

CS146: Data Structures and Algorithms

Lecture 6



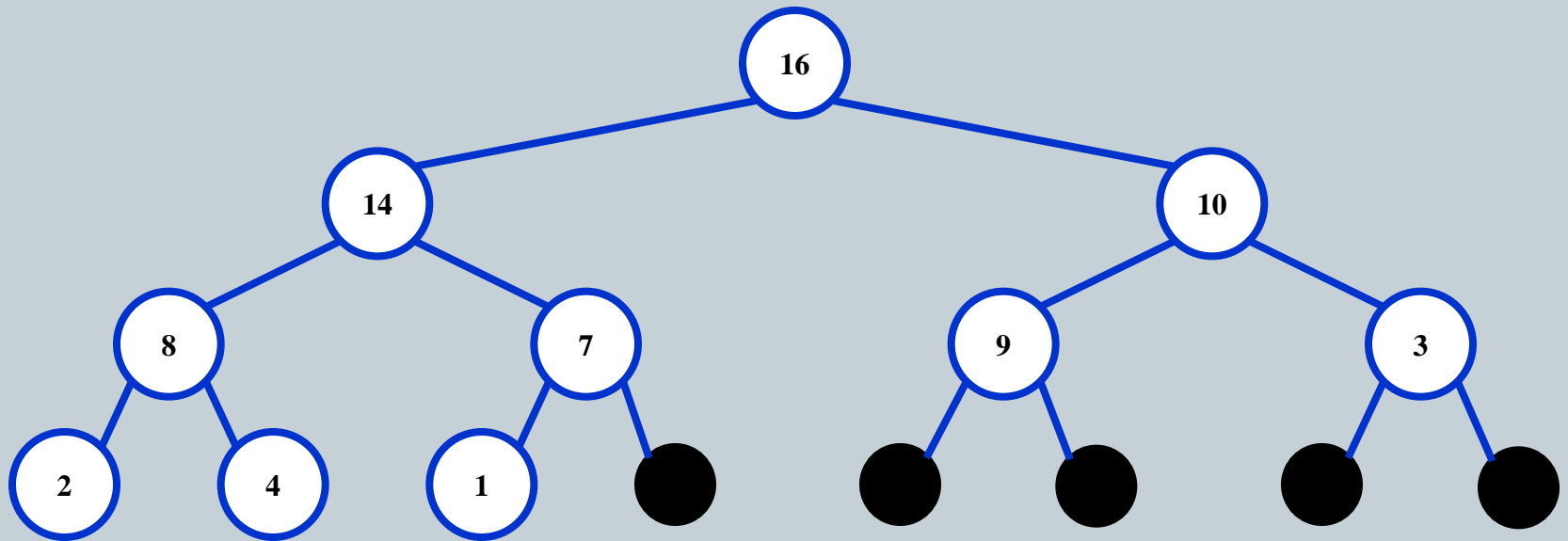
HEAPS - HEAPSORT

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CS SJSU

Heaps (chapter 6)

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A *heap* can be seen as a complete binary tree:



“nearly complete” binary trees: you can think of unfilled slots as null pointers

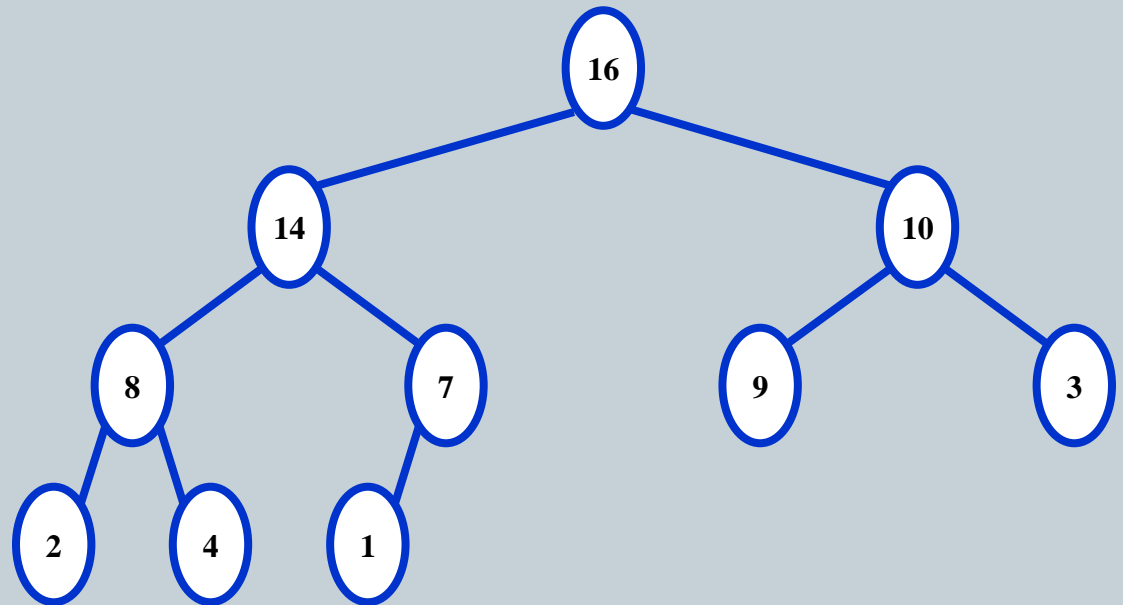
Heaps

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In practice, heaps are usually implemented as arrays:

<i>1</i>	<i>2</i>	<i>3</i>	<i>4</i>	<i>5</i>	<i>6</i>	<i>7</i>	<i>8</i>	<i>9</i>	<i>10</i>
16	14	10	8	7	9	3	2	4	1

$A =$



Heaps

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To represent a complete binary tree as an array:

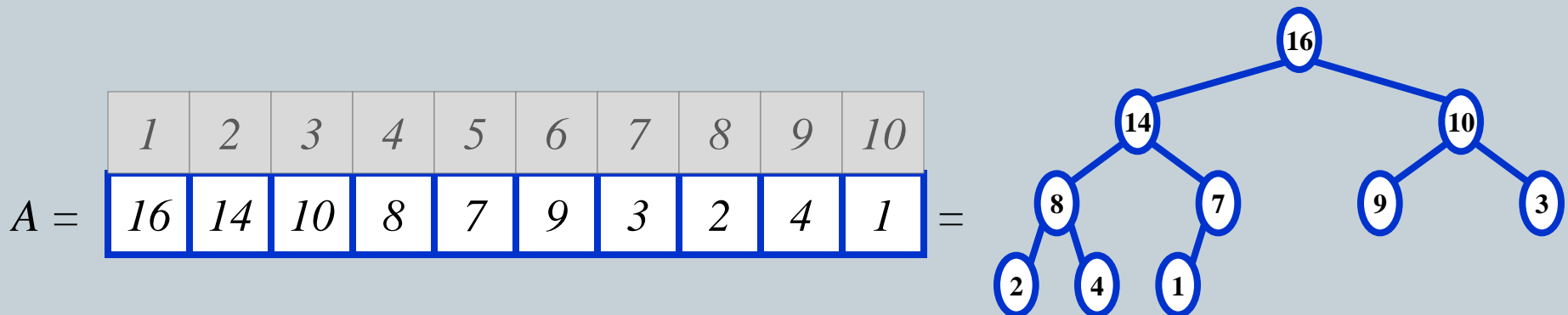
The root node is $A[1]$

Node i is $A[i]$

The parent of node i is $A[i/2]$ (note: integer divide)

The left child of node i is $A[2i]$

The right child of node i is $A[2i + 1]$



Referencing Heap Elements

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So...

```
parent(i) { return  $\lfloor i/2 \rfloor$ ; }  
left(i)   { return 2*i;   }  
right(i)  { return 2*i + 1; }
```

How would you implement this most efficiently?

The Heap Property

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Heaps also satisfy the *heap property*:

$$A[\textit{parent}(i)] \geq A[i] \quad \text{for all nodes } i > 1$$

i.e. , the value of a node is at most the value of its parent

In case of **Max Heap**

Where is the largest element in a heap stored?

- The *height* of a node in the tree = the number of edges on the longest downward path to a leaf
- The height of a tree = the height of its root

Self test

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- Is the array with values 23; 17; 14; 6; 13; 10; 1; 5; 7; 12 a max-heap?

Heap Height

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What is the height of an n -element heap?

This is nice property: all basic heap operations take at most time proportional to the height of the heap!

Heap Operations: Heapify()

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Heapify(): maintain the heap property

- *Input: a node i in the heap with children l and r*
- *& two subtrees rooted at l and r , assumed to be heaps*

The subtree rooted at i may violate the heap property (*Give an example...*)

- *Output: the tree rooted at i is a heap*

Action: let the value of the parent node “float down” so subtree at i satisfies the heap property

- *What basic operation between i , l , and r must be used?*

Heap Operations: Heapify()

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```
Heapify(A, i)
```

```
l = Left(i)
```

```
r = Right(i)
```

```
if (l <= A.heap_size && A[l] > A[i])
```

```
    largest = l
```

```
else largest = i
```

```
if (r <= A.heap_size && A[r] > A[largest])
```

```
    largest = r
```

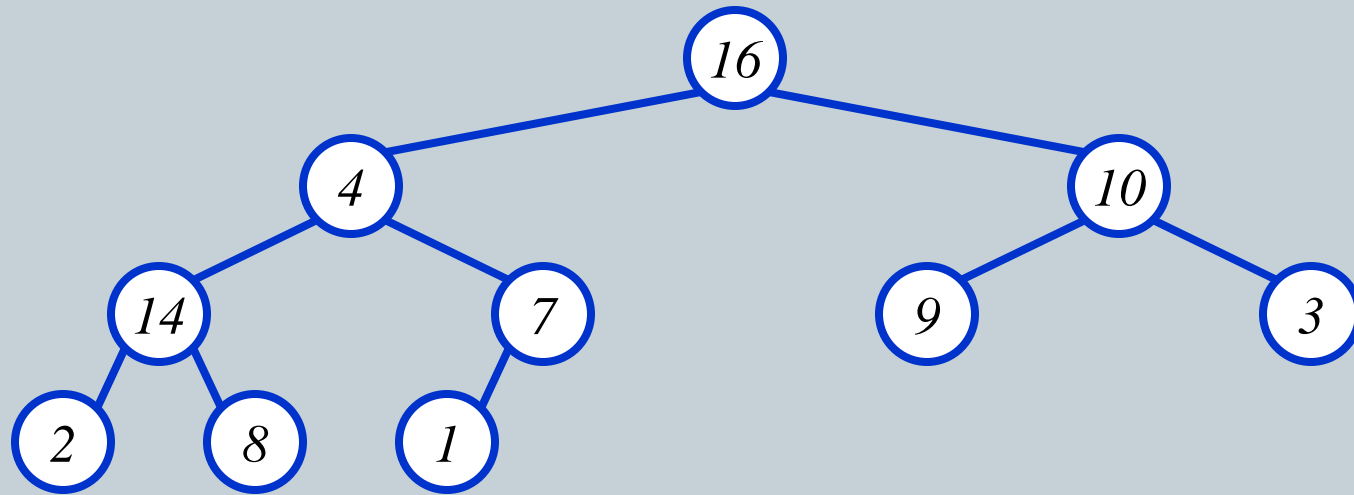
```
if (largest != i)
```

