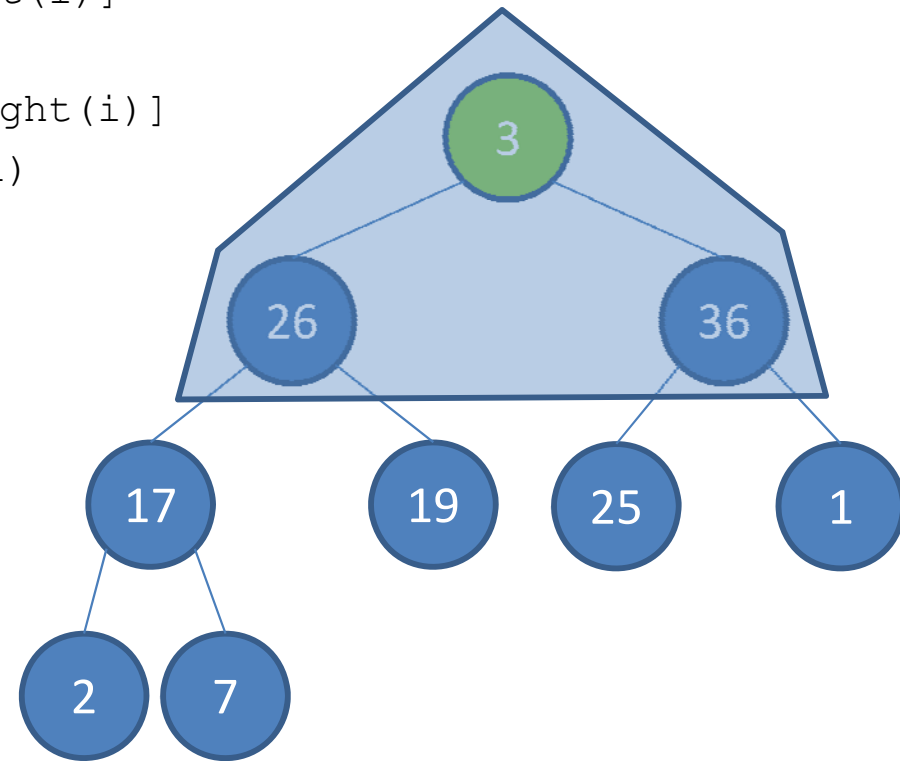


0	1	2	3	4	5	6	7	8	[9]	10	11
0	3	26	36	17	19	25	1	2	7		

Animation (3)

```
shiftDown(i)
  while i <= heapsize
    maxV ← A[i]; max_id = i;
    if Left(i) <= heapsize and maxV < A[Left(i)]
      maxV ← A[Left(i)]; max_id ← Left(i)
    if Right(i) <= heapsize and maxV < A[Right(i)]
      maxV ← A[Right(i)]; max_id ← Right(i)

    if (max_id != i)
      swap(A[i], A[max_id])
      i = max_id;
    else
      break;
```

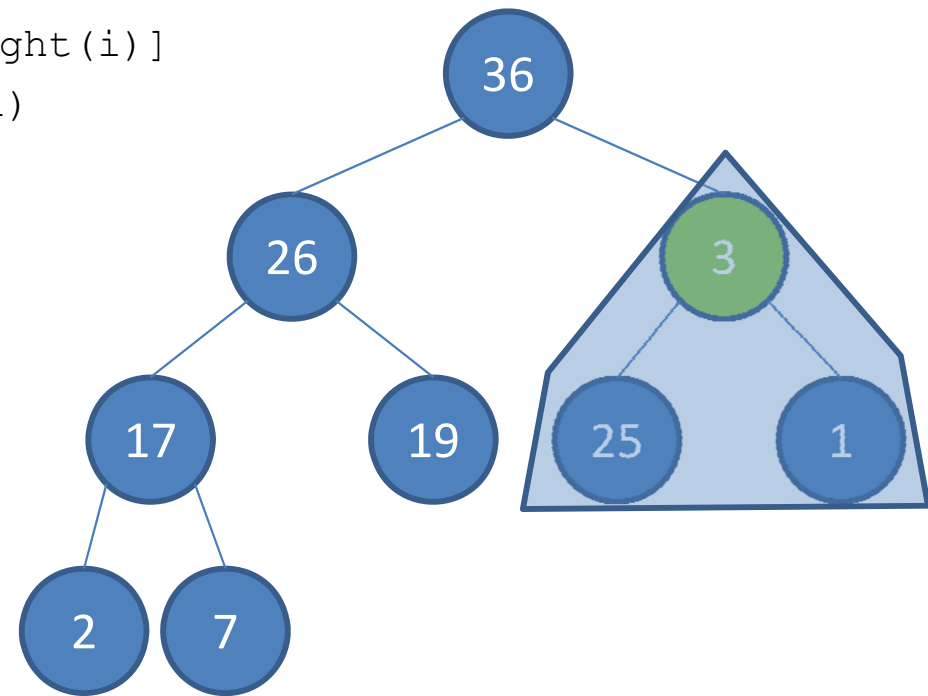


0	1	2	3	4	5	6	7	8	[9]	10	11
0	3	26	36	17	19	25	1	2	7		

Animation (4)

```
shiftDown(i)
while i <= heapsize
    maxV ← A[i]; max_id = i;
    if Left(i) <= heapsize and maxV < A[Left(i)]
        maxV ← A[Left(i)]; max_id ← Left(i)
    if Right(i) <= heapsize and maxV < A[Right(i)]
        maxV ← A[Right(i)]; max_id ← Right(i)

    if (max_id != i)
        swap(A[i], A[max_id])
        i = max_id;
    else
        break;
```



