

# Lecture 12

## Priority Queue

OOP

Linked Lists

Arrays

Queues

Stacks

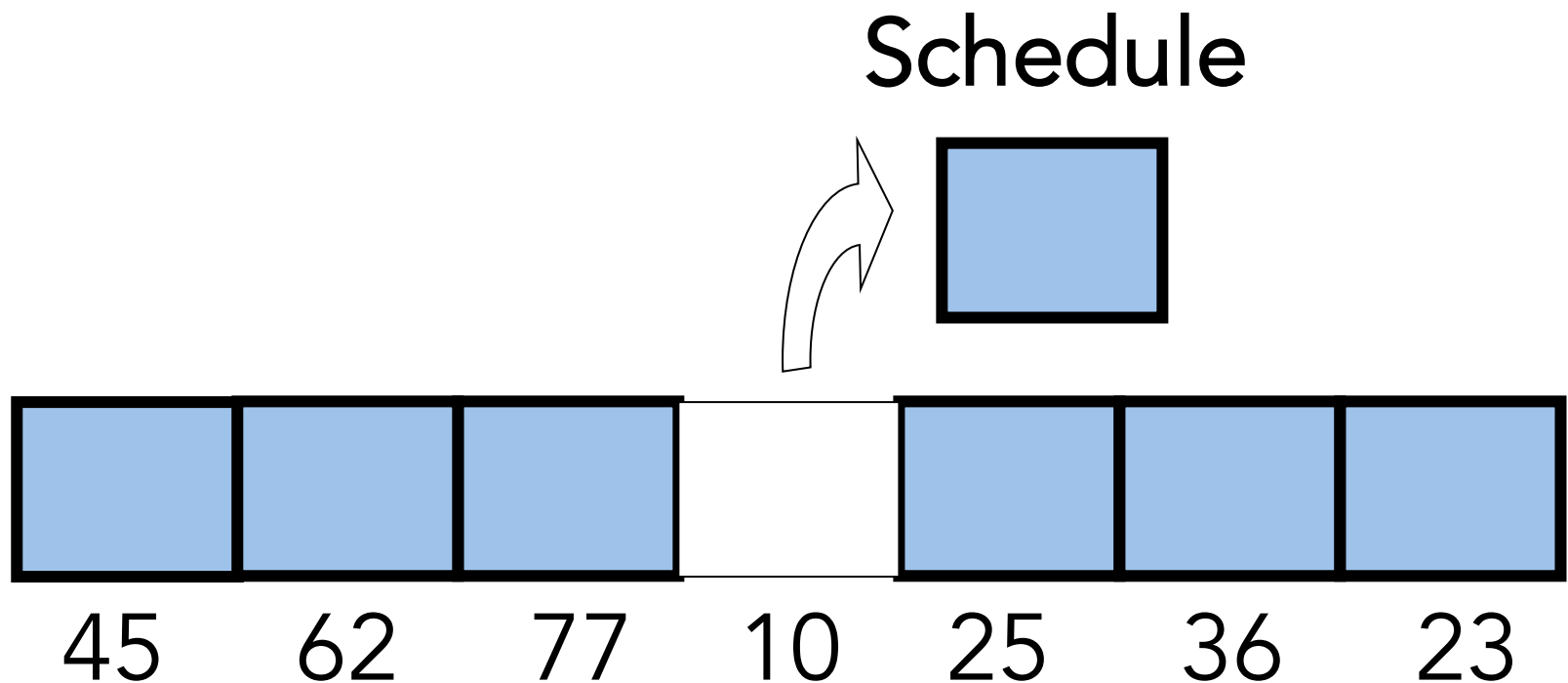
Trees

Algorithm Analysis

# Scheduler

- ready programs are added to the scheduler
- it decides which program to execute next

# SRTF Scheduling Policy



# Main Operations

insert()

removeMin()

# Priority Queue ADT

- a collection of entries
- `entry` = (key, value)
- main methods
  - `insert(k, v)`
  - `removeMin()`
- additional methods
  - `min()`
  - `size()`,
  - `isEmpty()`

# Example

Method	Return Value	Priority Queue Contents
insert(5,A)		{ (5,A) }
insert(9,C)		{ (5,A), (9,C) }
insert(3,B)		{ (3,B), (5,A), (9,C) }
min()	(3,B)	{ (3,B), (5,A), (9,C) }
removeMin()	(3,B)	{ (5,A), (9,C) }
insert(7,D)		{ (5,A), (7,D), (9,C) }
removeMin()	(5,A)	{ (7,D), (9,C) }
removeMin()	(7,D)	{ (9,C) }
removeMin()	(9,C)	{ }
removeMin()	null	{ }
isEmpty()	true	{ }

# Keys

- specific attribute of the element
- many times assigned to the element by user of application
- key may change for the same element, e.g. popularity



# Total Order Relations

1. Comparability property:

either  $x \leq y$  or  $y \leq x$

2. Antisymmetric property:

$x \leq y$  and  $y \leq x \Rightarrow x = y$

3. Transitive property:

$x \leq y$  and  $y \leq z \Rightarrow x \leq z$

- keys with total order
  - weight
- keys not having total order
  - 2D point

# Comparator ADT

- implements `isLess(p,q)`
- can derive other relations from this:
  - $-(p == q)?$
- for STL, in C++ overload `"()`

# Comparator Examples

```
class LeftRight {  
public:  
    bool operator()(Point2D& p, Point2D& q)  
    { return p.getX() < q.getX(); }  
};
```

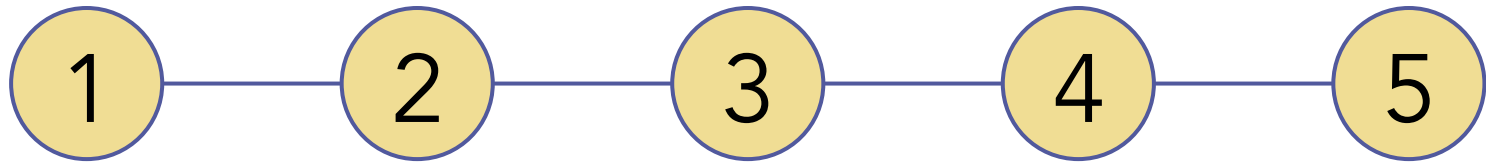
```
class BottomTop {  
public:  
    bool operator()( Point2D& p, Point2D& q)  
    { return p.getY() < q.getY(); }  
};
```

# Sort Sequence L with Priority Queue P

```
while !L.empty()  
    e ← L.front();  
    L.eraseFront()  
    P.insert (e)  
while !P.empty()  
    e ← P.removeMin()  
    L.insertBack(e)
```

What is the time  
complexity of this  
sorting?

# Implementation with sorted sequence



- insert()?
- removeMin()?

# Insertion-Sort

1. insert at right place to keep the list sorted
2. remove head repeatedly



# Insertion-Sort Example

	Sequence S	Priority queue P
Input:	(7,4,8,2,5,3,9)	()

## Phase 1

(a)	(4,8,2,5,3,9)	(7)
(b)	(8,2,5,3,9)	(4,7)
(c)	(2,5,3,9)	(4,7,8)
(d)	(5,3,9)	(2,4,7,8)
(e)	(3,9)	(2,4,5,7,8)
(f)	(9)	(2,3,4,5,7,8)
(g)	()	(2,3,4,5,7,8,9)

## Phase 2

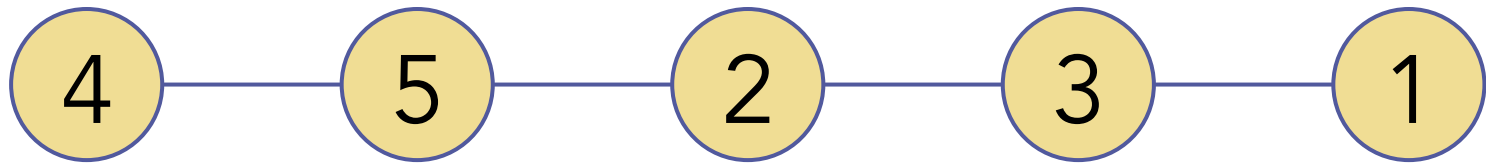
(a)	(2)	(3,4,5,7,8,9)
(b)	(2,3)	(4,5,7,8,9)
..	..	..
(g)	(2,3,4,5,7,8,9)	()

# Time complexity

## Best case?

## Worst case?

# Implementation with unsorted sequence



- insert()?
- removeMin()?

# Selection-Sort

- 1.insert elements in  
unsorted list
- 2.remove min repeatedly

# Selection-Sort Example

	Sequence L	Priority Queue P
Input:	(7,4,8,2,5,3,9)	()

Phase 1

(a)	(4,8,2,5,3,9)	(7)
(b)	(8,2,5,3,9)	(7,4)
..	.. ..	
(g)	()	(7,4,8,2,5,3,9)

Phase 2

(a)	(2)	(7,4,8,5,3,9)
(b)	(2,3)	(7,4,8,5,9)
(c)	(2,3,4)	(7,8,5,9)
(d)	(2,3,4,5)	(7,8,9)
(e)	(2,3,4,5,7)	(8,9)
(f)	(2,3,4,5,7,8)	(9)
(g)	(2,3,4,5,7,8,9)	()

What is time complexity  
of selection sort?

Worst case?

Best case?

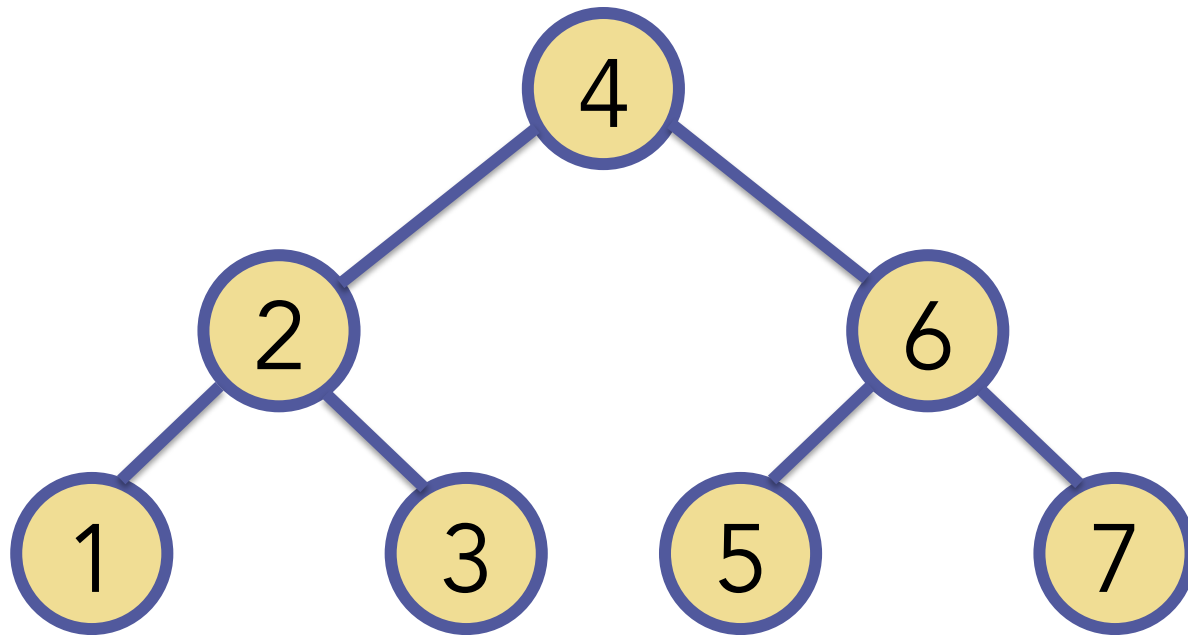
Give two advantages of selection sort over others?

