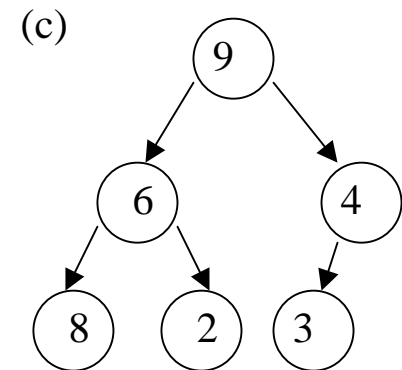
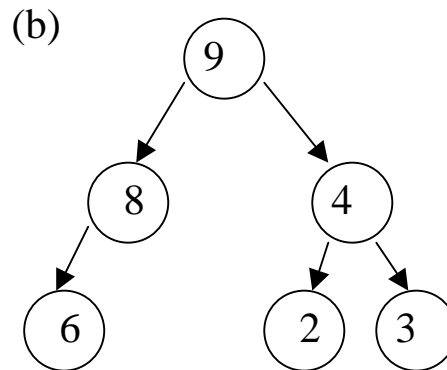
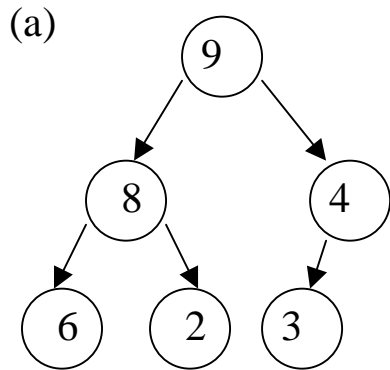


What is a Heap Data Structure?

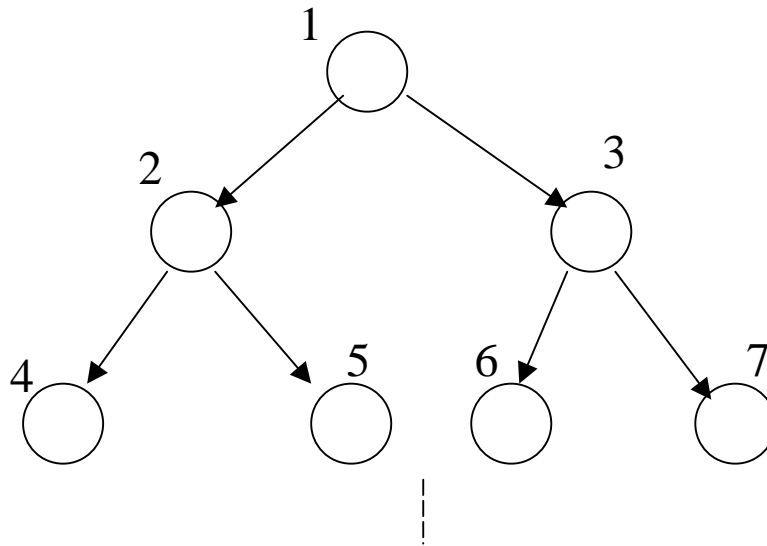
- A Heap data structure is a binary tree with the following properties :
 1. It is a ***complete binary tree***; that is, each level of the tree is completely filled, except possibly the bottom level. At this level, it is filled from left to right.
 2. It satisfies the heap-order property: The data item stored in each node is greater than or equal to the data items stored in its children.
- Examples:



- In the above examples **only (a) is a heap**. (b) is not a heap as it is not complete and (c) is complete but does not satisfy the second property defined for heaps.