```
    Swap(A, i, largest);
```

```
    Heapify(A, largest);
```

Heapify() Example

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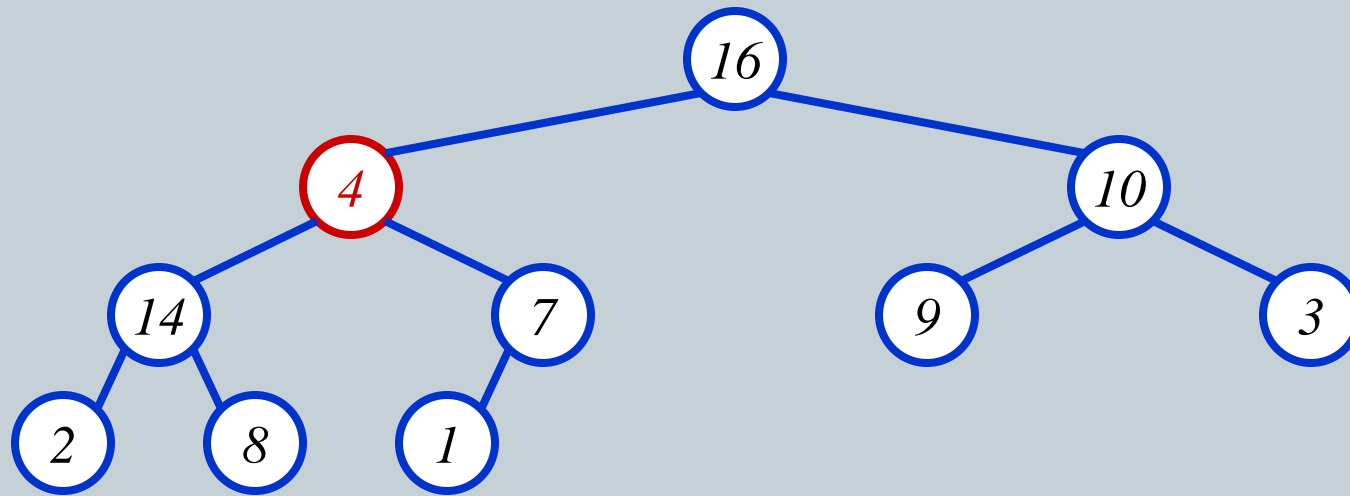


$A =$

1	2	3	4	5	6	7	8	9	10
16	4	10	14	7	9	3	2	8	1

Heapify() Example

12

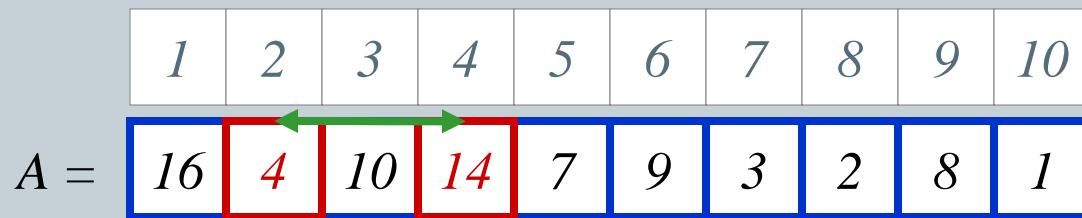
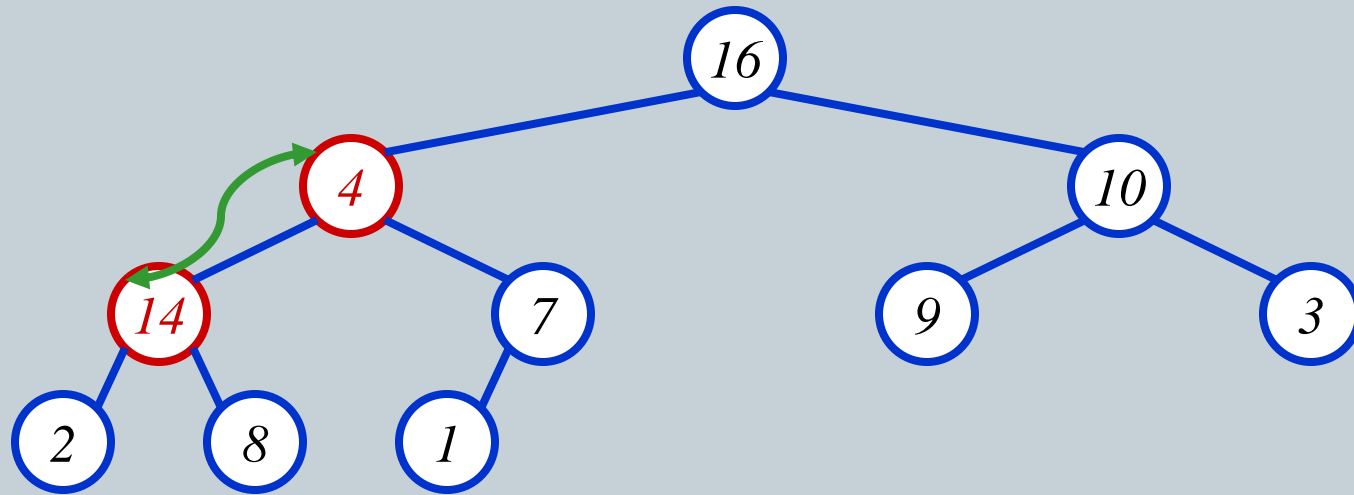


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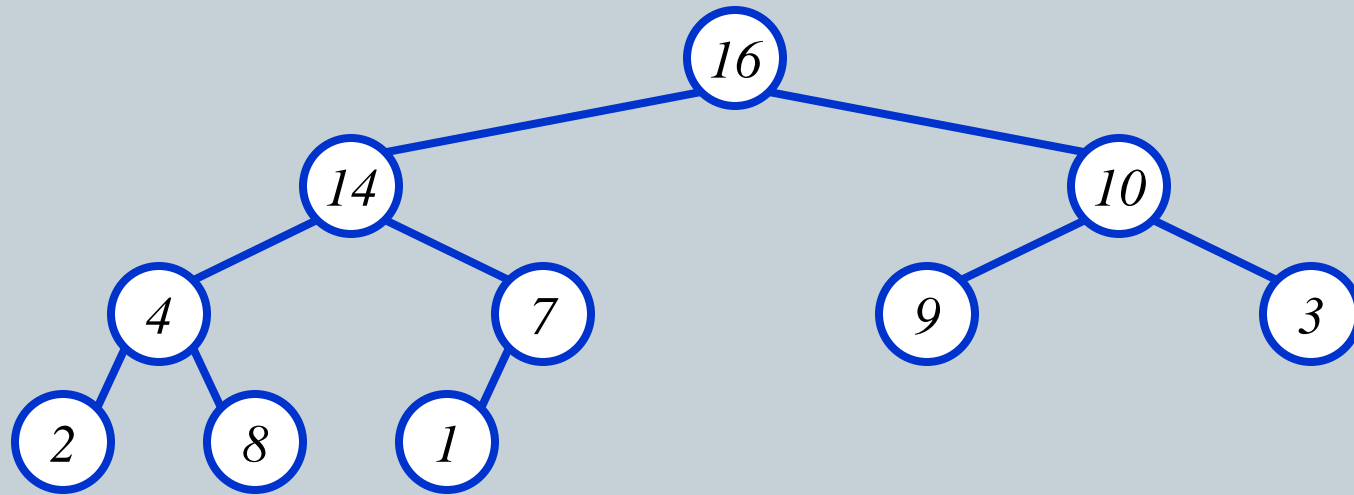
Heapify() Example

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Heapify() Example

14

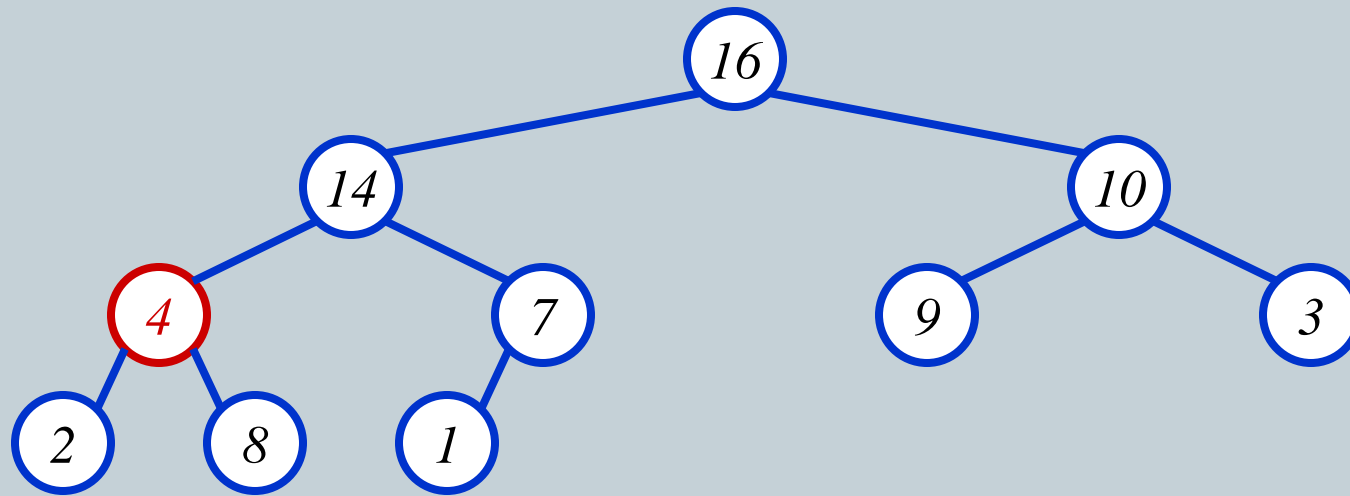


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Heapify() Example

15

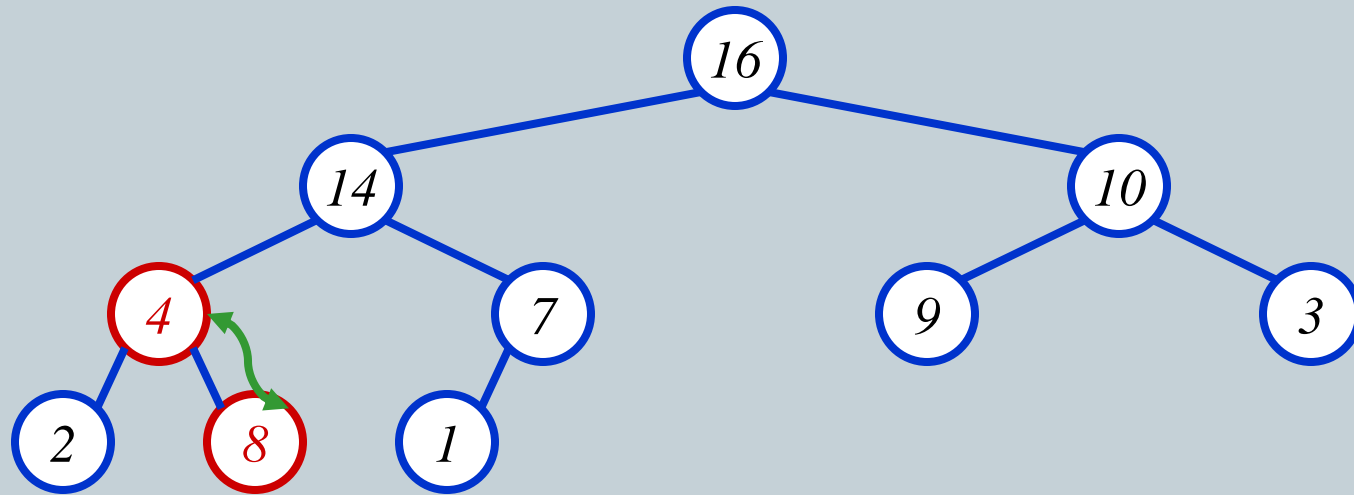


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Heapify() Example