Summary: either  
insert is  $O(n)$  or  
removeMin() is  $O(n)$ !



Can we do better?

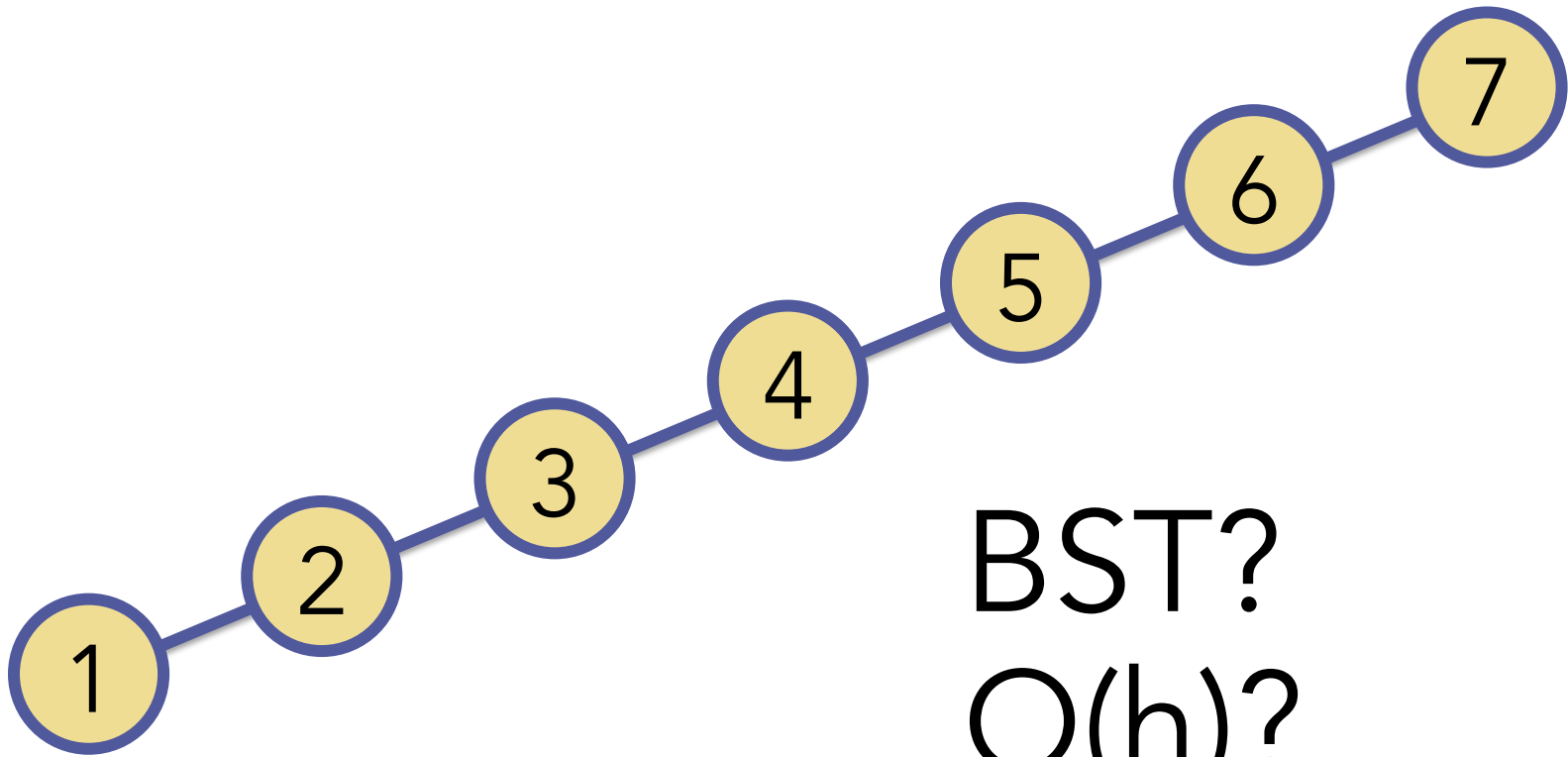
# Lets Try BST



Time complexity  
removeMin()?  
insert()?

Is  $O(h)$  good  
enough?

# Look at this tree...



BST?  
 $O(h)$ ?