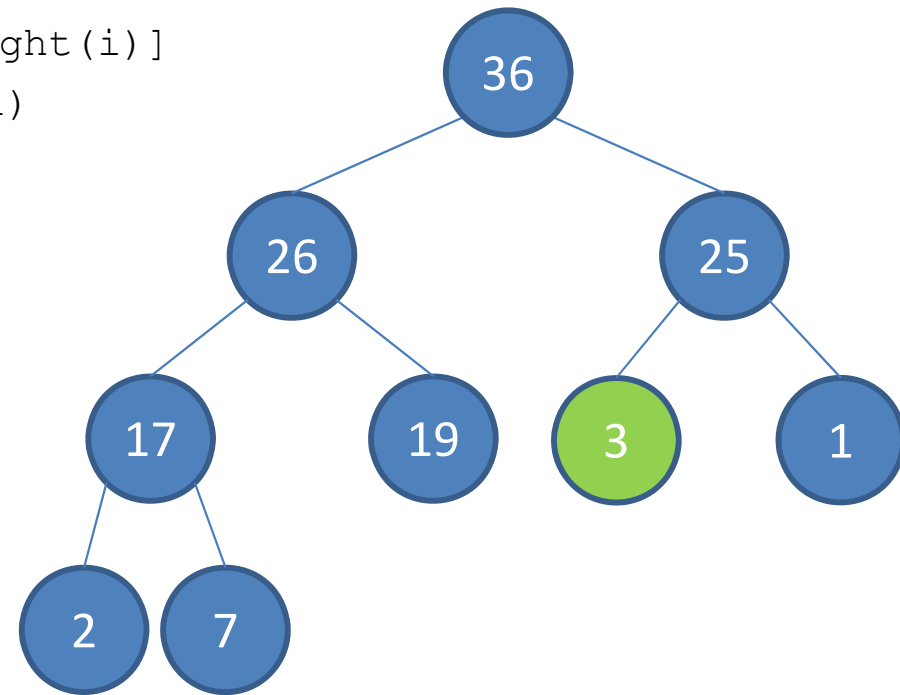
0	1	2	3	4	5	6	7	8	[9]	10	11
0	36	26	3	17	19	25	1	2	7		

Animation (5)

```
shiftDown(i)
  while i <= heapsize // at most root to leaf!  $O(h) = O(\log N)$ 
    maxV  $\leftarrow$  A[i]; max_id = i;
    if Left(i) <= heapsize and maxV < A[Left(i)]
      maxV  $\leftarrow$  A[Left(i)]; max_id  $\leftarrow$  Left(i)
    if Right(i) <= heapsize and maxV < A[Right(i)]
      maxV  $\leftarrow$  A[Right(i)]; max_id  $\leftarrow$  Right(i)

    if (max_id != i)
      swap(A[i], A[max_id])
      i = max_id;
    else
      break;

// In overall, shiftDown AND
// Heap_ExtractMax runs in
//  $O(h) = O(\log N)$  time
```



0	1	2	3	4	5	6	7	8	[9]	10	11
0	36	26	25	17	19	3	1	2	7		

PriorityQueue Implementation (4)

Strategy	Enqueue	Dequeue
Array-Based PQ (1)	$O(N)$	$O(1)$
Array-Based PQ (2)	$O(1)$	$O(N)$
Binary-Heap	Heap_Insert(key) $O(\log N)$	Heap_ExtractMax() $O(\log N)$

Summary so far:

Heap data structure is an efficient data structure -- $O(\log N)$ operations for enqueue/dequeue -- to implement ADT priority queue where 'key' represent the 'priority' of each item

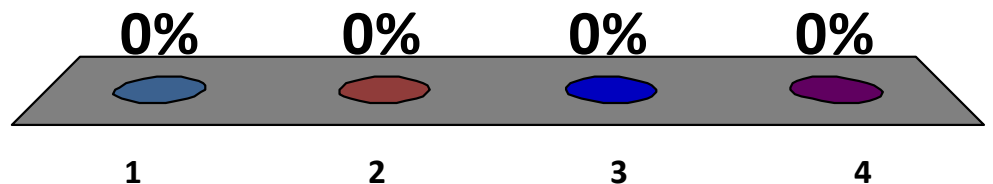
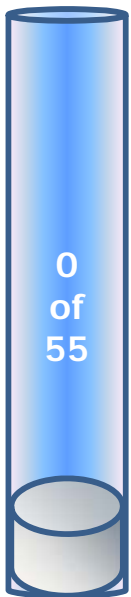
Next Items:

- Heap Sort
- Building Max Heap from an ordinary Array
- Java Implementation of Max Heap

10 MINUTES BREAK

Review: We have seen MergeSort in Lect2. It can sort N items in...

1. $O(N^2)$
2. $O(N \log N)$
3. $O(N)$
4. $O(\log N)$



Heap_Sort Pseudo Code

- With a max heap, we can do sorting too 😊
 - Just call Heap_ExtractMax N times
 - If we don't have a max heap yet, simply build one!

```
Heap_Sort(Array)
  Build_Heap(Array) // O(?)
  N ← size(Array)
  for i from 1 to N // O(N)
    A[N - i + 1] ← Heap_ExtractMax() // O(log N)
  return A
```

```
// Preliminary analysis:
// Heap_Sort runs in O(? + N log N)
```

Build_Heap (Version 1)

```
Build_Heap(Array)
  N ← size(Array)
  A[0] ← 0 // dummy entry
  for i = 1 to N // O(N)
    Heap_Insert(Array[i]) // O(log N)

// Analysis: This runs in O(N log N)
```

- Can we do better?