16

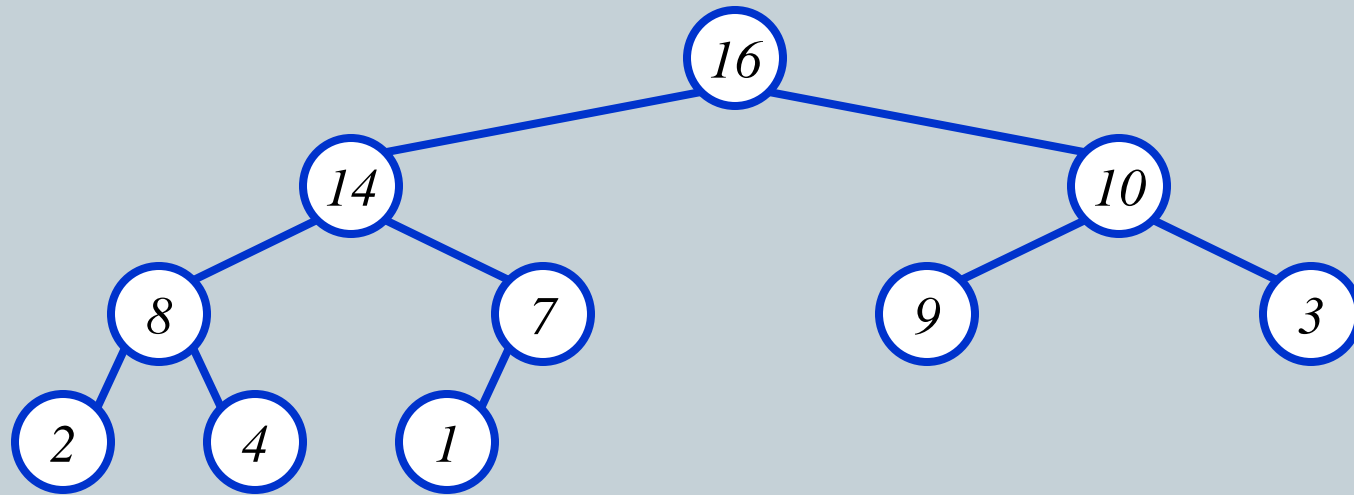


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Heapify() Example

17

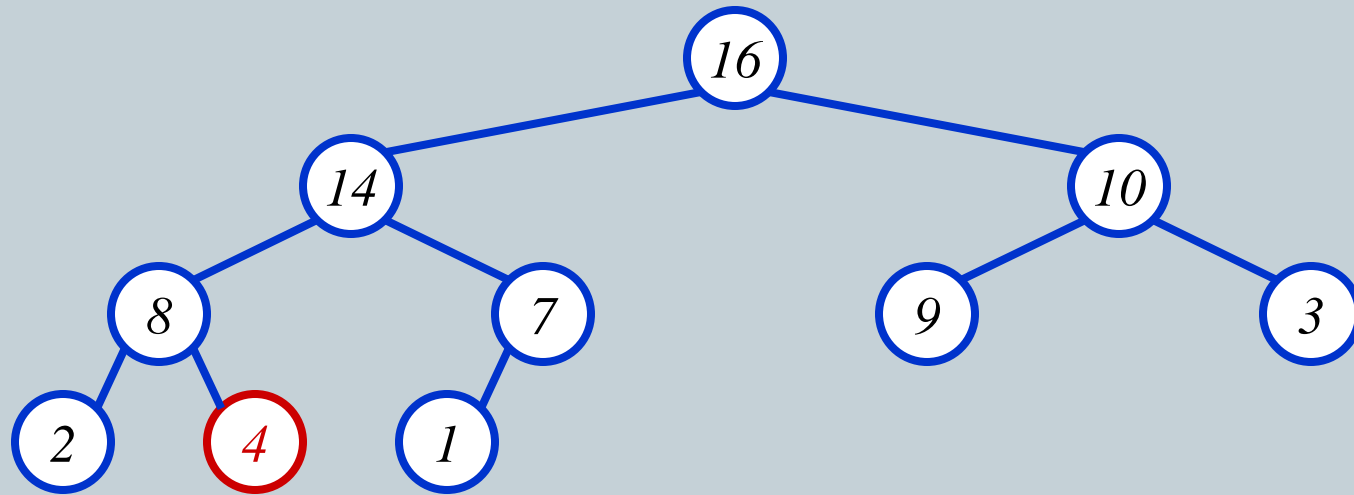


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Heapify() Example

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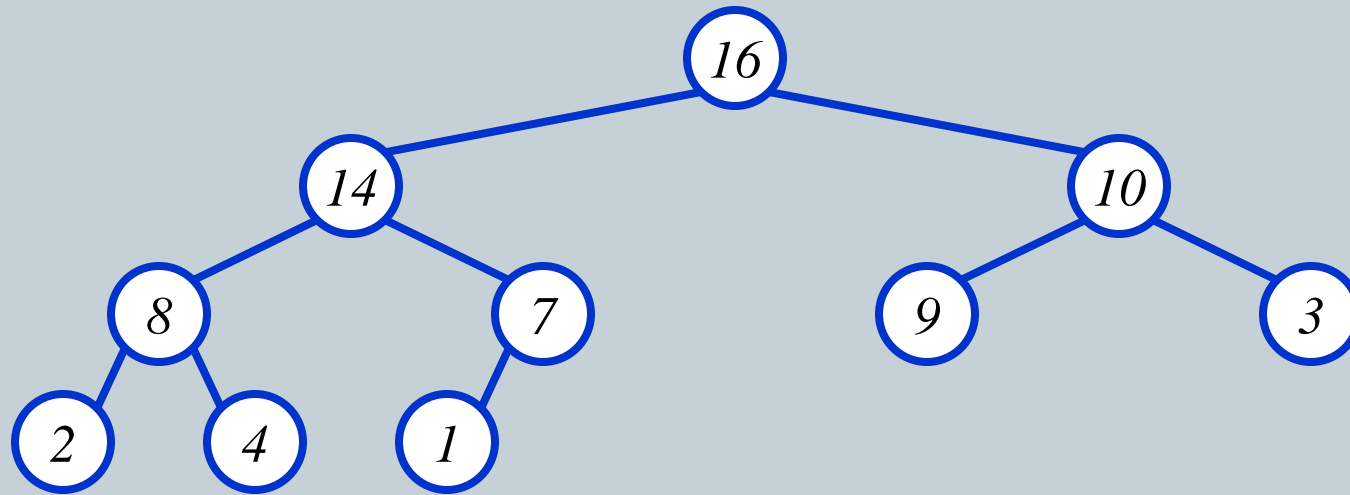


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16	14	10	8	7	9	3	2	4	1

Heapify() Example

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$A =$

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16	14	10	8	7	9	3	2	4	1

Analyzing Heapify()

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*Except the recursive call, what is the running time of **Heapify()**?*

*How many times can **Heapify()** recursively call itself?*

*What is the worst-case running time of **Heapify()** on a heap of size n ?*

Analyzing Heapify(): Formal

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Fixing up relationships between i , l , and r takes $\Theta(1)$ time

If the heap at i has n elements, how many elements can the subtrees at l or r have?

Draw it

Answer: $2n/3$ (worst case: bottom row 1/2 full)

So time taken by **Heapify()** is given by

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

Analyzing Heapify(): Formal

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So we have

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

By case 2 of the Master Theorem,

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

Thus, **Heapify()** takes logarithmic time

Heap Operations: BuildHeap()

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We can build a heap in a *bottom-up* manner by running **Heapify()** on successive subarrays




Fact: for array of length n , all elements in range $A[\lfloor n/2 \rfloor + 1 .. n]$ are heaps (*Why is that?*)

Key idea:

- Walk backwards through the array from $n/2$ to 1, calling **Heapify()** on each node.
- Order of processing guarantees that the children of node i are heaps when i is processed

BuildHeap()

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```
// given an unsorted array A, make A a heap  
BuildHeap(A)  
{  
    A.heap_size = A.length;   
    for (i = ⌊A.length/2⌋ downto 1)   
        Heapify(A, i);   
}
```