# BST Order Restrictions

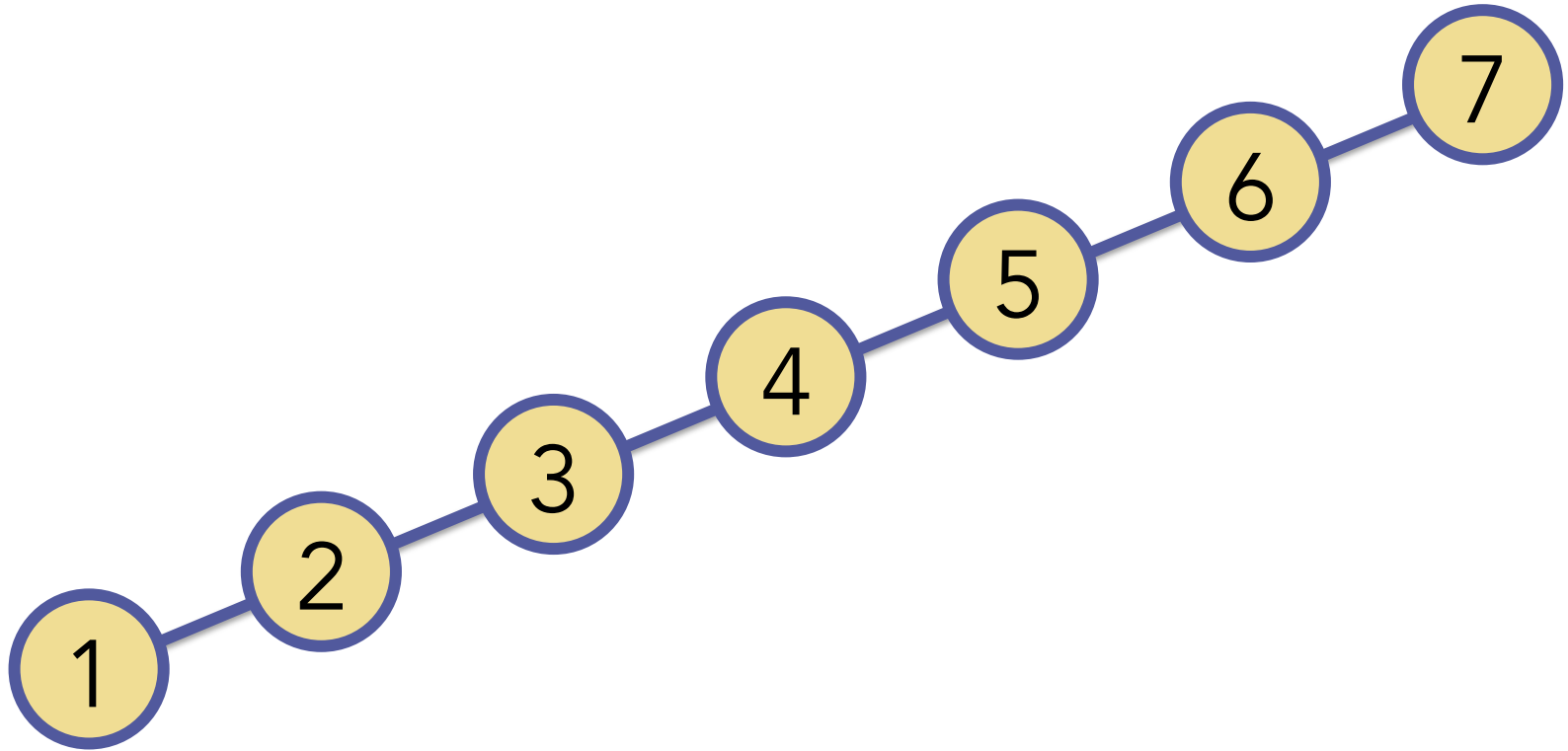
1. left subtree  $\leq$  node

2. right subtree  $\geq$  node

# A simpler order restriction...

1.  $\text{parent} \leq \text{child}$

# Remake this tree with new order restriction...





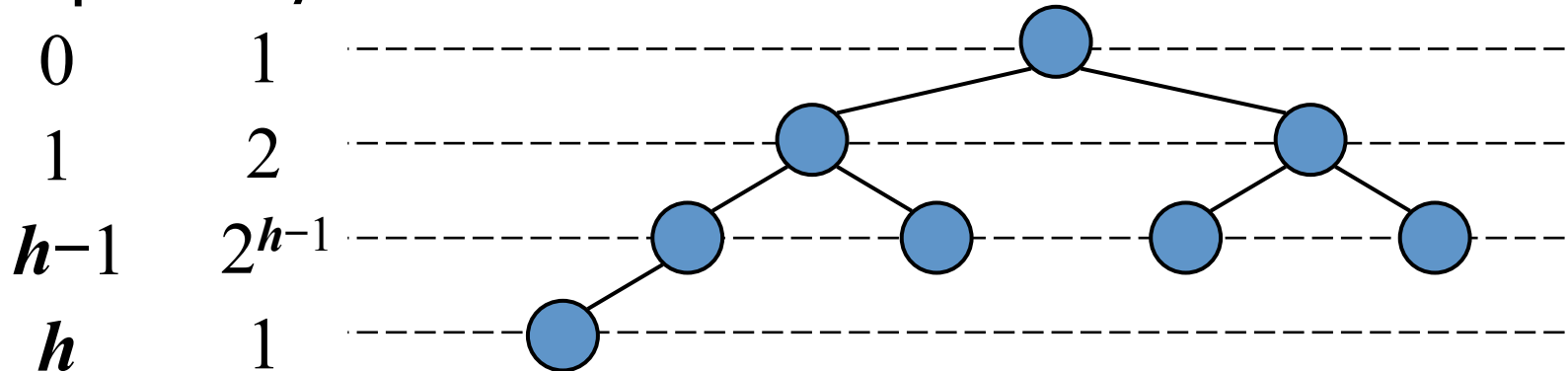
# Heap

1. Order restriction:  
 $K(\text{parent}) \leq K(\text{child})$
2. Structural restriction:  
complete binary

# Height of Heap

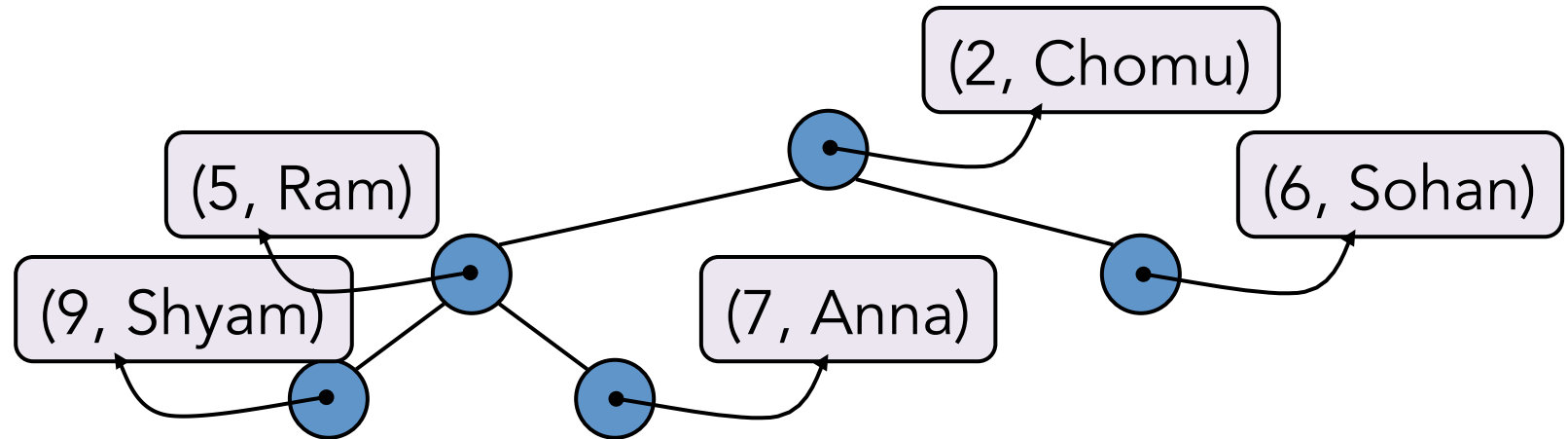
$h \leq \log n$ , so,  $h$  is  $O(\log n)$

depth keys

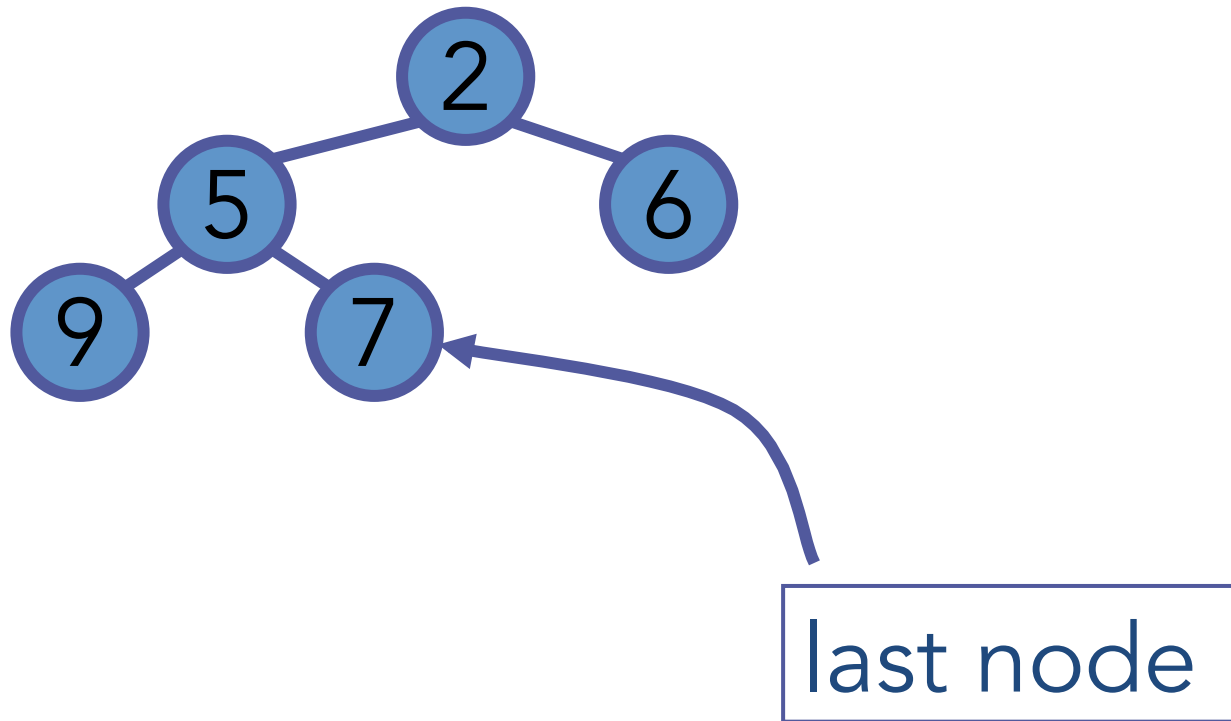


# Heaps and Priority Queues

- Heap can be used to implement a priority queue
- store a (key, element) item at each internal node
- keep track of the last node



The **last node** of a heap is  
the rightmost node of  
maximum depth!

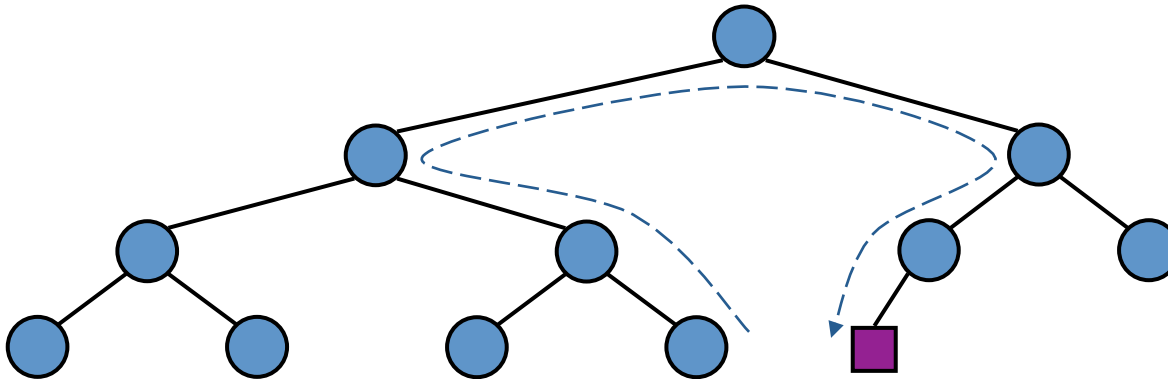


# insert()

- 1.find insertion node
- 2.add new element (k)  
at this node
- 3.restore heap order

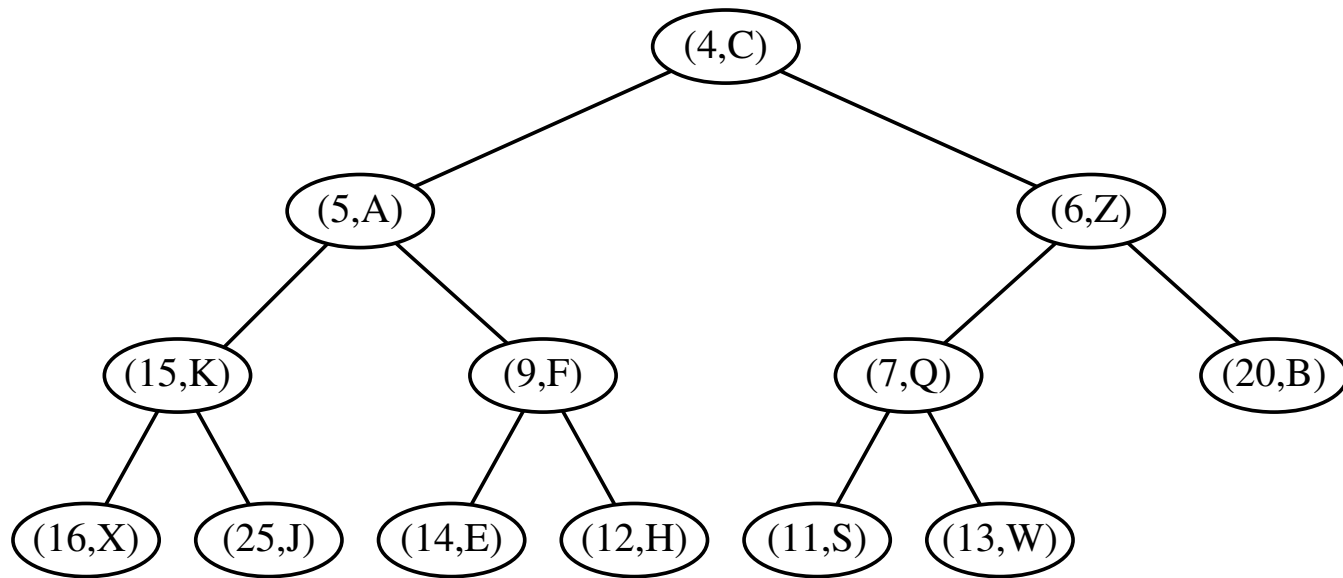
# Finding the Insertion Node

- go up until node becomes a left child or the root is reached
- if root is reached, go left until leaf node is reached
- if node becomes left child, go to the right sibling and go down left until a leaf is reached



