

Lecture 7: Binary Heaps, Heapsort, Union-Find

Instructor: Saravanan Thirumuruganathan

- ① Data Structures to speed up algorithms
 - Binary Heap
 - Heapsort
 - Union Find

- **URL:** `http://m.socrative.com/`
- **Room Name:** **4f2bb99e**

Key Things to Know for Data Structures

- Motivation
- Distinctive Property
- Major operations
- Key Helper Routines
- Representation
- Algorithms for major operations
- Applications

Data Structures for Algorithmic Speedup

- BST and RBT are two examples of data structures to represent dynamic set
- Today's topic, Heap and Union-Find, can also be used to represent dynamic set
- However, these are used more often to speed up algorithms

Binary Heap

Motivation

- Heap Sort (CLRS is organized that way!)
- Priority Queue
- Most space efficient data structure

Priority Queue

- “Queue” data structure has a FIFO property
- Some times it is useful to consider **priority**
- Output element with highest priority first

Priority Queue - Major Operations

- Insert
- FindMin (resp. FindMax)
- DeleteMin (resp. DeleteMax)
- DecreaseKey (resp. IncreaseKey)

Priority Queue - Applications¹

- Dijkstra's shortest path algorithm
- Prim's MST algorithm
- Heapsort
- Online median
- Huffman Encoding
- A* Search (or any Best first search)
- Discrete event simulation
- CPU Scheduling
- ...
- See Wikipedia entry for priority for details

¹Kleinberg-Tardos Book and Wikipedia

Priority Queue - Candidate Implementations

- Assume: for DeleteMin and DecreaseKey, pointer to element is given
- LinkedList
 - Insert:

Priority Queue - Candidate Implementations

- Assume: for DeleteMin and DecreaseKey, pointer to element is given
- LinkedList
 - Insert: $O(1)$
 - FindMin:

Priority Queue - Candidate Implementations

- Assume: for DeleteMin and DecreaseKey, pointer to element is given
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Priority Queue - Candidate Implementations

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Priority Queue - Candidate Implementations

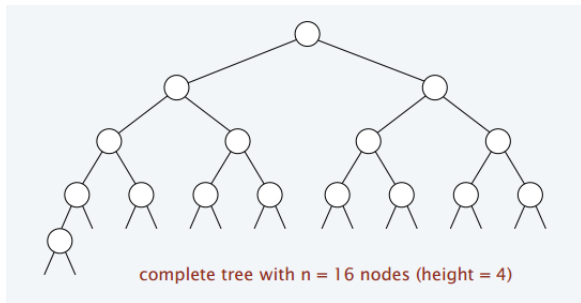
- Assume: for DeleteMin and DecreaseKey, pointer to element is given
- LinkedList
 - Insert: $O(1)$
 - FindMin: $O(n)$
 - DeleteMin: $O(1)$
 - DecreaseKey: $O(1)$
- Binary Heap
 - Insert: $O(\lg n)$
 - FindMin: $O(1)$
 - DeleteMin: $O(\lg n)$
 - DecreaseKey: $O(\lg n)$
- Binomial Heaps, Fibonacci Heaps etc.

Binary Heaps

- Perfect data structure for implementing Priority Queue
- MaxHeap and MinHeap
- We will focus on MaxHeaps in this lecture

Complete Tree²

- Perfectly balanced, except for bottom level
- Elements were inserted top-to-bottom and left-to-right

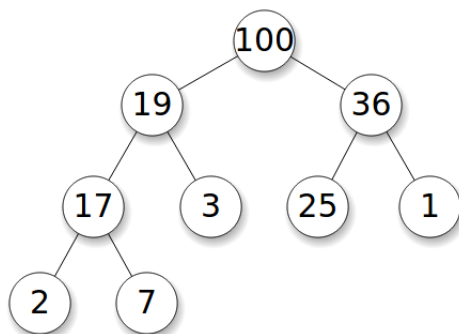


²<http://www.cs.princeton.edu/courses/archive/spring13/cos423/lectures/BinomialHeaps.pdf>

Heap Property

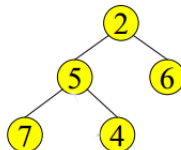
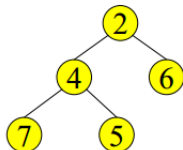
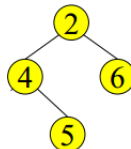
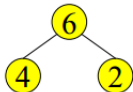
- Heap is a binary tree (**NOT BST**)
- Heap:
 - **Completeness** Property: Heap has restricted structure. It must be a complete binary tree .
 - **Ordering** Property: Relates parent value with that of its children
- MaxHeap property: Value of parent must be greater than **both** its children
- MinHeap property: Value of parent must be less than **both** its children
- Heap with n elements has height $O(\lg n)$