Build_Heap v1, can we do better?

1. Yes, you must have some more trick
2. No, this is already good enough



Build_Heap (Version 2)

```
Build_Heap(Array)
    heapsize ← size(Array)
    A[0] ← 0 // dummy entry
    for i = 1 to heapsize // copy the content O(N)
        A[i] ← Array[i]
    for i = Parent(heapsize) down to 1 // O(N/2)
        shiftDown(i) // O(log N)

// Analysis: Is this also O(N log N) ??
```

Animation (1)

```
Build_Heap(Array)
```

```
  heapsize  $\leftarrow$  size(Array)
```

```
  A[0]  $\leftarrow$  0
```

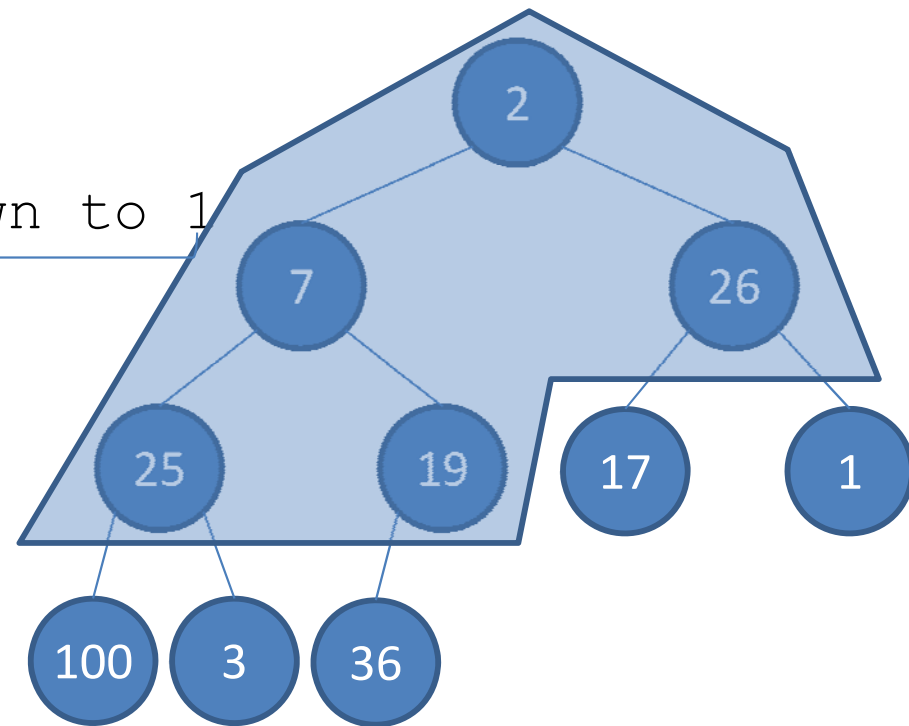
```
  for i = 1 to heapsize
```

```
    A[i]  $\leftarrow$  Array[i]
```

```
  for i = Parent(heapsize) down to 1
```

```
    shiftDown(i)
```

Internal Nodes Only!



0	1	2	3	4	5	6	7	8	9	10	11
0	2	7	26	25	19	17	1	100	3	36	

Animation (2)

```
Build_Heap(Array)
```

```
  heapsize  $\leftarrow$  size(Array)
```

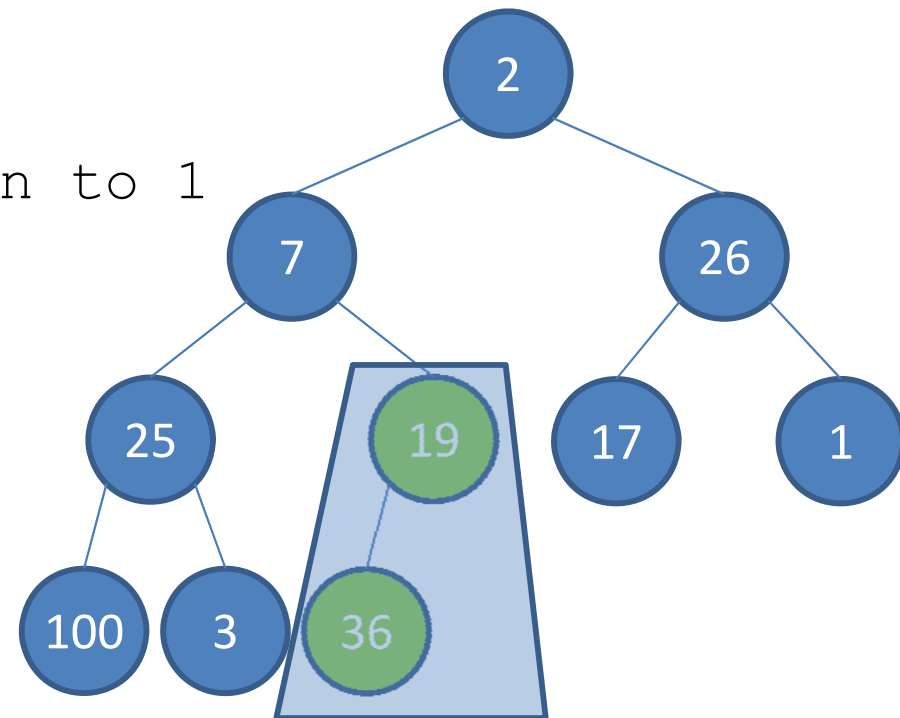
```
  A[0]  $\leftarrow$  0
```

```
  for i = 1 to heapsize
```

```
    A[i]  $\leftarrow$  Array[i]
```

```
  for i = Parent(heapsize) down to 1
```

```
    shiftDown(i)
```