time complexity of  
finding the insertion  
node

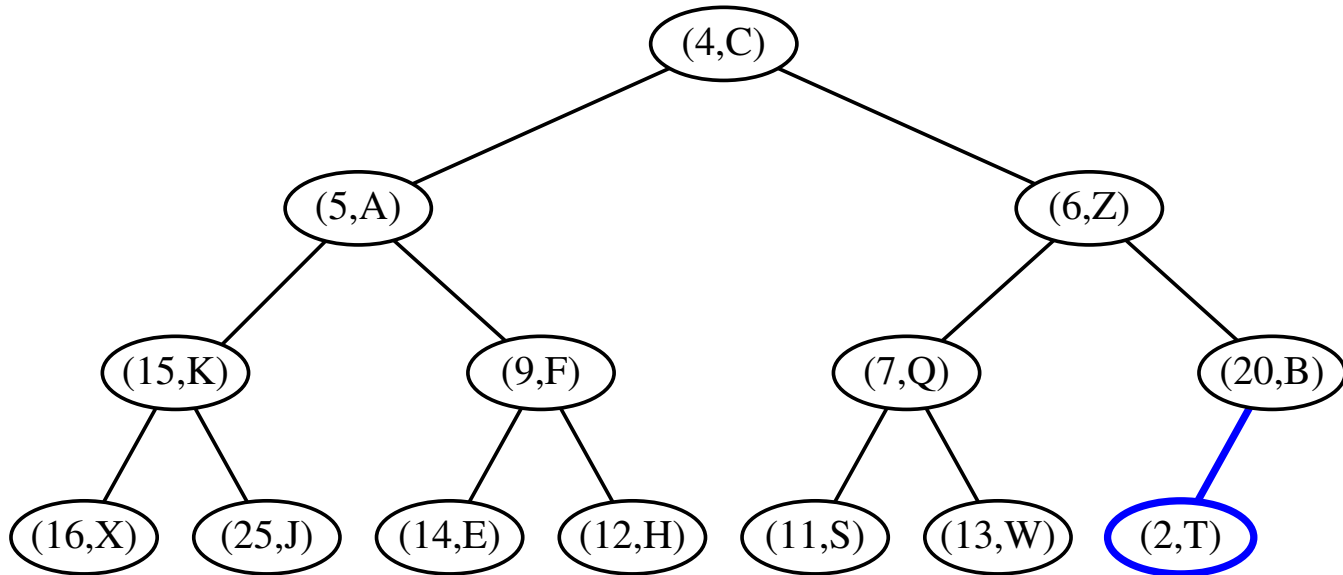
# Insertion into a Heap



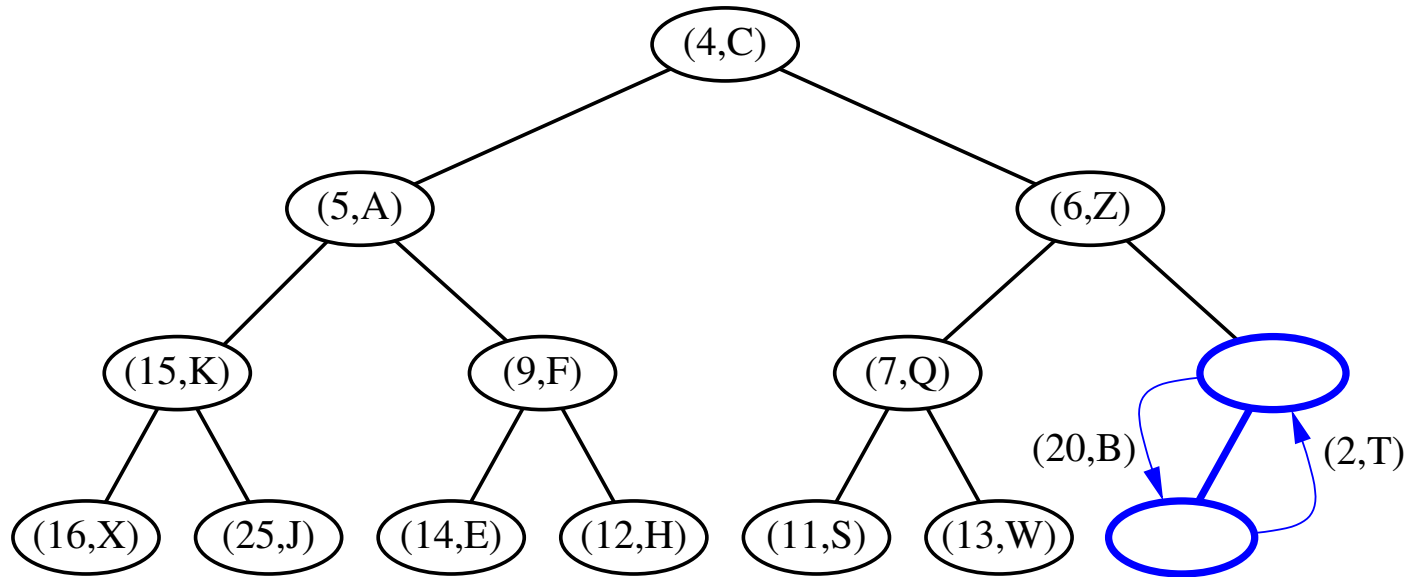
(2, T)