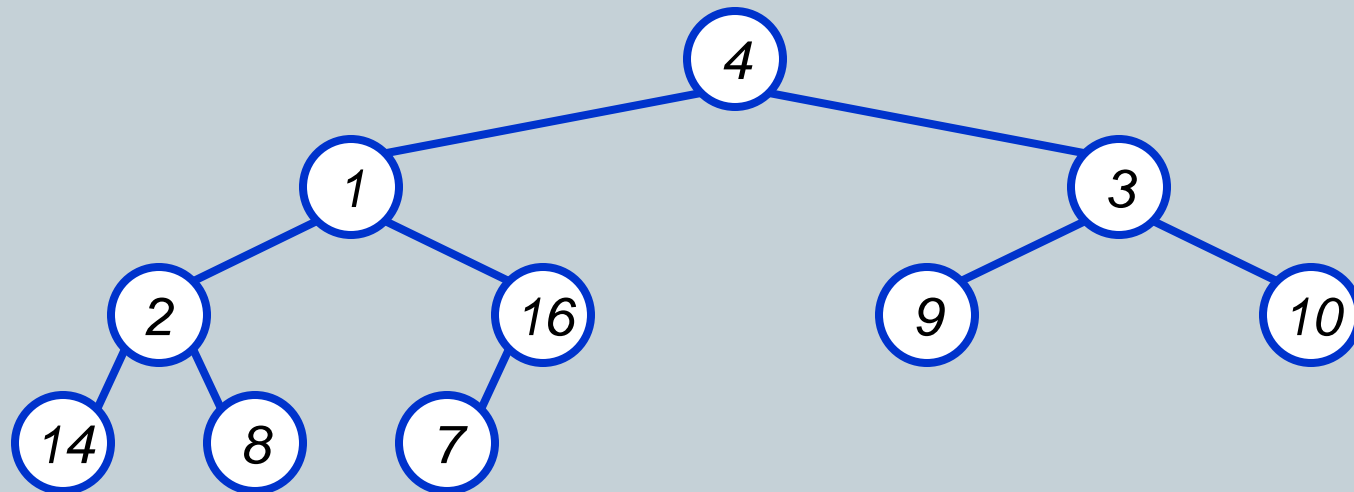


BuildHeap() Example

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Work through example

$A = \{4, 1, 3, 2, 16, 9, 10, 14, 8, 7\}$



Analyzing BuildHeap()

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Each call to **Heapify()** takes $O(\log n)$ time

There are $O(n)$ such calls ($\lfloor n/2 \rfloor$)

Thus the running time is $O(n \log n)$

Is this a correct asymptotic upper bound?

Is this an asymptotically tight bound?

Can we do better?

A tighter bound is $O(n)$

Is there a flaw in the above reasoning? No more careful analysis

Analyzing BuildHeap(): Tight

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To **Heapify** () a subtree takes $O(h)$ time where h is the height of the subtree

$h = O(\lg m)$, $m = \#$ nodes in subtree

The height of most subtrees is small

Fact: an n -element heap has at most $\lceil n/2^{h+1} \rceil$ nodes of height h

Uses this fact to prove that **BuildHeap** () takes $O(n)$ time

BuildHeap() is O(n)!

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

- Due $X=1/2$ it is

$$\begin{aligned} \sum_{h=0}^{\infty} \frac{h}{2^h} &= \frac{1/2}{(1 - 1/2)^2} \\ &= 2 . \end{aligned}$$

$$\begin{aligned} O \left(n \sum_{h=0}^{\lfloor \lg n \rfloor} \frac{h}{2^h} \right) &= O \left(n \sum_{h=0}^{\infty} \frac{h}{2^h} \right) \\ &= O(n) . \end{aligned}$$

Self Test

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- illustrate the operation of BUILD-MAX-HEAP on the array 5; 3; 17; 10; 84; 19; 6; 22; 9.

Heapsort!