0	1	2	3	4	5	6	7	8	9	10	11
0	2	7	26	25	19	17	1	100	3	36	

Animation (3)

```
Build_Heap(Array)
```

```
  heapsize  $\leftarrow$  size(Array)
```

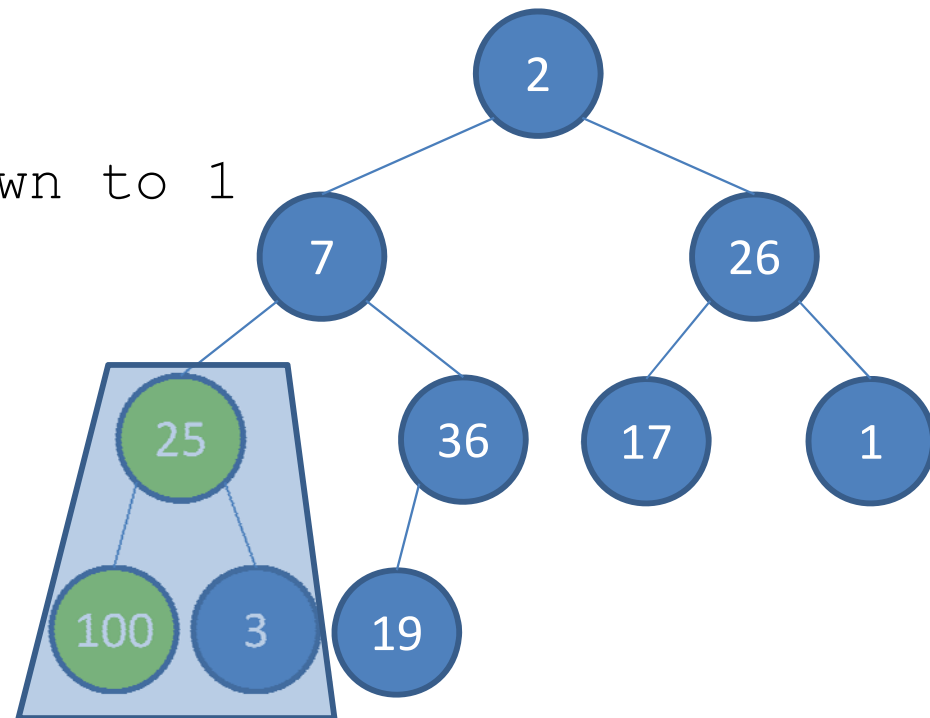
```
  A[0]  $\leftarrow$  0
```

```
  for i = 1 to heapsize
```

```
    A[i]  $\leftarrow$  Array[i]
```

```
  for i = Parent(heapsize) down to 1
```

```
    shiftDown(i)
```



0	1	2	3	4	5	6	7	8	9	10	11
0	2	7	26	25	36	17	1	100	3	19	

Animation (4)

```
Build_Heap(Array)
```

```
  heapsize  $\leftarrow$  size(Array)
```

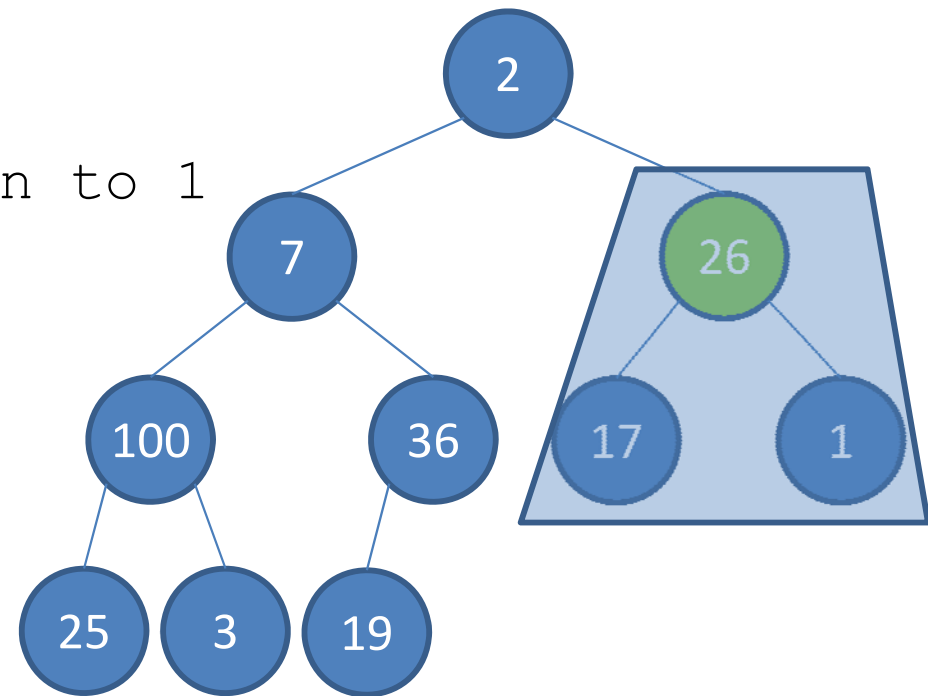
```
  A[0]  $\leftarrow$  0
```

```
  for i = 1 to heapsize
```

```
    A[i]  $\leftarrow$  Array[i]
```

```
  for i = Parent(heapsize) down to 1
```

```
    shiftDown(i)
```