Max Heap Example³



³Wikipedia page for Heap

Heap Property⁴



⁴<http://courses.cs.washington.edu/courses/cse373/06sp/handouts/lecture10.pdf>

Major Operations

- Insert
- FindMax
- DeleteMax (aka ExtractMax)
- IncreaseKey

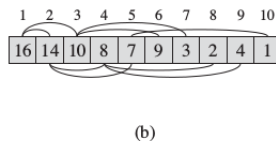
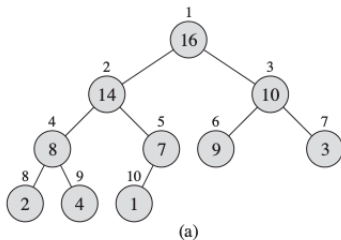
Key Helper Routines

- Max-Heapify (or Min-Heapify)
- Bubble-Up
- Bubble-Down
- Heapify

Representation: Arrays

- Very efficient implementation using arrays
- Possible due to completeness property
- $\text{Parent}(i)$: return $\lfloor i/2 \rfloor$
- $\text{LeftChild}(i)$: return $2i$
- $\text{RightChild}(i)$: return $2i + 1$

Representation: Arrays⁵

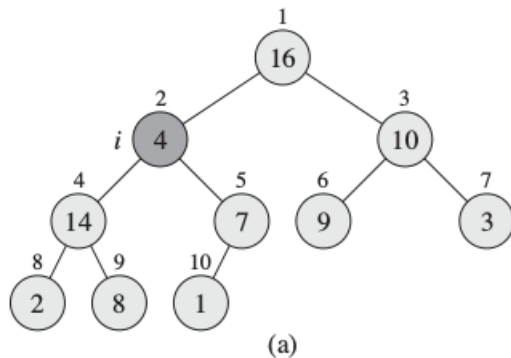


⁵CLRS Fig 6.1

Max-Heapify

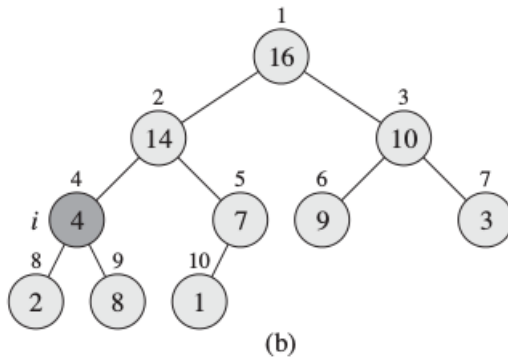
- Objective: Maintain heap property
- Invocation: $\text{Max-Heapify}(A, i)$
- Assume: $\text{Left}(i)$ and $\text{Right}(i)$ are valid max-heaps
- $A[i]$ might violate max-heap property
- Bubble-Down the violation
- **Analysis:** $O(\lg n)$

Max-Heapify: Example⁶



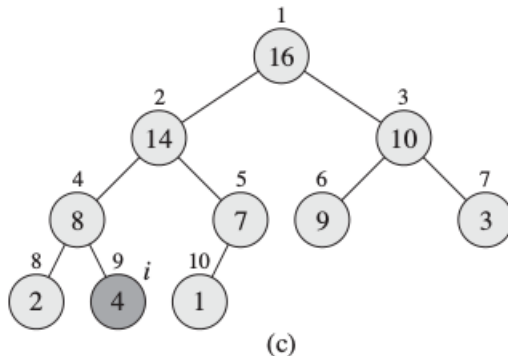
⁶CLRS Fig 6.2

Max-Heapify: Example⁷



⁷CLRS Fig 6.2

Max-Heapify: Example⁸

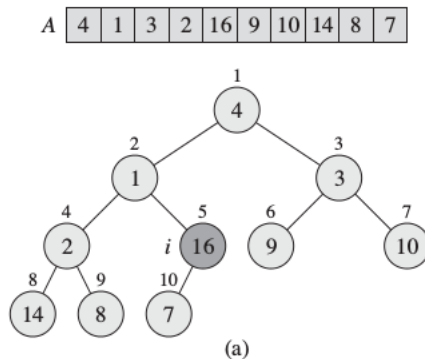


⁸CLRS Fig 6.2

Build-Max-Heap

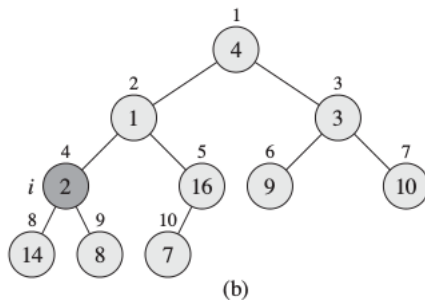
- Given an array A , convert it to a max-heap
- $A.length$: Length of the array
- $A.heapSize$: Elements from $1 \dots A.heapSize$ form a heap
- $Build_Max_Heap(A)$:
 - $A.heapSize = A.length$
 - for $i = \lfloor A.length/2 \rfloor$ down to 1
 $Max_Heapify(A, i)$
- **Analysis:** $O(n)$ (See book for details)