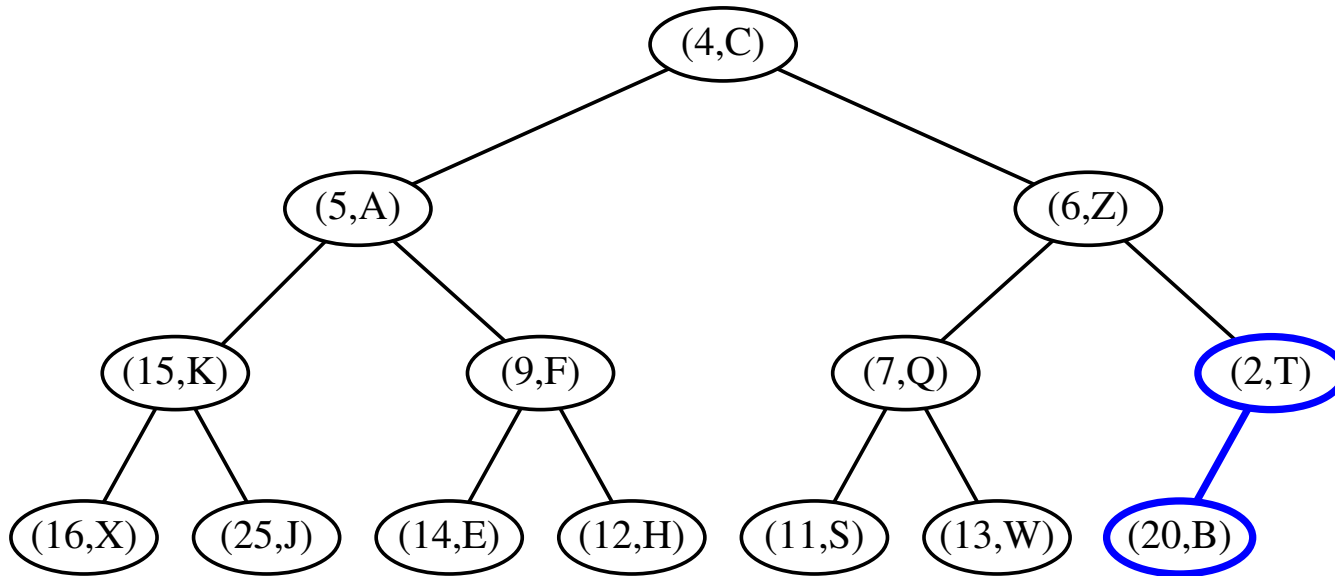
# up-heap bubbling



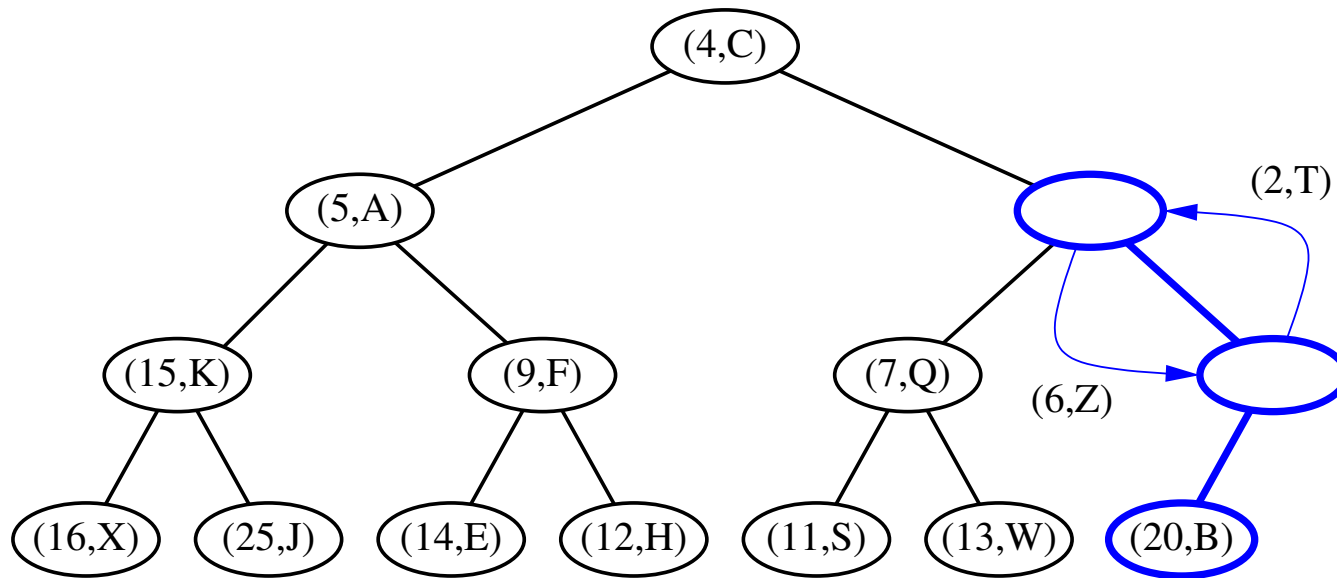
# up-heap bubbling



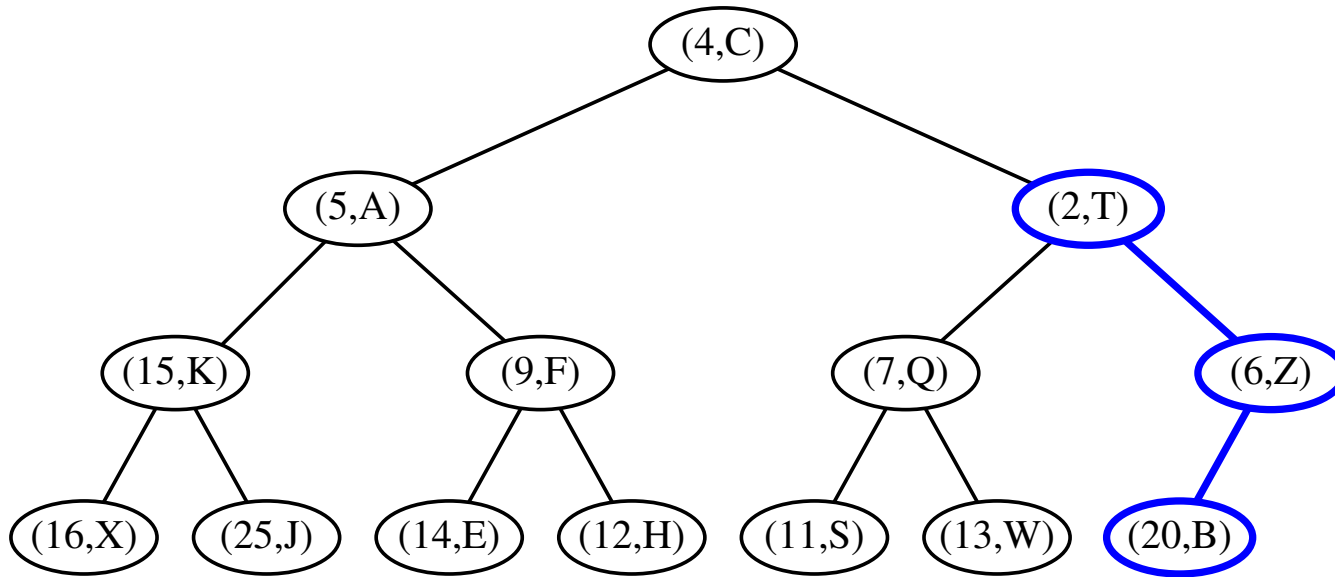
# up-heap bubbling



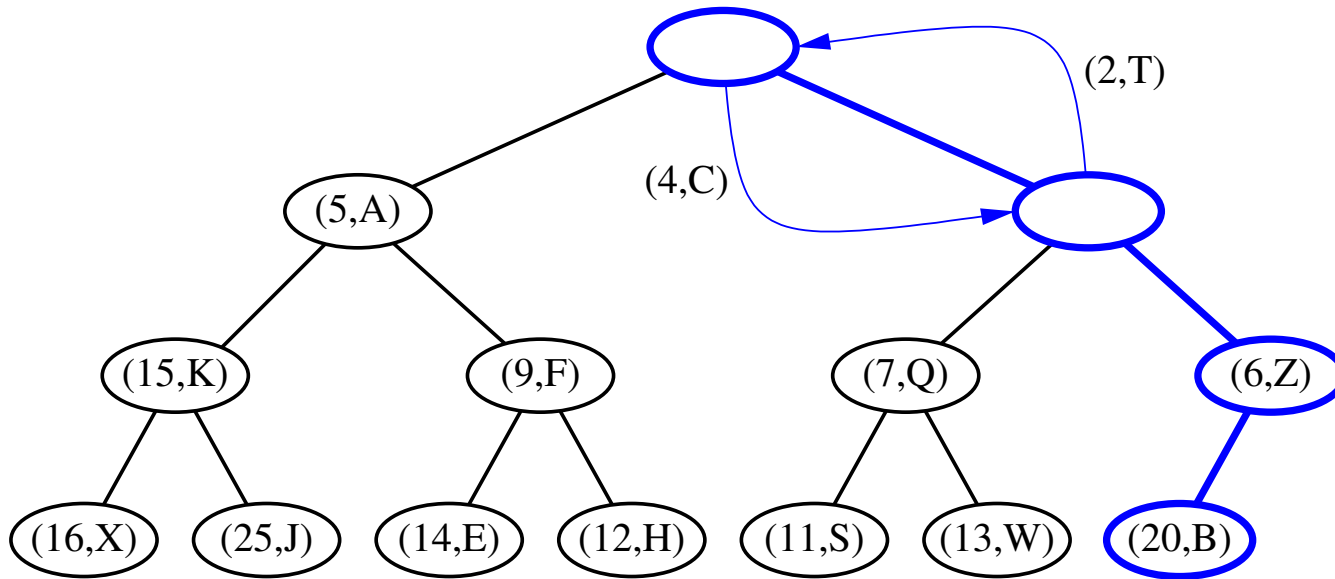
# up-heap bubbling



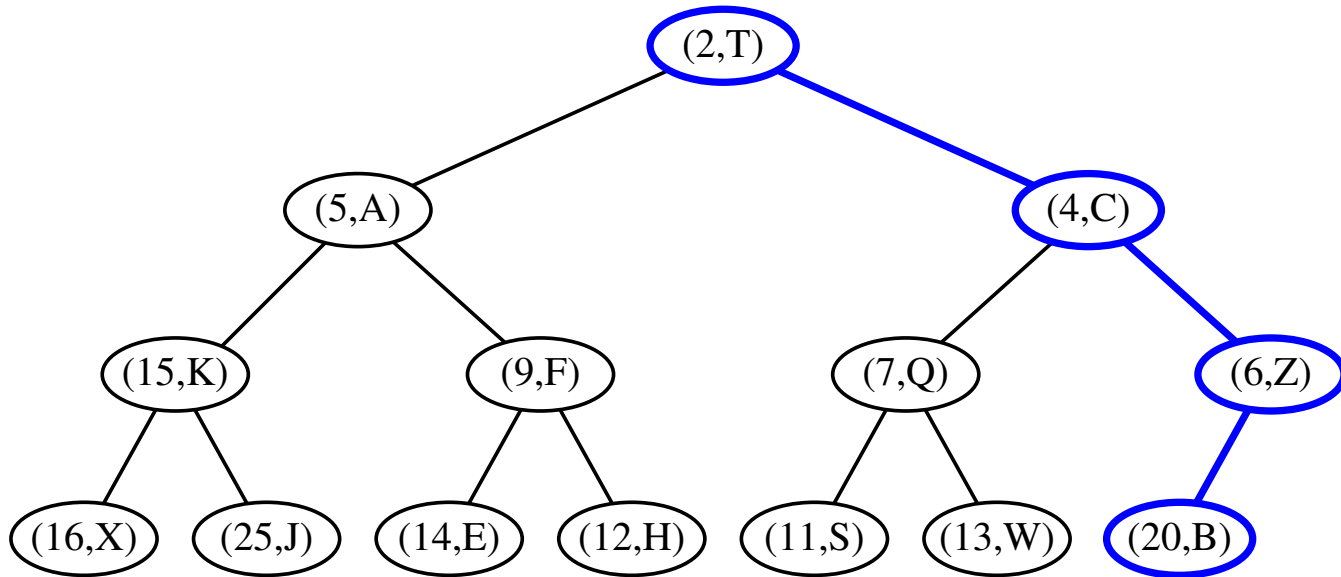
# up-heap bubbling



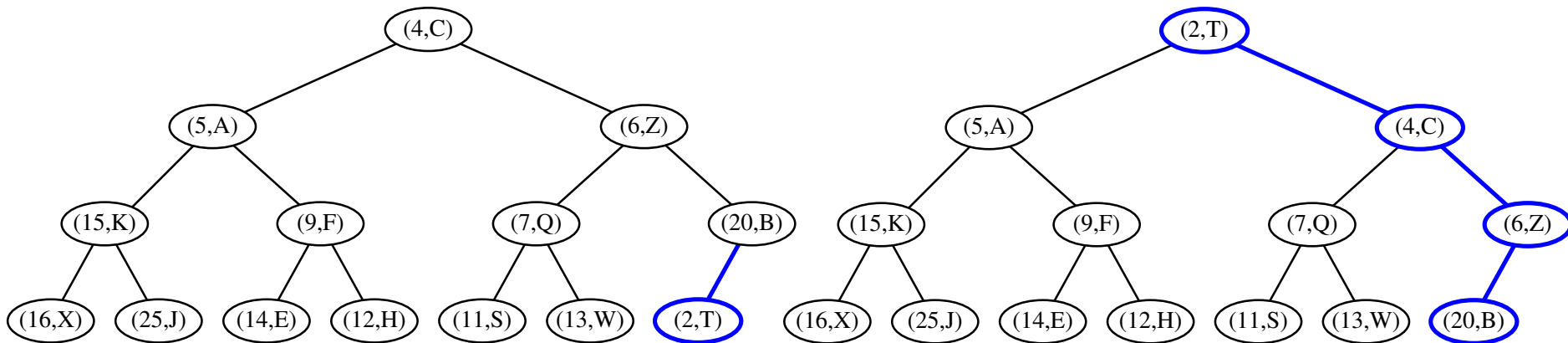
# up-heap bubbling



# up-heap bubbling

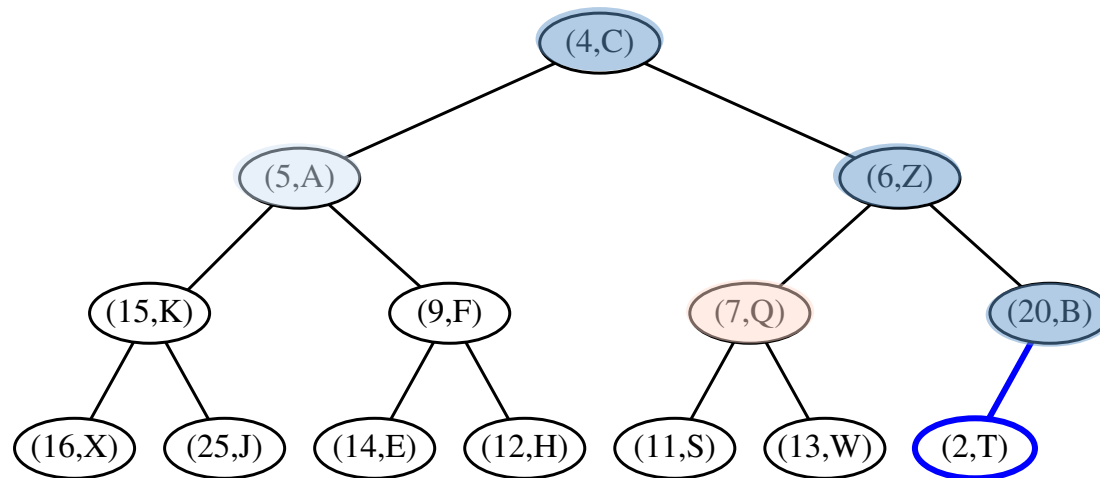


# Another View of Insertion

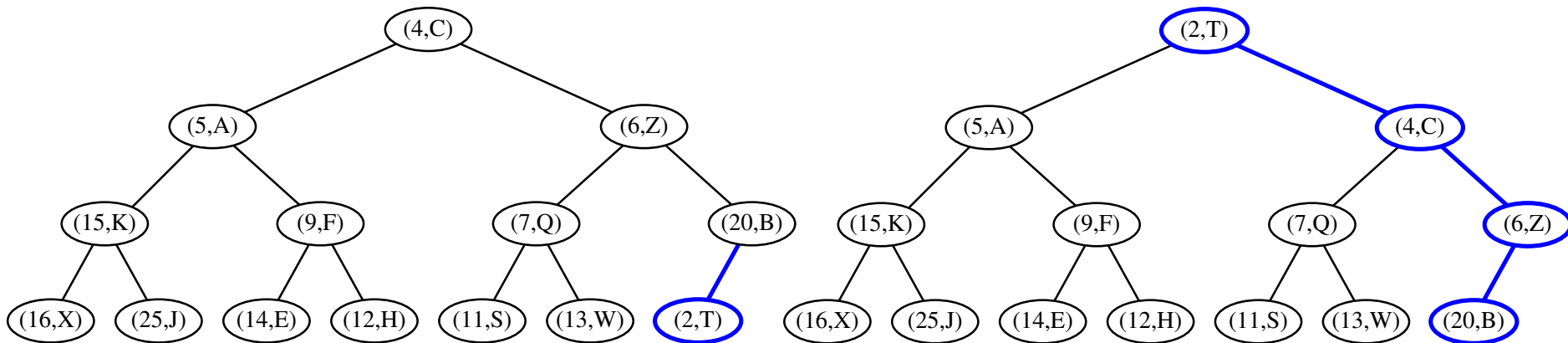




# Correctness of Upheap



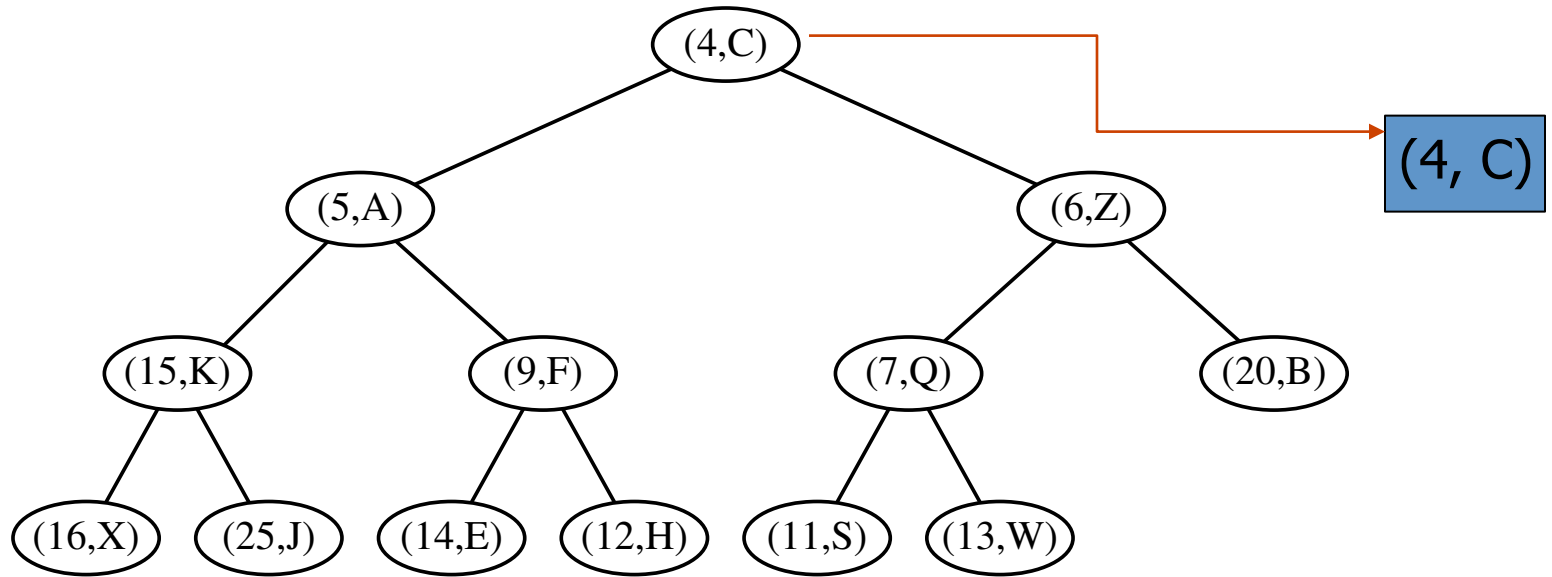
# Always replaced with smaller key!