0	1	2	3	4	5	6	7	8	9	10	11
0	2	7	26	100	36	17	1	25	3	19	

Animation (5)

```
Build_Heap(Array)
```

```
  heapsize  $\leftarrow$  size(Array)
```

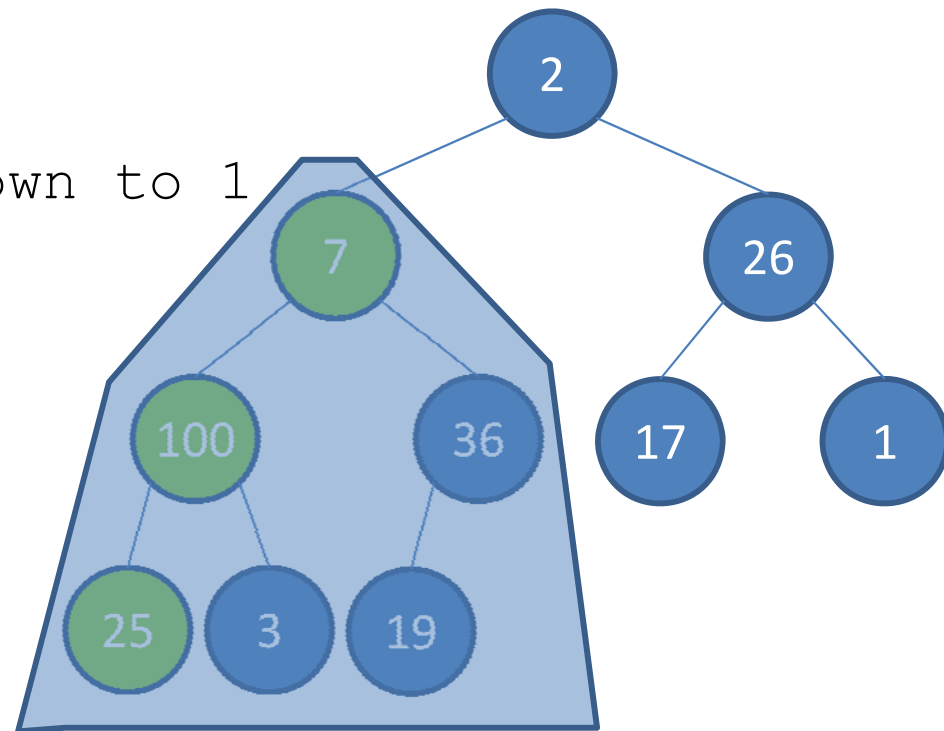
```
  A[0]  $\leftarrow$  0
```

```
  for i = 1 to heapsize
```

```
    A[i]  $\leftarrow$  Array[i]
```

```
  for i = Parent(heapsize) down to 1
```

```
    shiftDown(i)
```



0	1	2	3	4	5	6	7	8	9	10	11
0	2	7	26	100	36	17	1	25	3	19	

Animation (6)

```
Build_Heap(Array)
```

```
  heapsize  $\leftarrow$  size(Array)
```

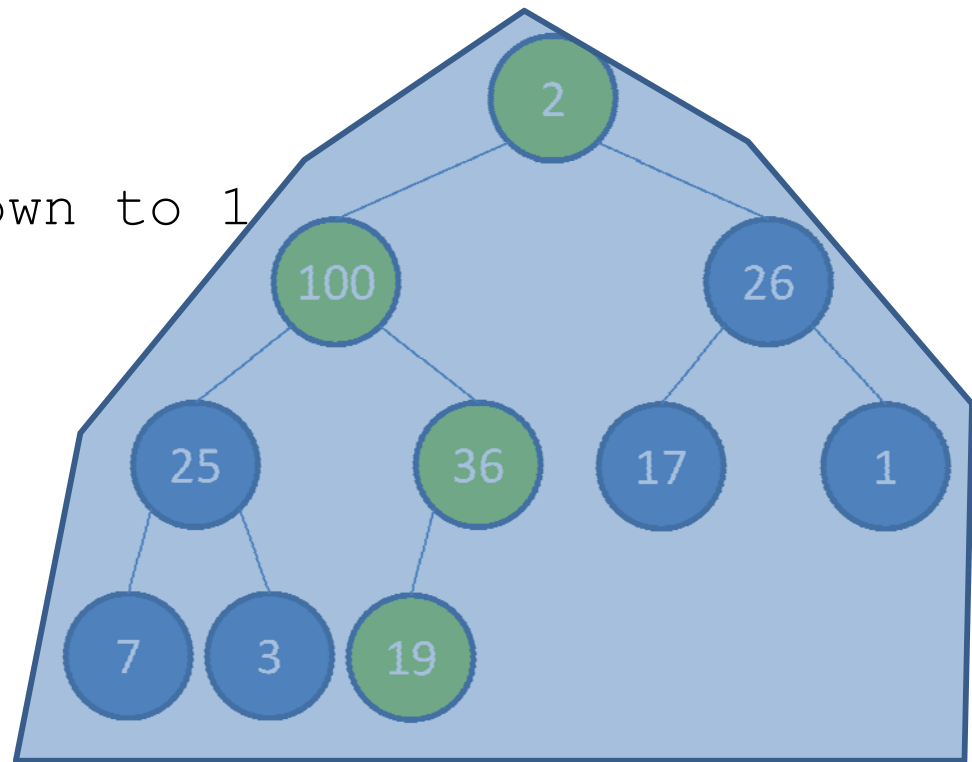
```
  A[0]  $\leftarrow$  0
```

```
  for i = 1 to heapsize
```

```
    A[i]  $\leftarrow$  Array[i]
```

```
  for i = Parent(heapsize) down to 1
```

```
    shiftDown(i)
```