Build-Max-Heap : Example ⁹

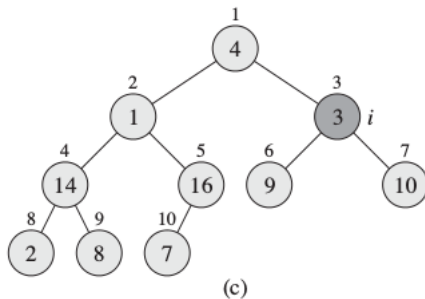


⁹CLRS Fig 6.3

Build-Max-Heap : Example ¹⁰

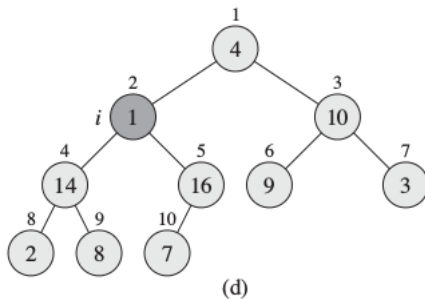


Build-Max-Heap : Example ¹¹



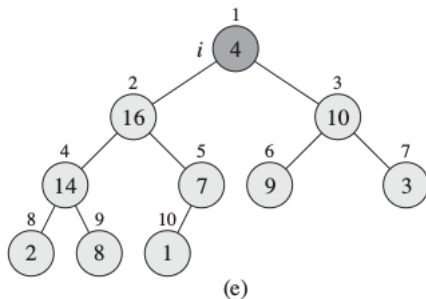
¹¹CLRS Fig 6.3

Build-Max-Heap : Example ¹²

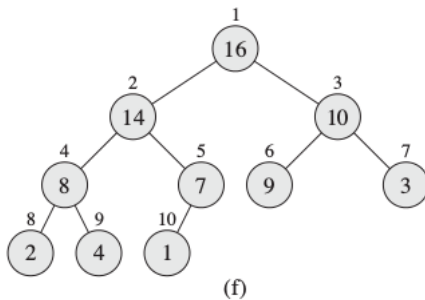


¹²CLRS Fig 6.3

Build-Max-Heap : Example ¹³

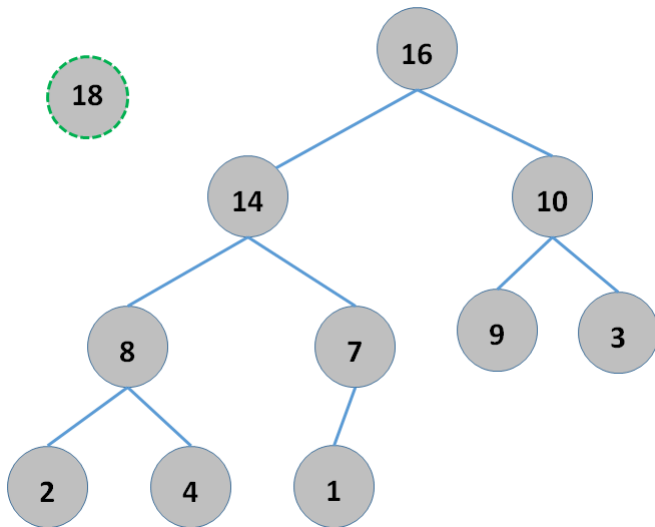


Build-Max-Heap : Example ¹⁴

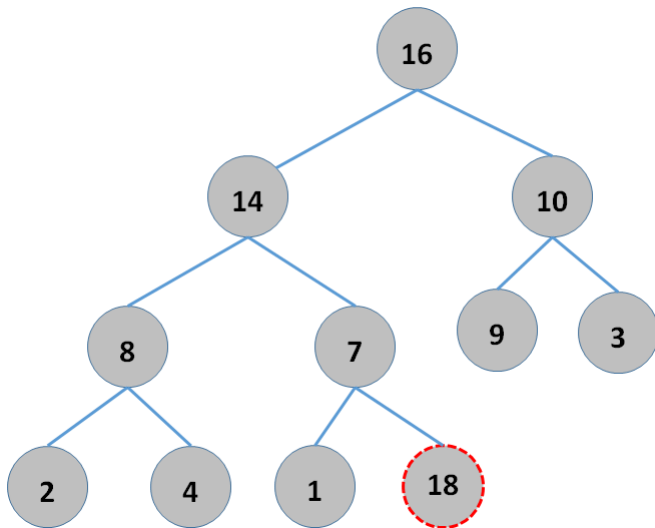


¹⁴CLRS Fig 6.3

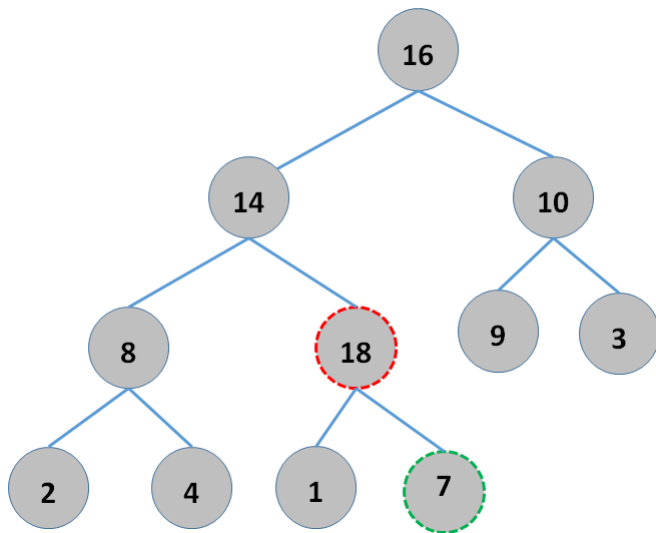