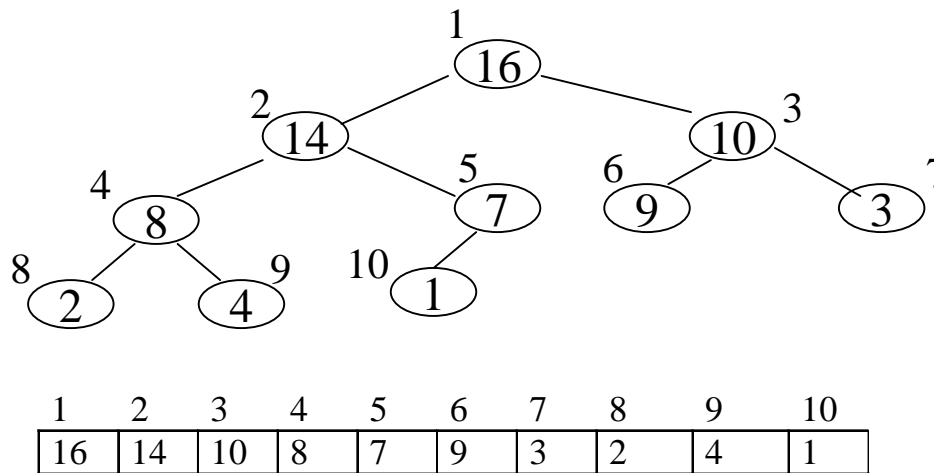
Heap as an Array

- We could implement heaps using a linked list like structure, like we did with binary trees but in this instance it is actually easier to implement heaps using arrays.
- We simply number the nodes in the heap from top to bottom, numbering the nodes on each level from left to right and store the i th node in the i th location of the array. (of course remembering that array indexing in Java starts at zero).



Heap as an Array (2)

- An array A that represents a heap is an array with two attributes
 - *length*, the number of elements in the array
 - *heap-size*, the number of heap elements stored in the array
- Viewed as a binary tree and as an array :



- The root of the tree is stored at A[0], its left-child at A[1], its right child at A[2] etc.

Accessing the Heap Values

- Given the index i of a node, the indices of its parent $Parent(i)$, left-child $LeftChild(i)$ and right child $RightChild(i)$ can be computed simply :

Parent(i)
return $i/2$

LeftChild(i)
return $2i$

RightChild(i)
return $2i+1$

- Heaps must also satisfy the **heap property** for every node, i , other than the root.

$$A[Parent(i)] \geq A[i]$$

- Therefore, the largest element in a heap is stored at the root, and the subtrees rooted at a node contain smaller values than does the node itself.

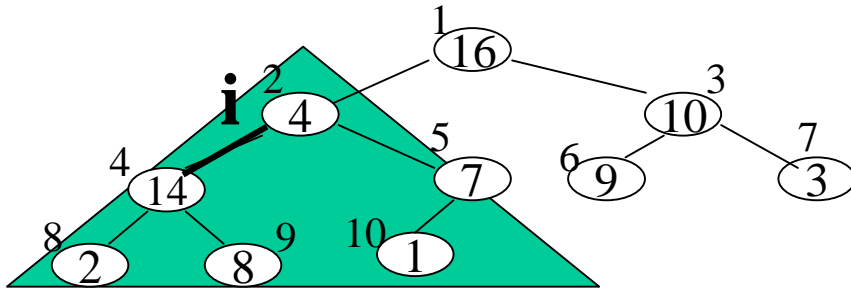
Heap Operations

- The **height** of a node in a tree is the number of edges on the longest simple downward path from the node to a leaf. (i.e. maximum depth from that node – Remember?)
- The height of an n -element heap based on a binary tree is $\lg n$
- The basic operations on heaps run in time at most proportional to the height of the tree and thus take $O(\lg n)$ time.

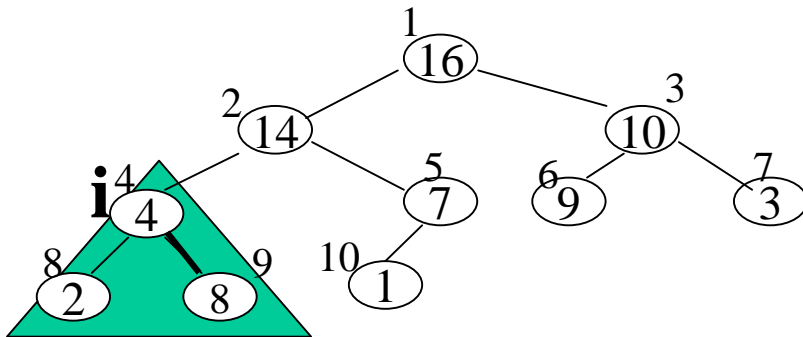
Maintaining the Heap Property

- One of the more basic heap operations is converting a complete binary tree to a heap.
- Such an operation is called “Heapify”.
- Its inputs are an array A and an index i into the array.
- When Heapify is called, it is assumed that the binary trees rooted at $\text{LeftChild}(i)$ and $\text{RightChild}(i)$ are heaps, but that $A[i]$ may be smaller than its children, thus violating the 2nd heap property.
- The function of Heapify is to let the value at $A[i]$ “float down” in the heap so that the subtree rooted at index i becomes a heap.
- The action required from Heapify is as follows:

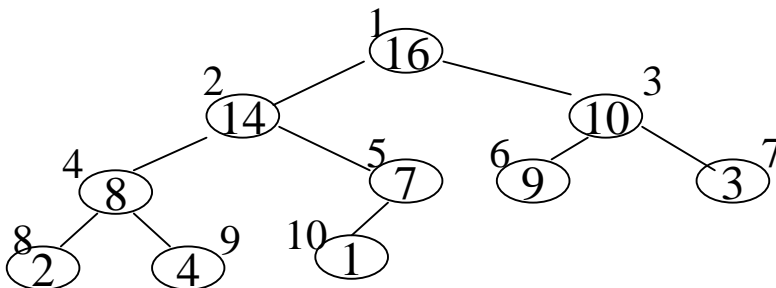
Example of Heapify



- Subtree to be heapified is marked by the shaded triangle.
- Node **i** if smaller than its children is swapped with the greater of its two children – in this case the left child with value 14 is swapped for value 4 in node **I**.



- If the next subtree down (again marked with the shaded triangle region) is not a heap, we “percolate” down to the next subtree level and find the larger of the two children again to swap with current root node.



- We keep “percolating” down until either the subtree conforms to the heap properties or a leaf has been reached.

The Heapify Algorithm

Begin

```
Heapify(A[], i)           // Heapify takes in our heap and index of current root node of subtree to be heapified
{
    left = LeftChild(i);           // index of left child
    right = RightChild(i);         // index of right child

    if left ≤ A.heap-size AND A[left] > A[i]    { // if still within bounds AND left child
        largest = left;                       // greather than parent – remember largest as left child
    }
    else {
        largest = i;                          // else parent still has largest value for now
    }

    if right ≤ A.heap-size AND A[right] > A[largest] { // if still within bounds AND right child
        largest = right                       // greather than parent – remember largest as right child
    }

    if largest NOT EQUAL i {                  // if parent does not hold largest value
        swap A[i] and A[largest]             // swap parent with child that does have largest value
        Heapify(A, largest)                  // Percolate down and Heapify next subtree
    }
}
End
```


The Heapify Algorithm

- At each step, the index of the largest of the elements $A[i]$, $A[\text{Left}(i)]$, and $A[\text{Right}(i)]$ is stored in the variable `largest`.
- If $A[i]$ is largest, then the subtree rooted at node i is a heap and the procedure ends.
- Otherwise, one of the two children has the largest element, and $A[i]$ is swapped with $A[\text{largest}]$, which causes node i and its children to satisfy the heap property.
- The node `largest`, however, now has the original value $A[i]$, and thus the subtree rooted at `largest` may violate the heap property.
- Therefore, `Heapify` must be called recursively on that subtree.

Analysis of Heapify

- The running time of Heapify on a subtree of size n rooted at a given node i is the $O(1)$ time to fix up the relationships among the elements $A[i]$, $A[\text{Left}(i)]$ and $A[\text{Right}(i)]$, plus the time to run Heapify on a subtree rooted at one of the children of node i .
- The subtrees can have size of at most $2n/3$, and the running time of Heapify can therefore be described by the recurrence,

$$T(n) \leq T(2n/3) + O(1)$$

- Using either the iteration method or the master's theorem we can find that the solution to this works out as:

$$T(n) = O(\log n)$$

Building a Heap

- For the general case of converting a complete binary tree to a heap, we begin at the last node that is not a leaf, apply the “percolate down” routine to convert the subtree rooted at this current root node to a heap.
- We then move onto the preceding node and “percolate down” that subtree.
- We continue on in this manner, working up the tree until we reach the root of the given tree.
- We use the Heapify procedure here in a bottom up manner. This means also means that to convert an array $A[1..n]$, where $n = A.length$, into a heap we can apply the same procedure as described above.
- Now, since the subarray $A[(\lfloor n/2 \rfloor + 1)..n]$ are all leaves of the tree, each is a 1-element heap.
- The procedure “Build-Heap” goes through the remaining nodes and runs Heapify on each.
- The order of processing guarantees that the subtrees rooted at children of node i are heaps before Heapify is run at that node.