0	1	2	3	4	5	6	7	8	9	10	11
0	2	100	26	25	36	17	1	7	3	19	

Animation (7)

```
Build_Heap(Array)
```

```
  heapsize  $\leftarrow$  size(Array)
```

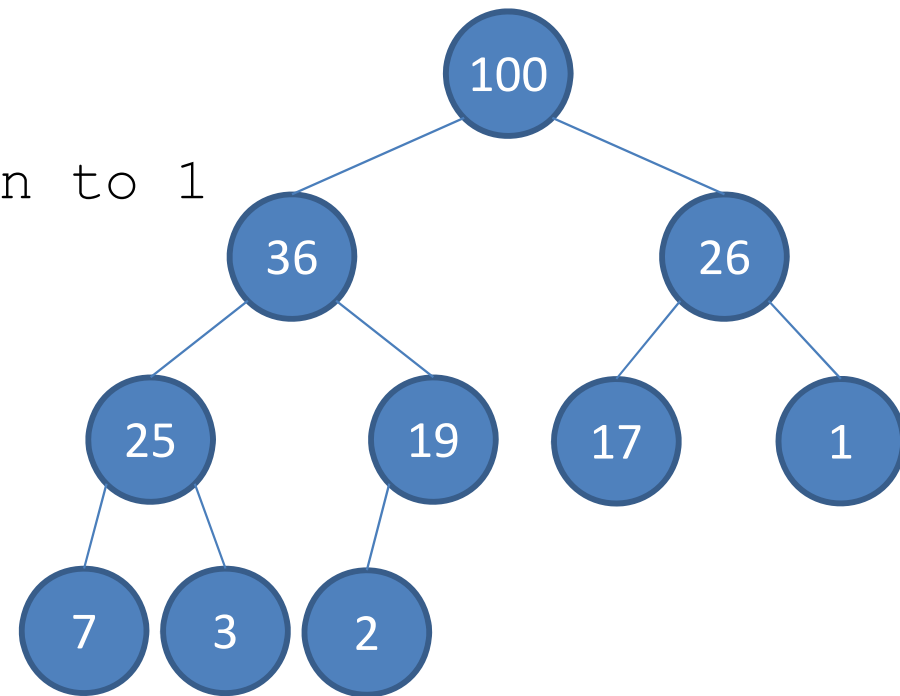
```
  A[0]  $\leftarrow$  0
```

```
  for i = 1 to heapsize
```

```
    A[i]  $\leftarrow$  Array[i]
```

```
  for i = Parent(heapsize) down to 1
```

```
    shiftDown(i)
```



0	1	2	3	4	5	6	7	8	9	10	11
0	100	36	26	25	19	17	1	7	3	2	

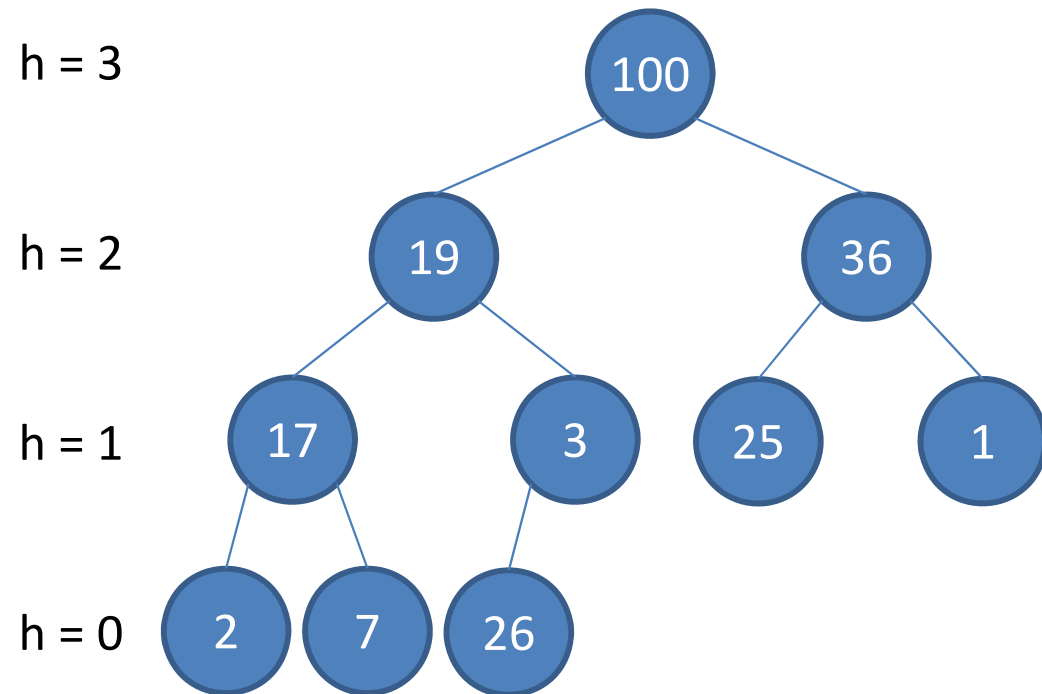
Build-Heap v2 runs in $O(N \log N)$?