Heap : Insert



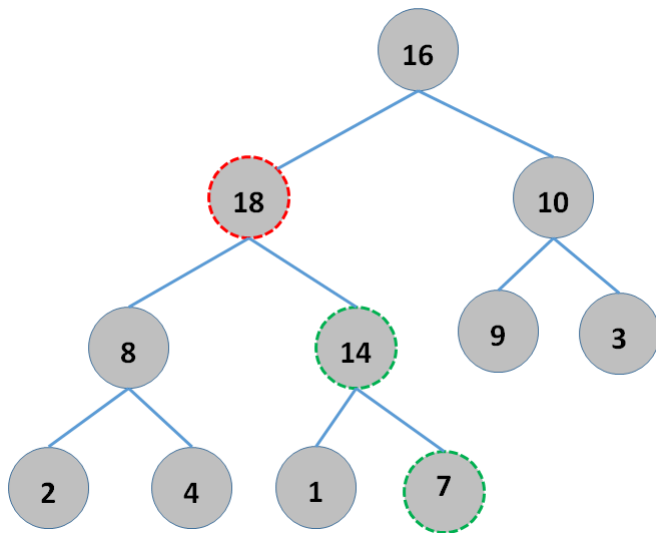
Heap : Insert



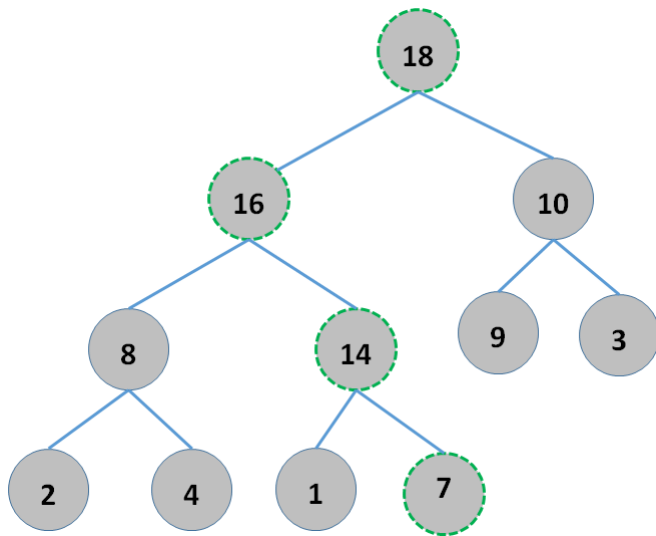
Heap : Insert



Heap : Insert



Heap : Insert



Heap : Insert

- Insert element at first available slot (no completeness property violation!)
- Fix heap property violations by bubbling up the violation till it is fixed
- Complexity:

Heap : Insert

- Insert element at first available slot (no completeness property violation!)
- Fix heap property violations by bubbling up the violation till it is fixed
- Complexity: $O(\lg n)$