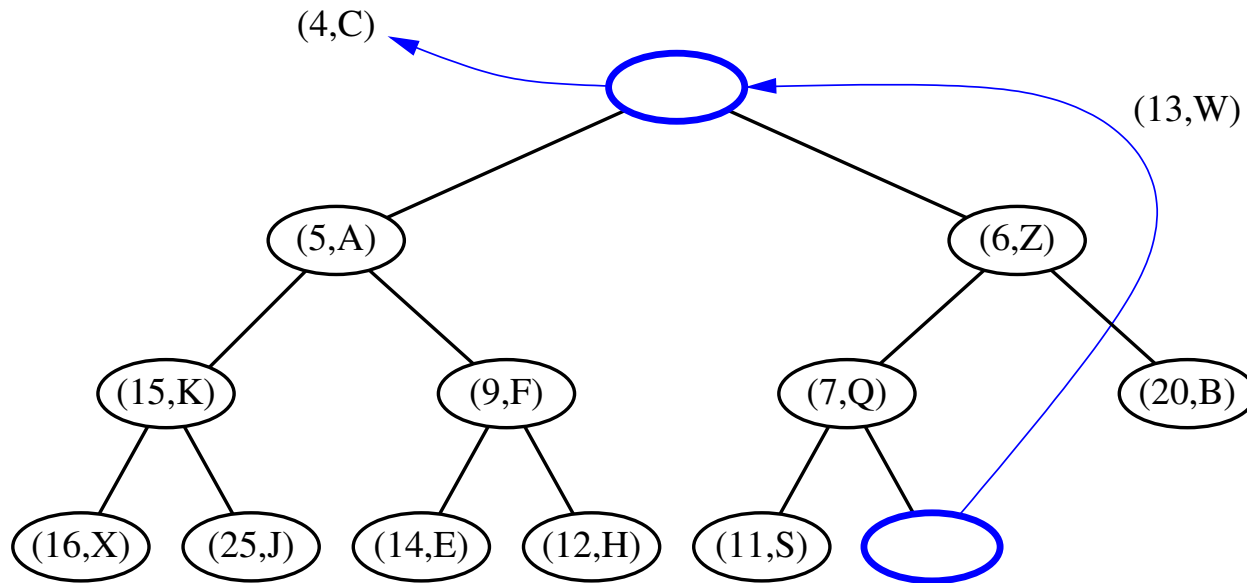
Since a heap has height  $O(\log n)$ , up-heap runs in  $O(\log n)$  time