1. Yes, obviously
 $O(N \log N)$
2. No, it is _____



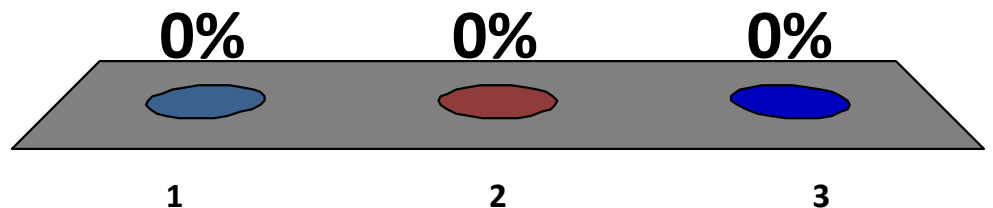
Build-Heap v2 Analysis... (1)

- Recall: How many levels (height) are there in a complete binary tree (heap) of size N ? _____
- Recall: What is the cost to run `shiftDown(i)`? _____
- How many nodes are there at height h of a complete binary tree?



Number of CBT nodes at height h ?

1. $\text{ceil}(n / (h + 1))$
2. $\text{floor}(2^{(h + 1)})$
3. $\text{ceil}(n / 2^{(h + 1)})$



Build-Heap v2 Analysis... (2)

- Cost of Build-Heap v2 is thus:

$$\sum_{h=0}^{\lfloor \lg(n) \rfloor} \underbrace{\left\lceil \frac{n}{2^{h+1}} \right\rceil}_{\substack{\text{\# of} \\ \text{nodes at} \\ \text{height } h}} \underbrace{O(h)}_{\substack{\text{Cost to} \\ \text{Heapify a} \\ \text{node at} \\ \text{height } h}}$$

Sum over all levels Cost for a level

$$= \sum_{h=0}^{\lfloor \lg(n) \rfloor} \left\lceil \frac{n}{2^{h+1}} \right\rceil c * h =$$



Heap-Sort Analysis

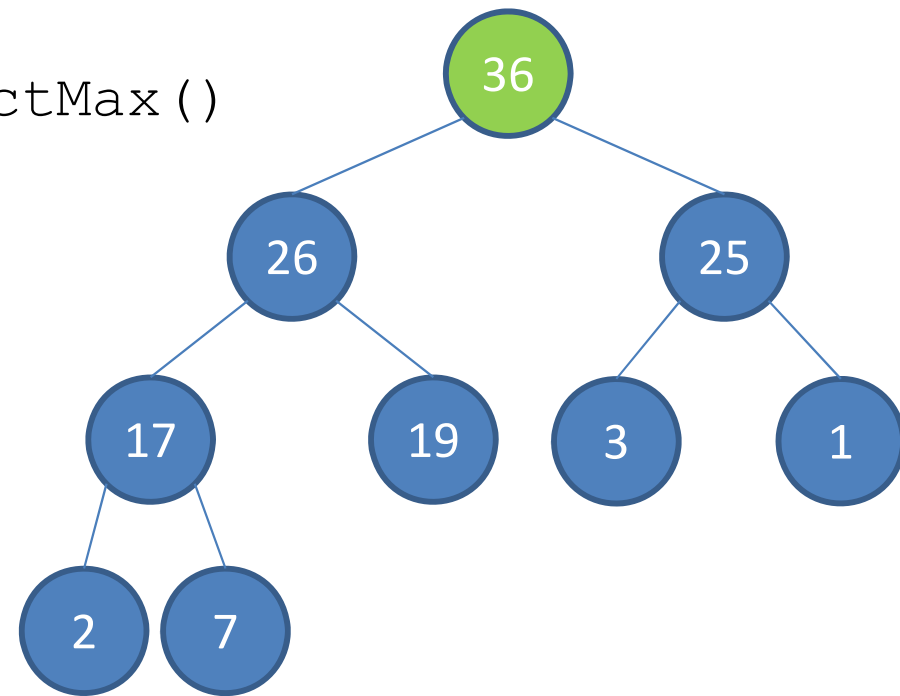
```
Heap_Sort(Array)
  Build_Heap(Array) // The best we can do is _____
  N ← size(Array)
  for i from 1 to N // O(N)
    A[N - i + 1] ← Heap_ExtractMax() // O(log N)
  return A

// Analysis: Thus Heap_Sort runs in O(_____)

// Do you notice that we do not need extra array
// like merge sort to perform sorting?
// Thus heap sort is more memory friendly
// This is called "in-place sorting"
```

Animation (1)

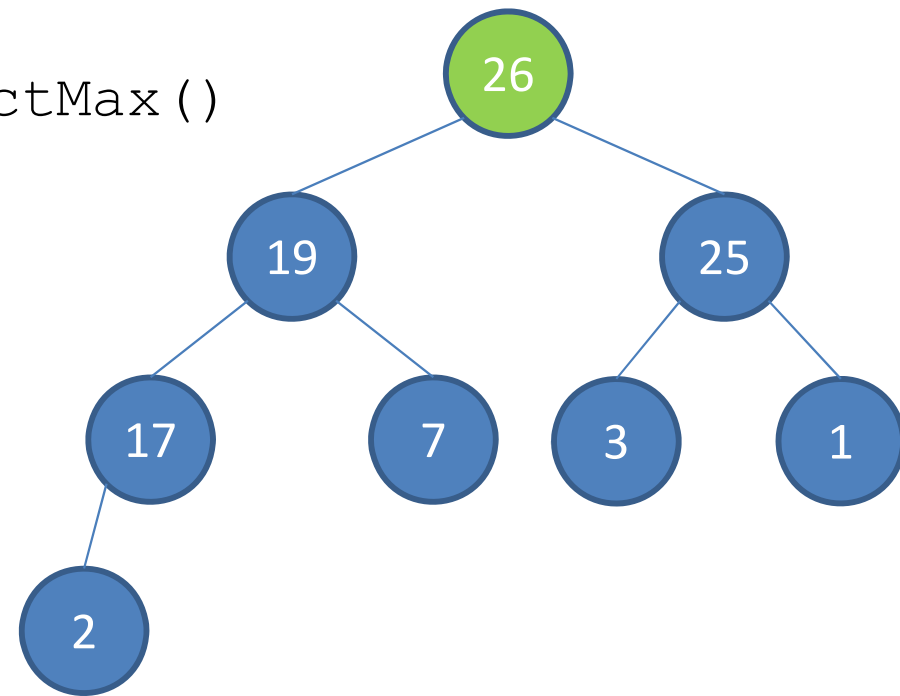
```
Heap_Sort(Array)
  Build_Heap(Array)
  N ← size(Array)
  for i from 1 to N
    A[N - i + 1] ← Heap_ExtractMax()
  return A
```