Build Heap Algorithm

Begin:

```
Build-Heap (A[])    {           // takes in an array to be heapified
    A.heap-size = A.length;      // heap size is number of elements in the array
    for  $i = \lfloor A.length/2 \rfloor$  downto 1 // starting at first subtree up to root full tree
    {
        Heapify(A, i);           // Heapify the current subtree
    }
}
```

End

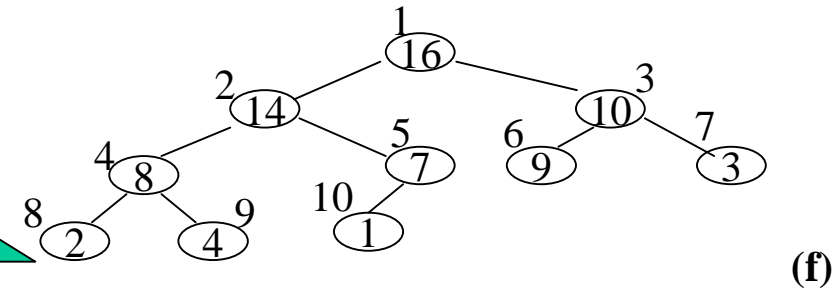
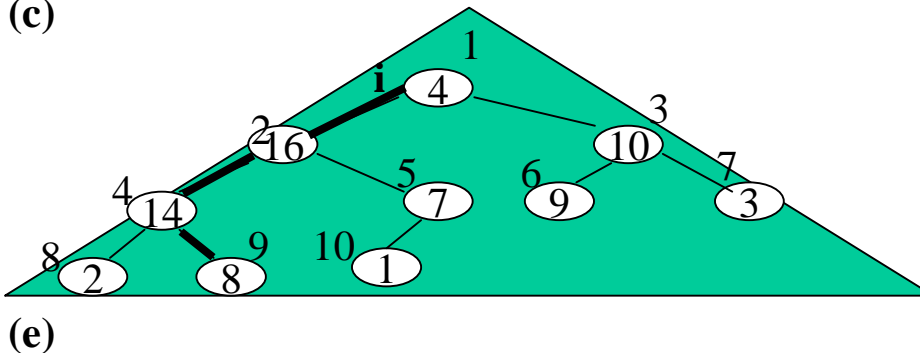
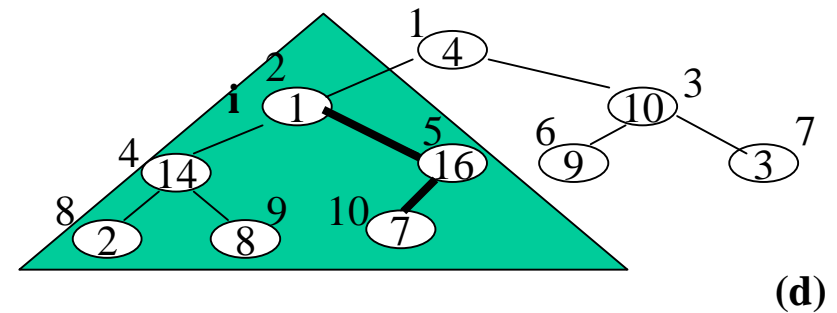
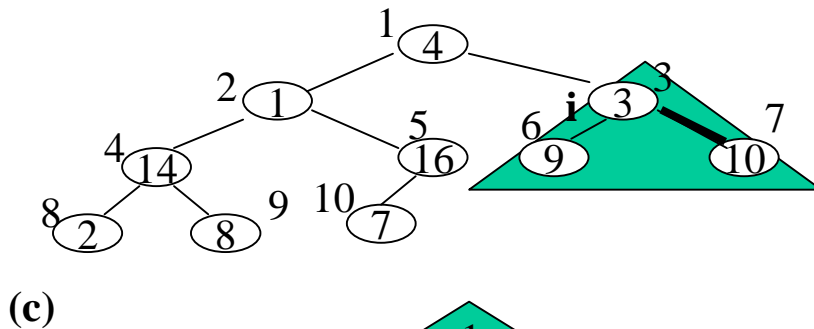
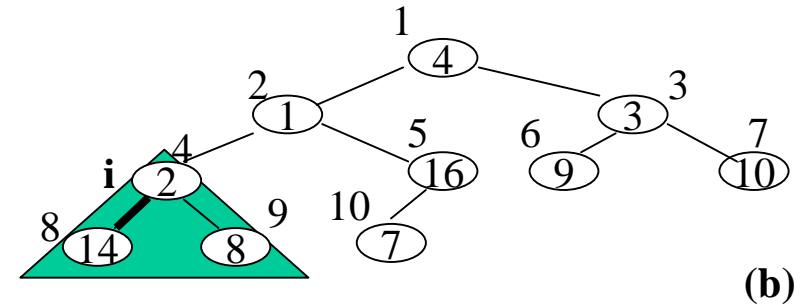
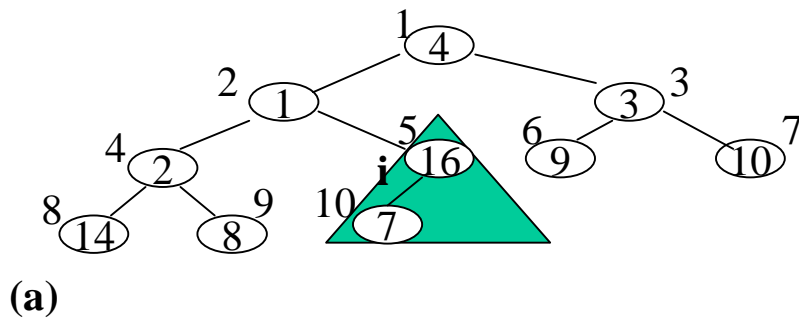
- Each call to Heapify costs $O(\lg n)$ time, and there are $O(n)$ such calls.
- Thus, the running time is at most $O(n \lg n)$
- A more complex analysis, however, gives us a tighter upper bound of $O(n)$.
- Hence, we can build a heap from an unordered array in linear time.