**removeMin()**

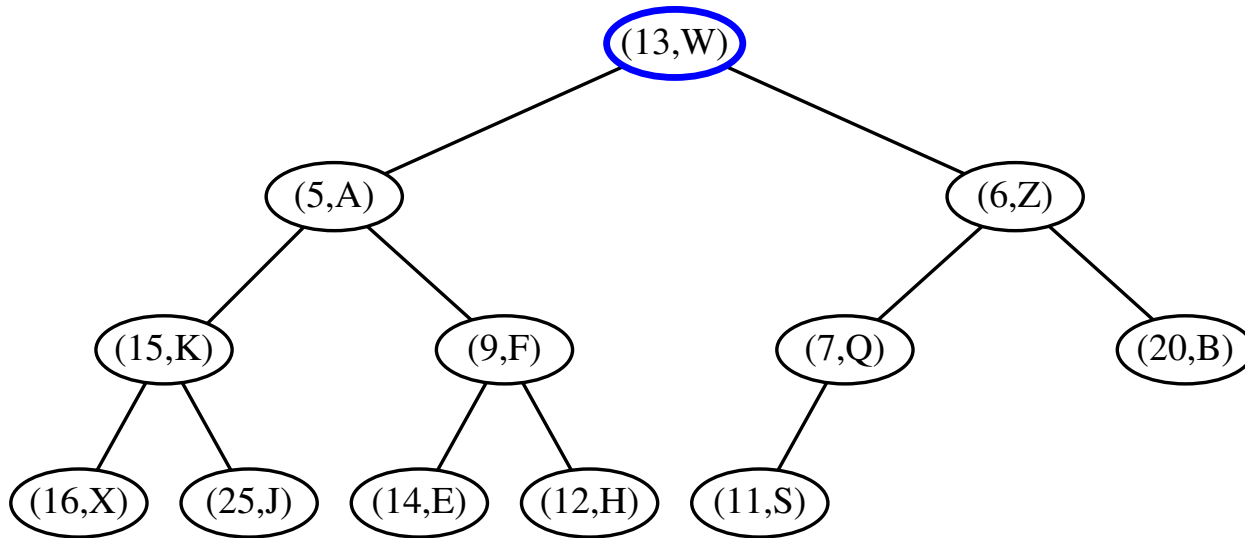
# Removal from a Heap



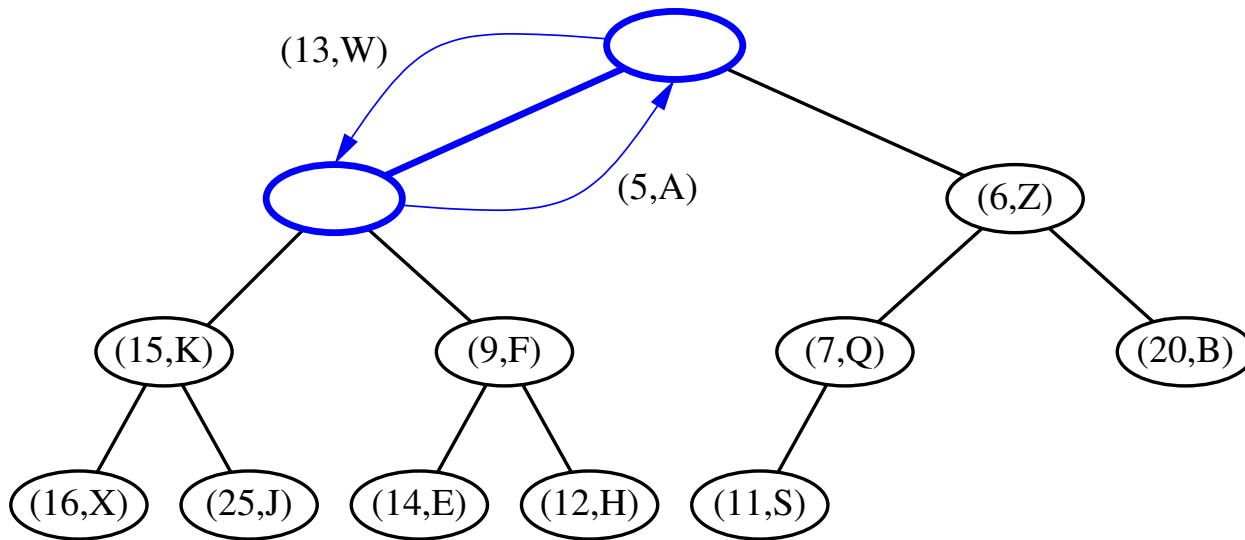
# Removal from a Heap



# down-heap bubbling

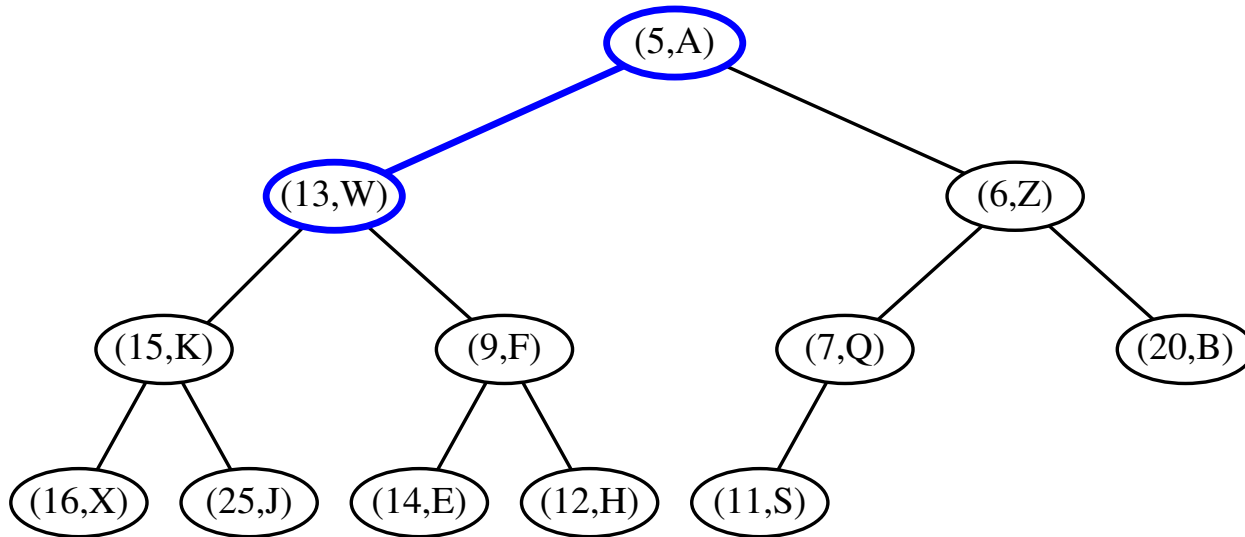


# down-heap bubbling

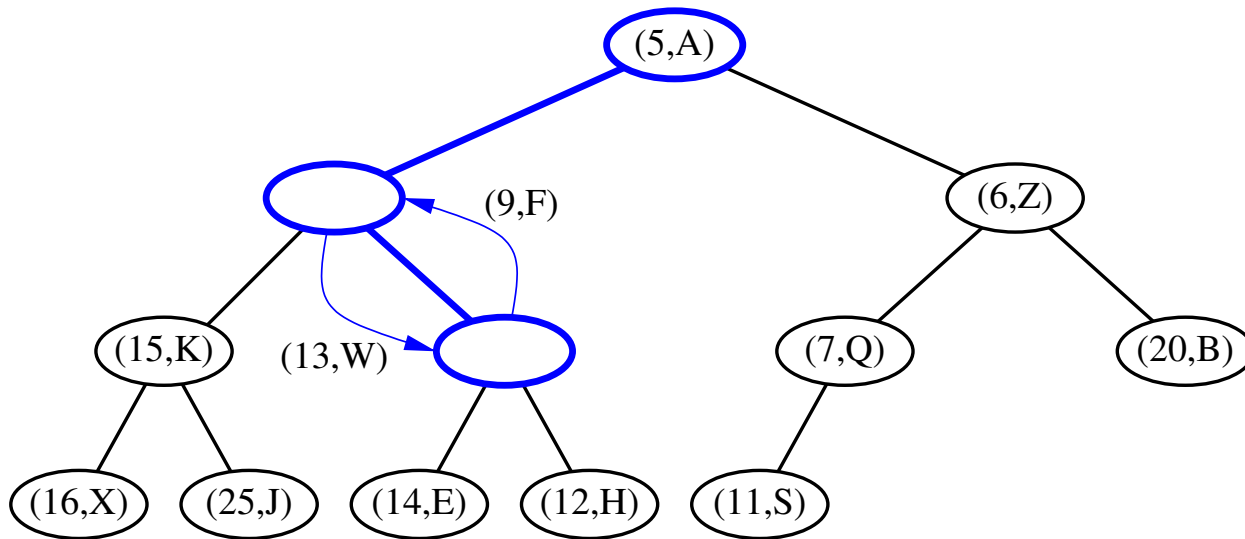




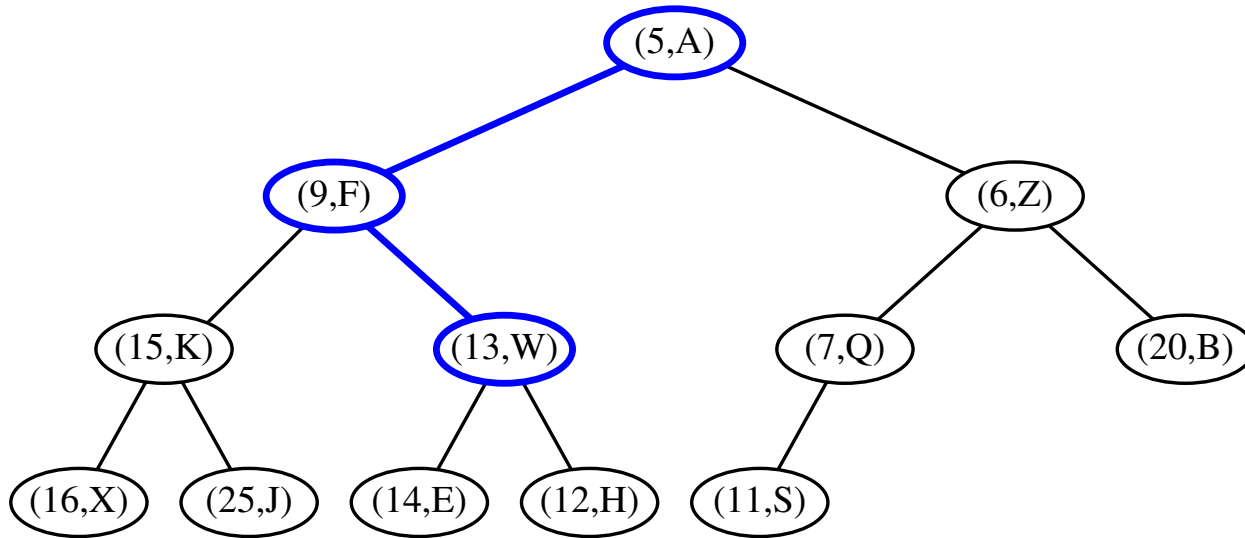
# down-heap bubbling



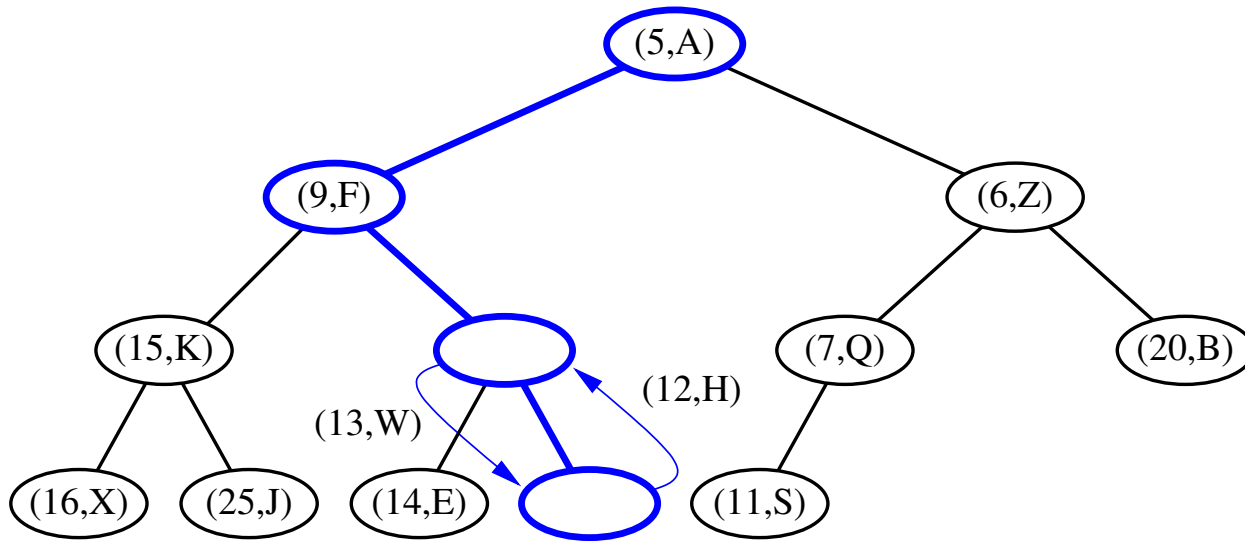
# down-heap bubbling



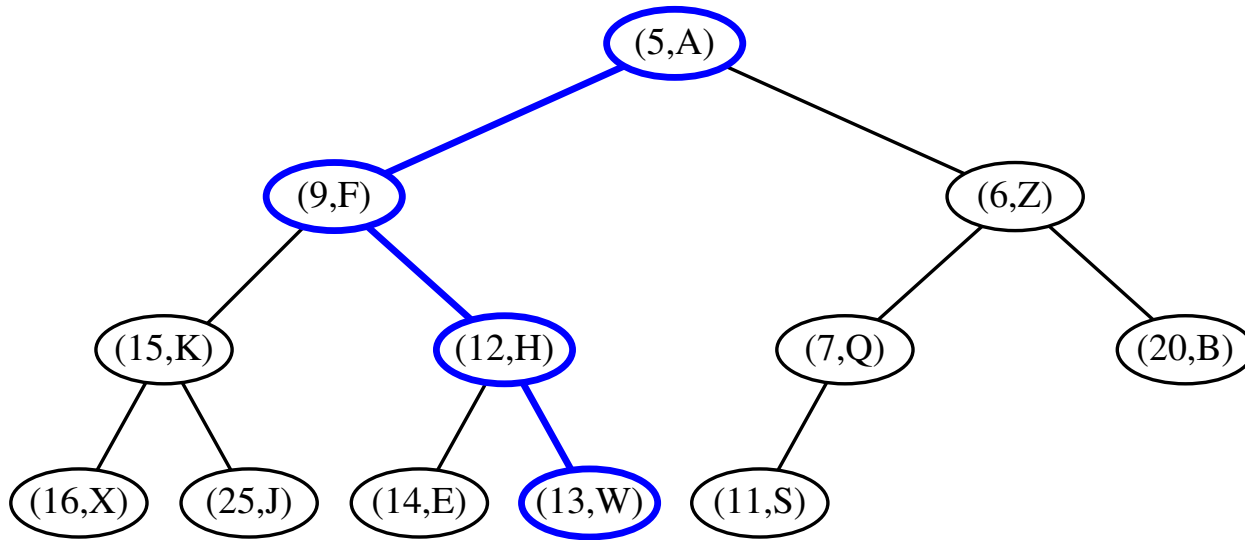
# down-heap bubbling



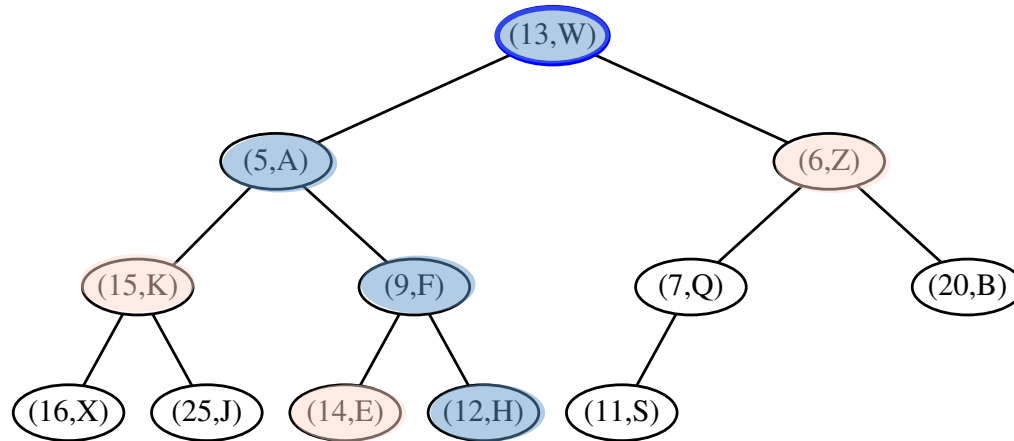
# down-heap bubbling



# down-heap bubbling



# Correctness of Downheap

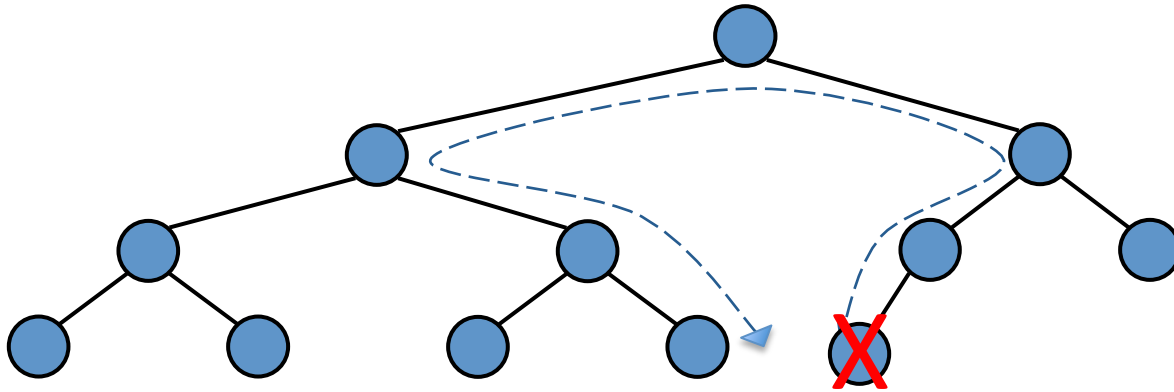


Since a heap has height  $O(\log n)$ , down-heap runs in  $O(\log n)$  time