0	1	2	3	4	5	6	7	8	[9]	10	11
0	36	26	25	17	19	3	1	2	7		

Animation (2)

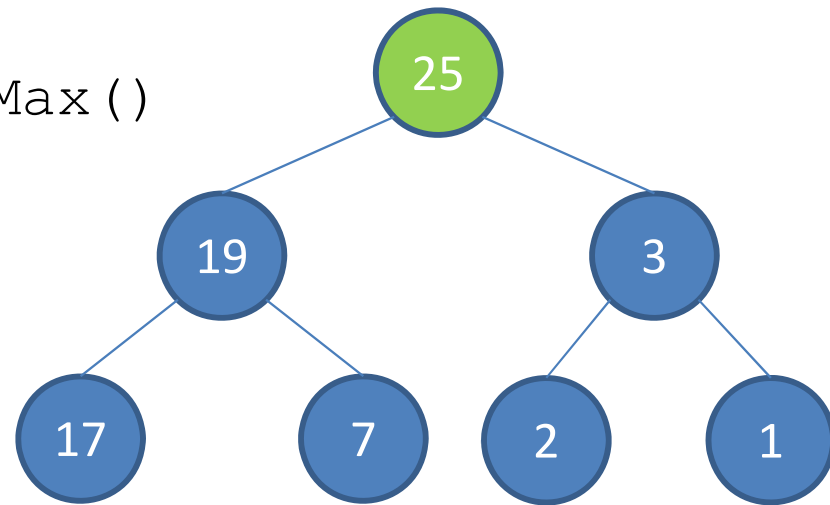
```
Heap_Sort(Array)
  Build_Heap(Array)
  N ← size(Array)
  for i from 1 to N
    A[N - i + 1] ← Heap_ExtractMax()
  return A
```



0	1	2	3	4	5	6	7	[8]	9	10	11
0	26	19	25	17	7	3	1	2	36		

Animation (3)

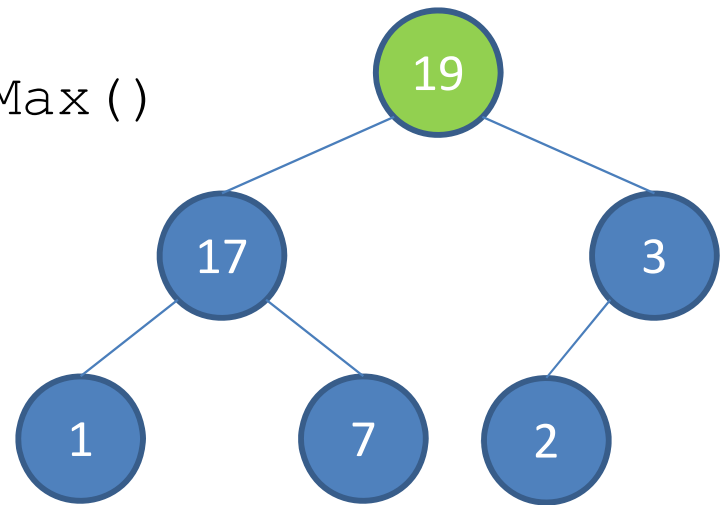
```
Heap_Sort(Array)
  Build_Heap(Array)
  N ← size(Array)
  for i from 1 to N
    A[N - i + 1] ← Heap_ExtractMax()
  return A
```



0	1	2	3	4	5	6	[7]	8	9	10	11
0	25	19	3	17	7	2	1	26	36		

Animation (3)

```
Heap_Sort(Array)
  Build_Heap(Array)
  N ← size(Array)
  for i from 1 to N
    A[N - i + 1] ← Heap_ExtractMax()
  return A
```



And so on until A[1..9] are sorted

0	1	2	3	4	5	[6]	7	8	9	10	11
0	19	17	3	1	7	2	25	26	36		

Java Implementation

- Priority Queue ADT
- Heap Class
 - shiftUp
 - Heap_Insert
 - shiftDown
 - Heap_ExtractMax
 - Build_Heap
 - Heap_Sort
- In OOP Style

PS4

- PS4 will only be released on Tuesday of Week05
- So, enjoy your CNY break 😊

Summary

- In this lecture we looked at:
 - Heap DS and its application for PriorityQueue
 - Storing heap as a compact array and its operations
 - Remember how we always try to maintain complete binary tree and heap property in all our operations!!!
 - Simple application of Heap DS: Heap_Sort
- See you again in the 2nd half of CS2020
 - We will use Heap/PriorityQueue for this algorithm:
 - Dijkstra's algorithm for Single Source Shortest Paths Problem
- Note about Today's Recitation Classes:
 - Still with Dr Seth Gilbert, about Balanced Trees