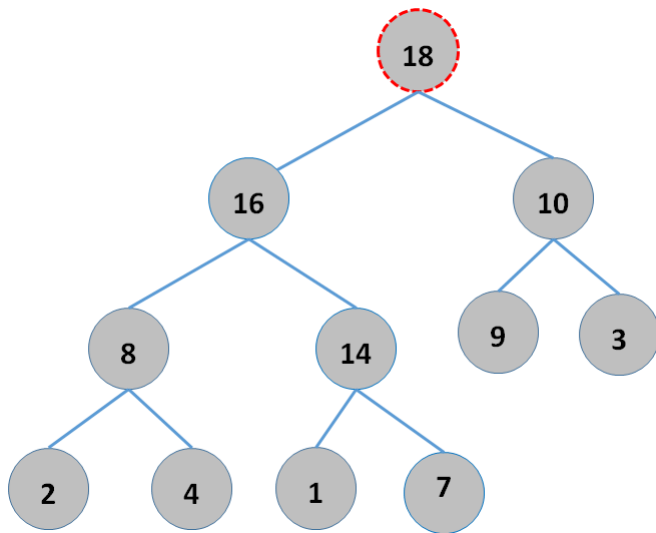
Heap : FindMax

- Look at the root element
- Time complexity: $O(1)$

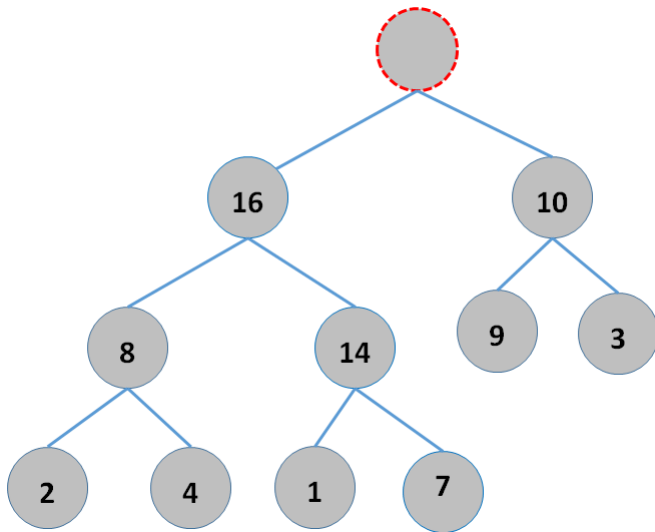
Heap : DeleteMax

- Delete the maximum element (root)
- Fix the heap

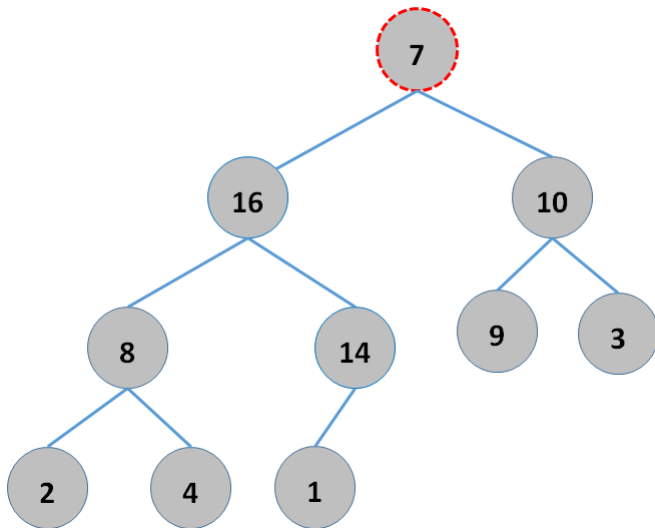
Heap : DeleteMax



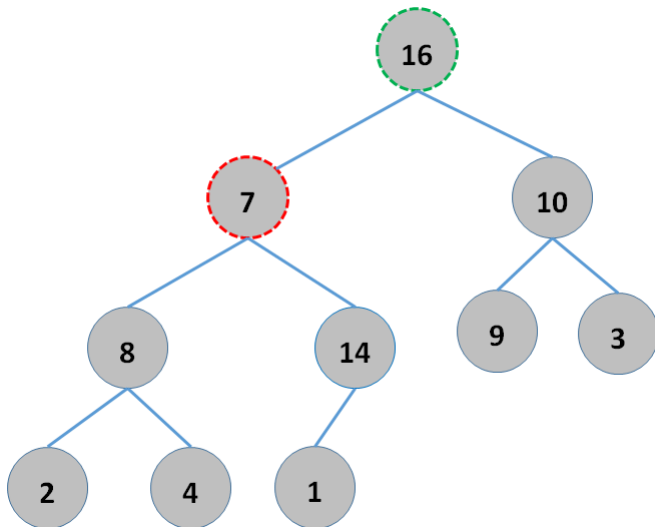
Heap : DeleteMax



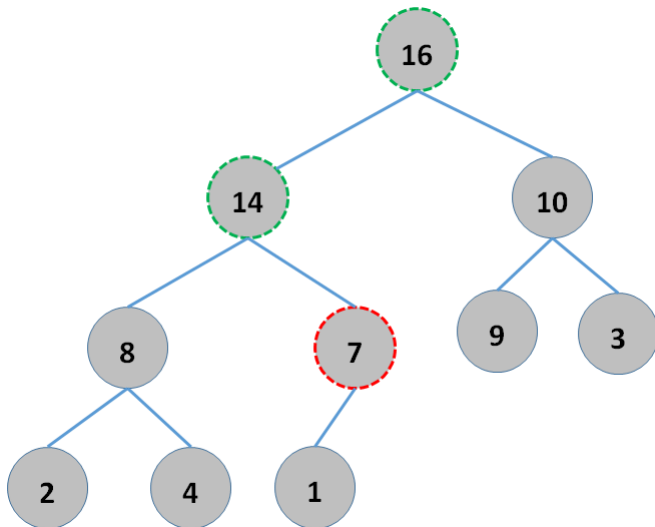
Heap : DeleteMax



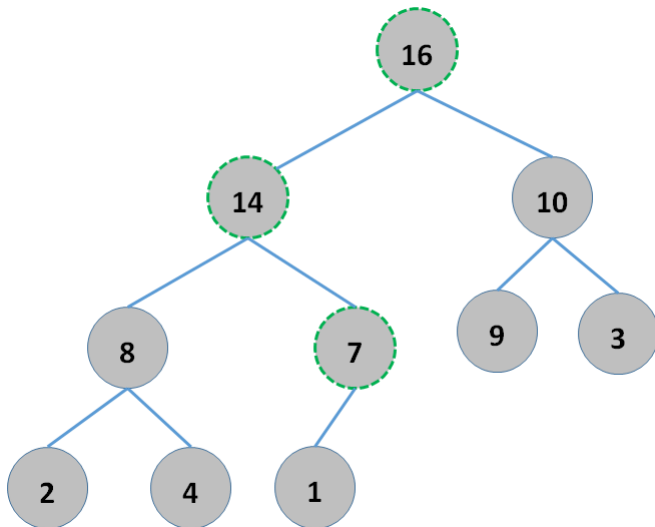
Heap : DeleteMax



Heap : DeleteMax



Heap : DeleteMax



Heap : DeleteMax

- Remove root
- Replace root with last element (does not affect Completeness property)
- Fix heap violations by bubbling it down till it is fixed
- Complexity:

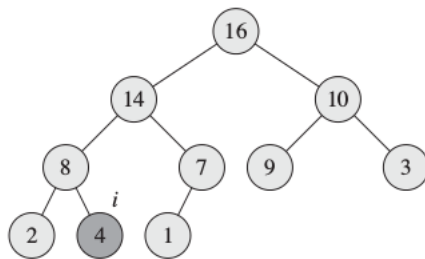
Heap : DeleteMax

- Remove root