- Lets have a look at a visual example of this operation (have a look at the next slide).
- The current subtrees that are being worked on are marked inside a shaded triangle.
- Any swapping that goes on is marked with a thicker edge.

Example of Build Heap

A

4	1	3	2	16	9	10	14	8	7
---	---	---	---	----	---	----	----	---	---



Heap Practical Example :Priority Queue

- Many applications require that we process records with keys/priorities in order, but not necessarily in full sorted order. (e.g. CPU process scheduling).
- Items in a priority queue are not processed strictly in order of entry into the queue
- Items can be placed on the queue at any time, but processing always takes the item with the largest key/ priority
- The priority queue is a generalised queue Abstract Data Structure (ADT)
- In fact, the priority queue is a proper generalisation of the stack and the queue ADTs, as we can implement either with priority queues, using appropriate priorities

Priority Queue - Definition

- A priority queue is a data structure of items with keys that support two basic operations : insert a new item, and delete the item with the largest key.

6	5	5	3	2	1
---	---	---	---	---	---

6	6	5	5	3	2	1
---	---	---	---	---	---	---

Add item with priority 6

6	6	5	5	4	3	2	1
---	---	---	---	---	---	---	---

Add item with priority 4

6	5	5	4	3	2	1
---	---	---	---	---	---	---

Remove item with highest priority

- Applications of priority queues include:
 - Simulation Systems, where events need to be processed chronologically.
 - Job scheduling in computer systems.
- The Heap Data Structure is an efficient structure for implementing a simple priority queue.
- In practice, priority queues are more complex than the simple definition above.

Priority Queue

- One of the reasons that priority queue implementations are so useful is their flexibility
- For that reason any implementation will need to support some of the following operations
 - *Construct* a priority queue from N given items
 - *Insert* a new item
 - *Delete the maximum* item
 - *Change the priority* of an arbitrary specified item
 - *Delete* an arbitrary specified item
- We take “maximum” to mean “any record with the largest key value”.

Other Heap Operations

- Other important heap operations include :
 - Heap Insert (inserting an element into a heap)
 - Heap Extract Max/Min (extracting an Max/Min element from a heap)
 - Heap Delete (remove an element from a heap)
 - And of course.... HeapSort - $O(N \lg N)$ sorting algorithm
- + very basic idea :
 - * swap and remove root value with final position element value
 - * decrement size of heap and Heapify from position one (i.e. new root) again to restore new heap
 - * Keep doing this while there is more nodes. This eventually gives us a sorted list.
- Unfortunately, due to time constraints, we cannot examine these operations in any more detail. ☹️