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Given **BuildHeap()**, an in-place sorting algorithm is easily constructed, key idea:

- Maximum element is at $A[1]$
- Discard by swapping with element at $A[n]$
 - Decrement $\text{heap_size}[A]$
 - $A[n]$ now contains correct value
- Restore heap property at $A[1]$ by calling **Heapify()**
- Repeat, always swapping $A[1]$ for $A[\text{heap_size}(A)]$

Heapsort

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Heapsort (A)

BuildHeap (A) ;

for (i = A.length downto 2)

Swap (A[1], A[i]) ;

A.heap_size (A) -= 1 ;

Heapify (A, 1) ;

Example

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- 21, 14, 16, 12, 10, 4, 8
- HeapSort using min heap

Analyzing Heapsort

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The call to **BuildHeap()** takes $O(n)$ time

Each of the $n - 1$ calls to **Heapify()** takes $O(\log n)$ time

Thus the total time taken by **HeapSort()**

$$= O(n) + (n - 1) O(\lg n)$$

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

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

Priority Queues

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Heapsort is a nice algorithm, but in practice Quicksort (coming up) usually wins

But the heap data structure is incredibly useful for implementing *priority queues*

A data structure for maintaining a set S of elements, each with an associated value or *key*

Supports the operations **Insert()**, **Maximum()**, and **ExtractMax()**

What might a priority queue be useful for?

Priority Queue Operations

35

Insert(S, x) inserts the element x into set S

Maximum(S) returns the element of S with the maximum key

ExtractMax(S) removes and returns the element of S with the maximum key

How could we implement these operations using a heap?

Priority Queues

36

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Priority Queue Operations

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How could we implement these operations using a heap?

Maximum and ExtractMaximum

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```
HeapMaximum(A)
```

```
    Return A[1]
```

```
HeapExtractMax(A)
```

```
{
```

```
    if (A.heap_size < 1) { error }
```

```
    max = A[1]
```

```
    A[1] = A[A.heap_size]
```

```
    A.heap_size --
```

```
    Heapify(A, 1)
```

```
    return max
```

```
}
```

HeapExtractMax

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$O(\lg n)$ (same as Heapify)

IncreaseKey

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IncreaseKey(A,i,key)

{

if key < A[i] error “new key is smaller than current”

A[i]=key

while (i > 1 && A[parent(i)] < A[i])

swap(A[i],A[parent(i)])

i=parent(i)

}

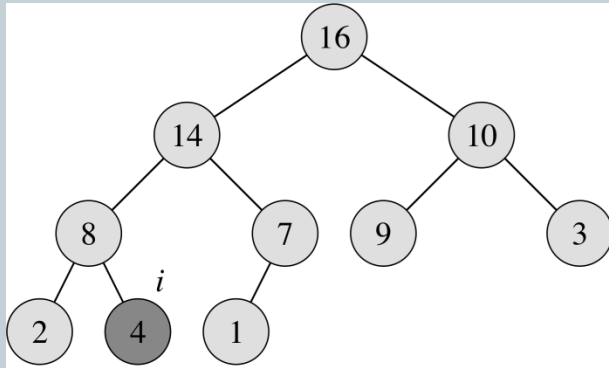
IncreaseKey complexity

41

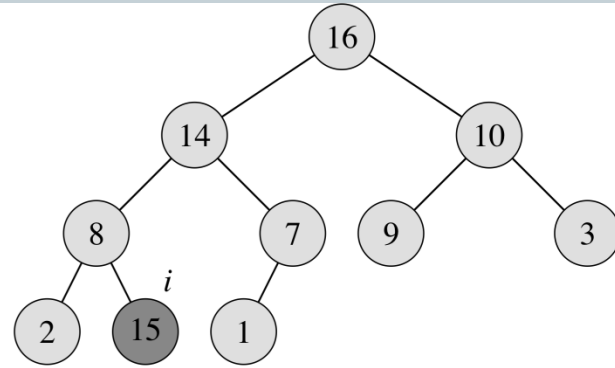
$O(\lg n)$ why?

example

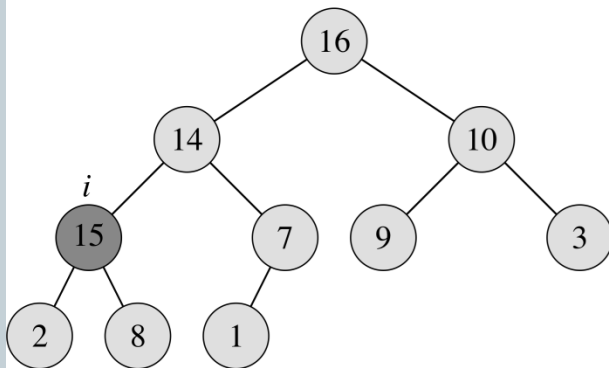
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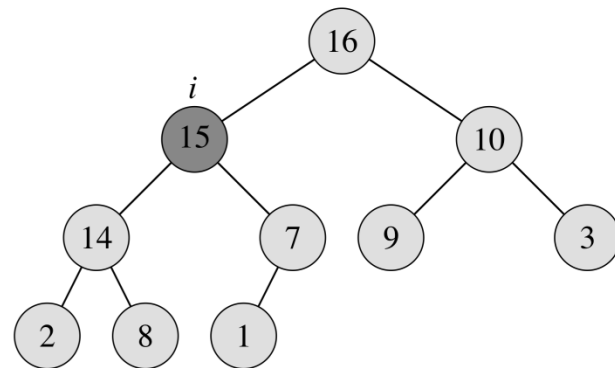
(a)



(b)



(c)



(d)

Insert

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```
Insert (A, key)
{
    A.heap_size++
    A[A.heap_size] = -infinity
    IncreaseKey (A, A.heap_size, key)
}
```

$O(\lg n)$ complexity