- Replace root with last element (does not affect Completeness property)
- Fix heap violations by bubbling it down till it is fixed
- Complexity: $O(\lg n)$

Heap : IncreaseKey

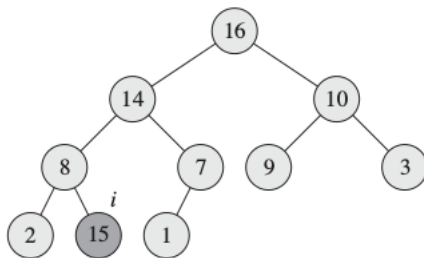
- Given a node, increase its priority to a new, higher value
- Fix heap property violations

Heap : IncreaseKey¹⁵



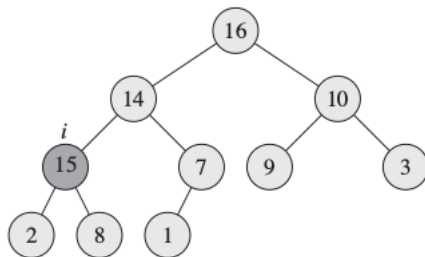
(a)

Heap : IncreaseKey¹⁶



(b)

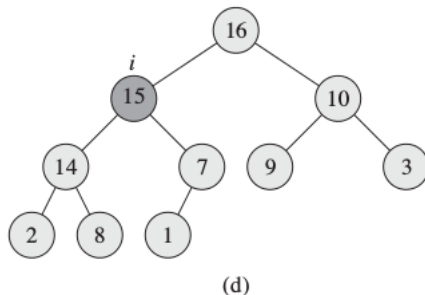
Heap : IncreaseKey¹⁷



(c)

¹⁷CLRS Fig 6.5

Heap : IncreaseKey¹⁸



¹⁸CLRS Fig 6.5

Heap : IncreaseKey

- Update element
- Fix heap violations by bubbling it up till it is fixed
- Complexity:

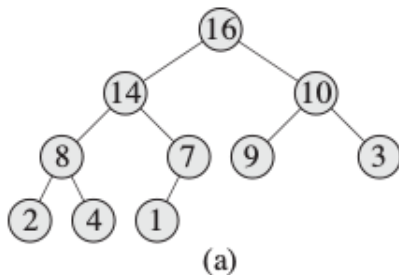
Heap : IncreaseKey

- Update element
- Fix heap violations by bubbling it up till it is fixed
- Complexity: $O(\lg n)$

HeapSort

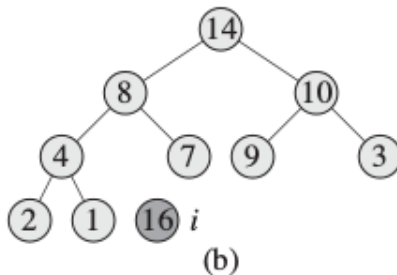
```
HeapSort(A):  
    Build-Max-Heap(A)  
    for i = A.length down to 2  
        Exchange A[1] with A[i]  
        A.heapSize = A.heapSize - 1  
        Max-Heapify(A, 1)
```

Heap Sort: Example¹⁹



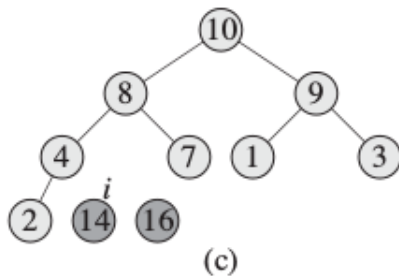
¹⁹CLRS Fig 6.4

Heap Sort: Example²⁰



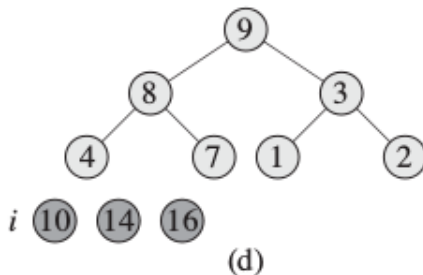
²⁰CLRS Fig 6.4

Heap Sort: Example²¹



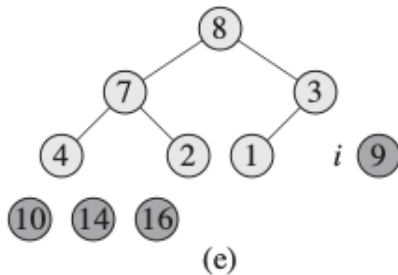
²¹CLRS Fig 6.4

Heap Sort: Example²²



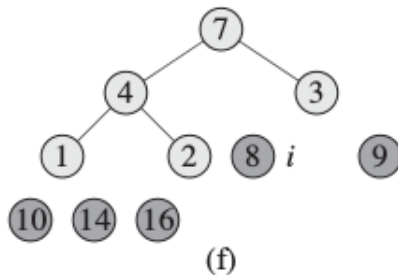
²²CLRS Fig 6.4

Heap Sort: Example²³



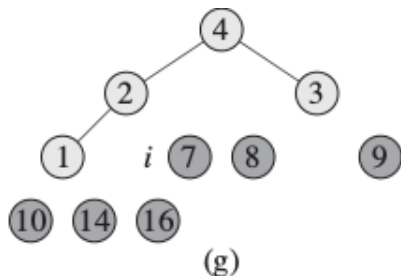
²³CLRS Fig 6.4

Heap Sort: Example²⁴



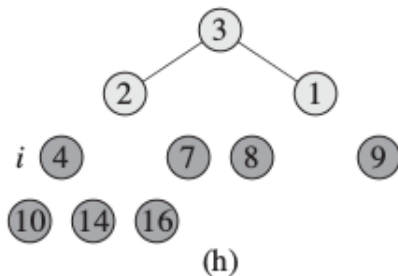
²⁴CLRS Fig 6.4

Heap Sort: Example²⁵



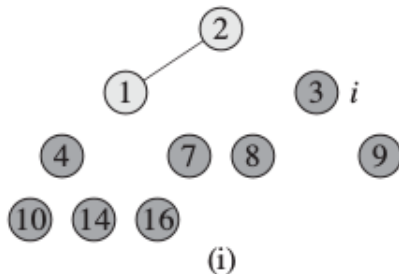
²⁵CLRS Fig 6.4

Heap Sort: Example²⁶



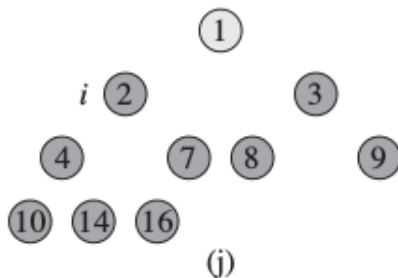
²⁶CLRS Fig 6.4

Heap Sort: Example²⁷



²⁷CLRS Fig 6.4

Heap Sort: Example²⁸



²⁸CLRS Fig 6.4

Heap Sort: Example²⁹

A

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

²⁹CLRS Fig 6.4

HeapSort: Analysis

- Operations:
 - Build-Max-Heap:

HeapSort: Analysis

- Operations:
 - Build-Max-Heap: $O(n)$
 - n Max-Heapify:

HeapSort: Analysis

- Operations:
 - Build-Max-Heap: $O(n)$
 - n Max-Heapify: $n \times \lg n = O(n \lg n)$
 - Complexity: $O(n) + O(n \lg n) = O(n \lg n)$

HeapSort

- Very efficient in practice - often competitive with QuickSort
- In-Place but not stable (why?)
- Requires constant extra space
- Best, average and worst case complexity is $O(n \lg n)$ (unlike Quicksort)

Data Structures for Disjoint Sets

Disjoint Sets ADT

- Objective: Represent and manipulate disjoint sets (sets that do not overlap)
- Required Operations
 - $\text{MakeSet}(x)$: Create a new set $\{x\}$ with single element x
 - $\text{Find}(x)$: Find the set containing x
 - $\text{Union}(x, y)$: Merge sets containing x and y

Disjoint Sets: Example³⁰

- Objects.

```
0 1 2 3 4 5 6 7 8 9
```

- Disjoint sets of objects.

```
0 1 { 2 3 9 } { 5 6 } 7 { 4 8 }
```

- **Find** query: are objects 2 and 9 in the same set?

```
0 1 { 2 3 9 } { 5 6 } 7 { 4 8 }
```

- **Union** command: merge sets containing 3 and 8.

```
0 1 { 2 3 4 8 9 } { 5 6 } 7
```

³⁰<https://www.cs.princeton.edu/~rs/AlgsDS07/01UnionFind.pdf>

Disjoint Sets: Applications³¹

- Network connectivity: are two computers connected?
- Compilers: are two variables aliases?
- Image segmentation: are both pixels in same segment?
- Chip design: are two transistors connected to each other?
- Maze design
- Speeding up Kruskal's MST algorithm
- Many many more

³¹<http://courses.cs.washington.edu/courses/cse326/08sp/lectures/18-disjoint-union-find.pdf>

Disjoint Sets: Naming

- Represent each disjoint set by a unique name
- For convenience, the name is one of its elements
- This element is called the **leader** of the set
- $\text{Find}(x)$ returns the leader of set containing x
- Typically, Union takes leaders as input. For eg, $\text{Union}(a, b)$. If not easily fixable by $\text{Union}(\text{Find}(a), \text{Find}(b))$

Data Structures for Disjoint Sets

Objective:

- Design an efficient data structure to represent DS ADT
- Assume that there are N elements - represented by $1, \dots, N$
- M operations (any mixture of union and find)

Candidate Representations:

- Array based
- Linked List based
- Tree based

Disjoint Sets Implementation: Arrays

Idea:

- Maintain an array A with N elements
- $A[i]$ stores the leader for set containing element i

Find: Find(3)=4, Find(6)=8

i	1	2	3	4	5	6	7	8	9	10
A[i]	1	2	4	4	5	8	5	8	8	5

Union: Union(4,8)

i	1	2	3	4	5	6	7	8	9	10
A[i]	1	2	8	8	5	8	5	8	8	5

Disjoint Sets Implementation: Arrays

Analysis:

- Find:

Disjoint Sets Implementation: Arrays

Analysis:

- Find: $O(1)$
- Union:

Disjoint Sets Implementation: Arrays

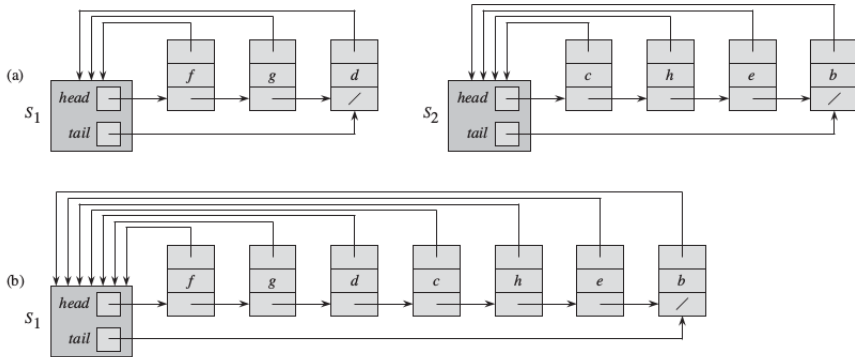
Analysis:

- Find: $O(1)$
- Union: $O(n)$
- Complexity for M operations: $O(MN)$

Disjoint Sets Implementation: Doubly Linked List³²

Idea:

- Represent each set as a doubly linked list
- Set first element of each linked list as the leader



- GiG

- GiG

Major Concepts:

- Binary Heap
- Heapsort
- Disjoint set data structures
- Union-Find