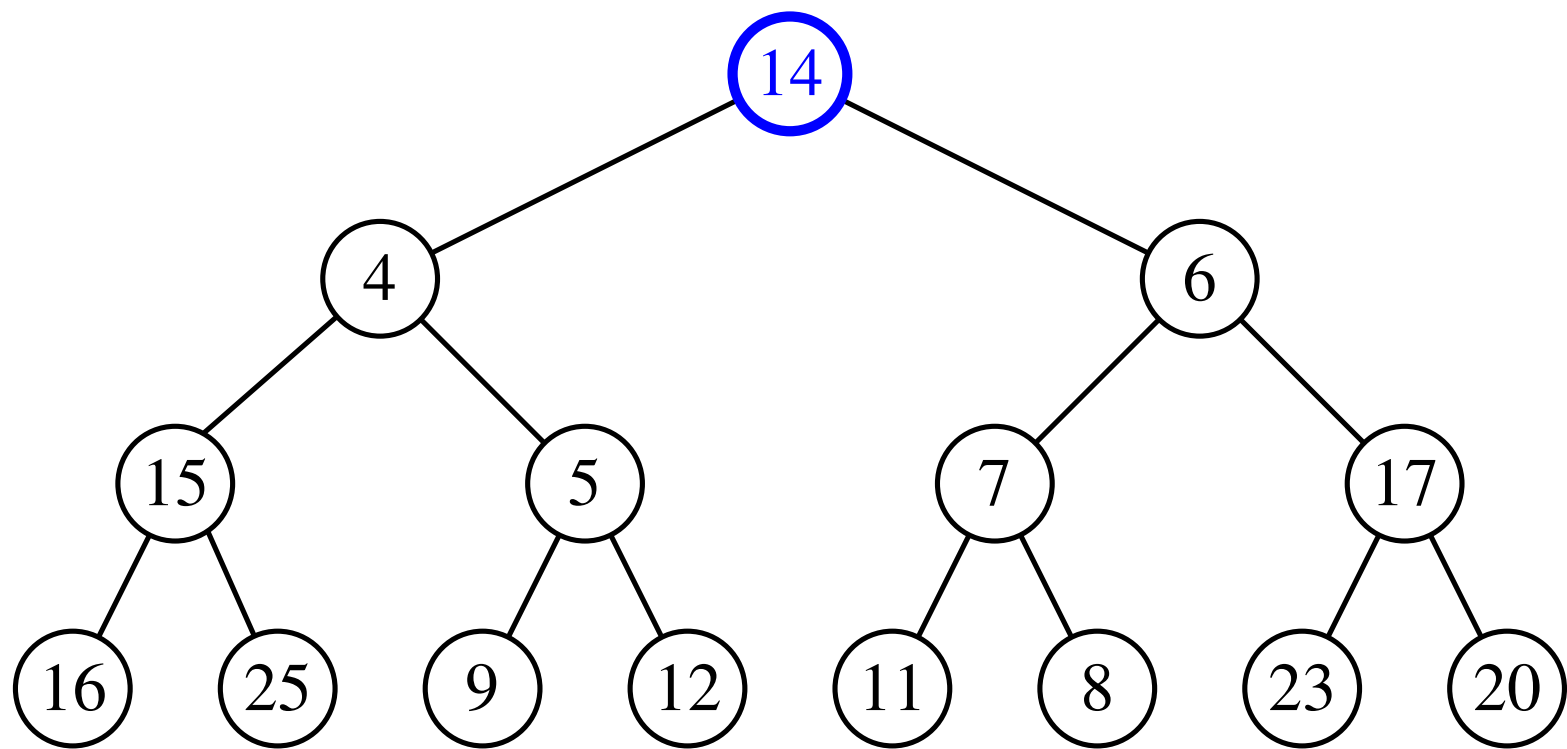
Keeping track of last  
node after  
removeMin()



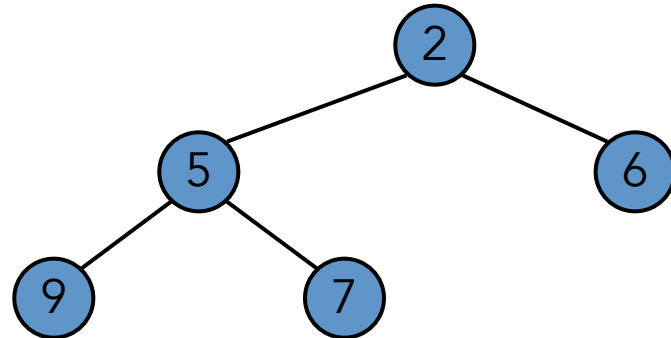
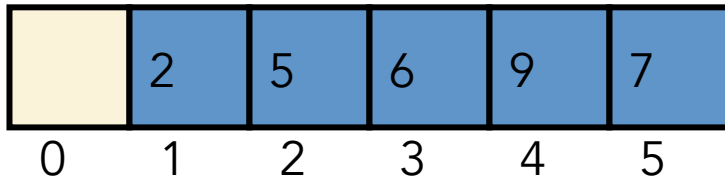
# Similar to finding node for insertion



# Vector Implementation of Complete Binary Tree



- start at rank 1
- for the node at rank  $i$ 
  - the left child is at rank  $2i$
  - the right child is at rank  $2i + 1$



`insert()` and  
`removeMin()` on  
vector heap!

# How to build a heap?

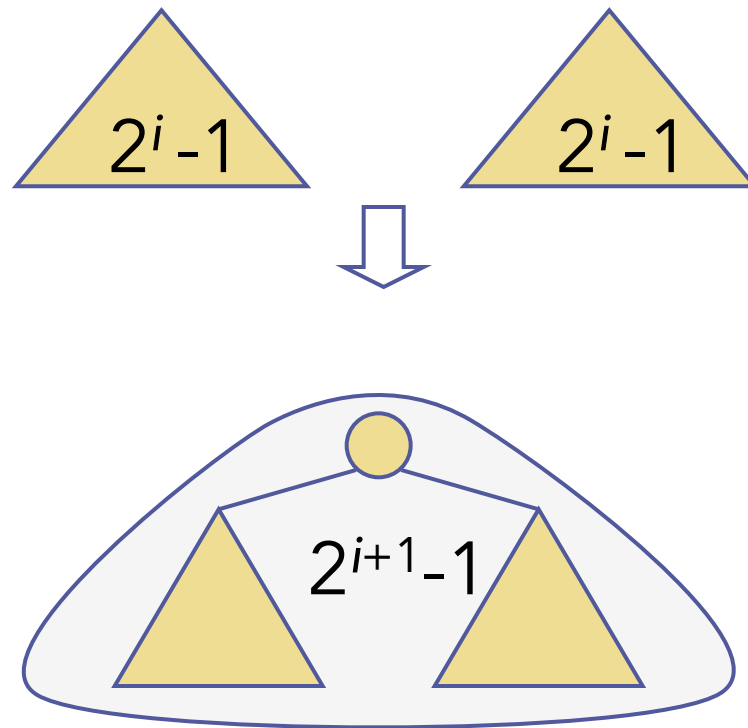
1. insert repeatedly  
 $O(n \log n)$

# Build a heap with the below keys!

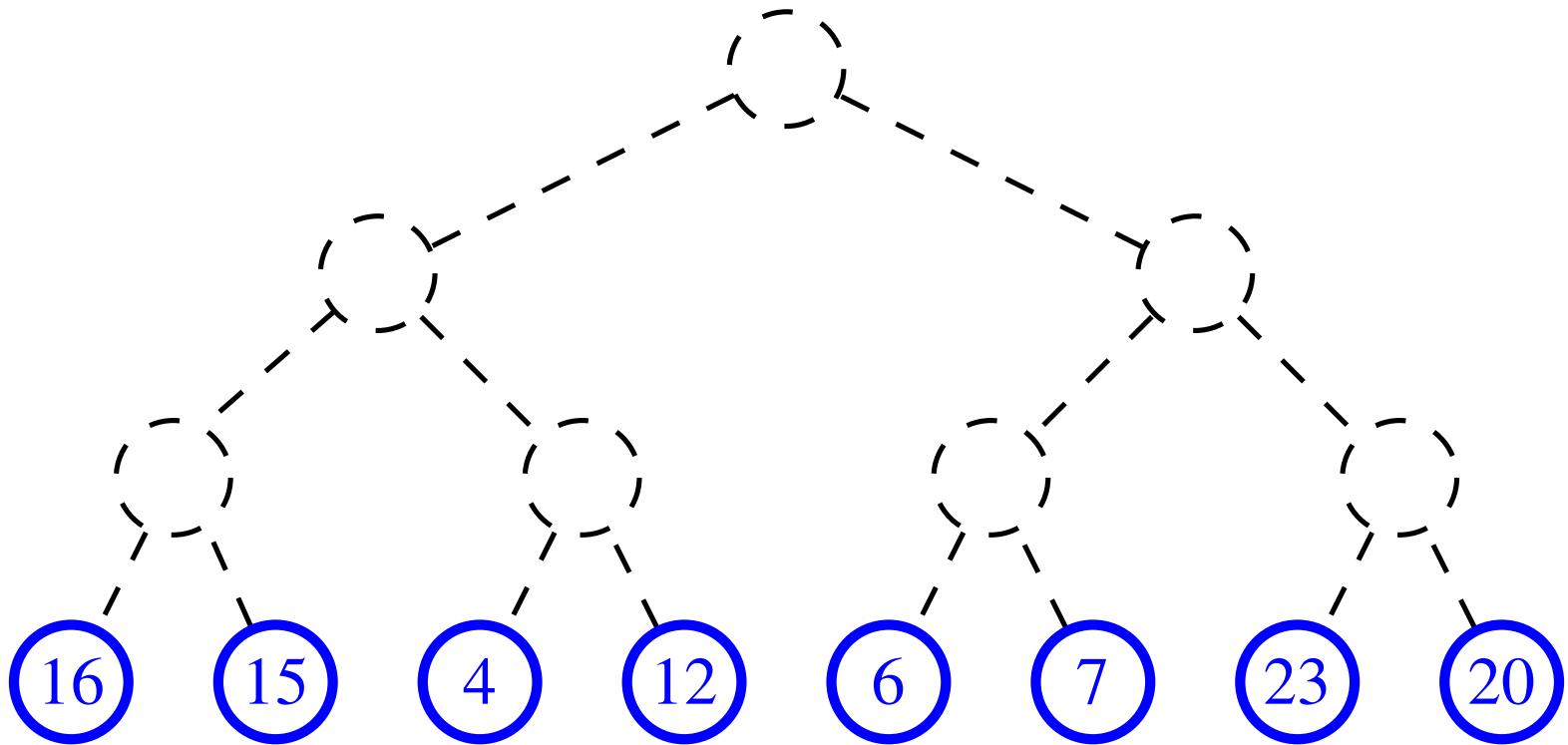
(16, 15, 4 12, 6, 7, 23, 20, 25, 9, 11, 17,  
5, 8, 14)



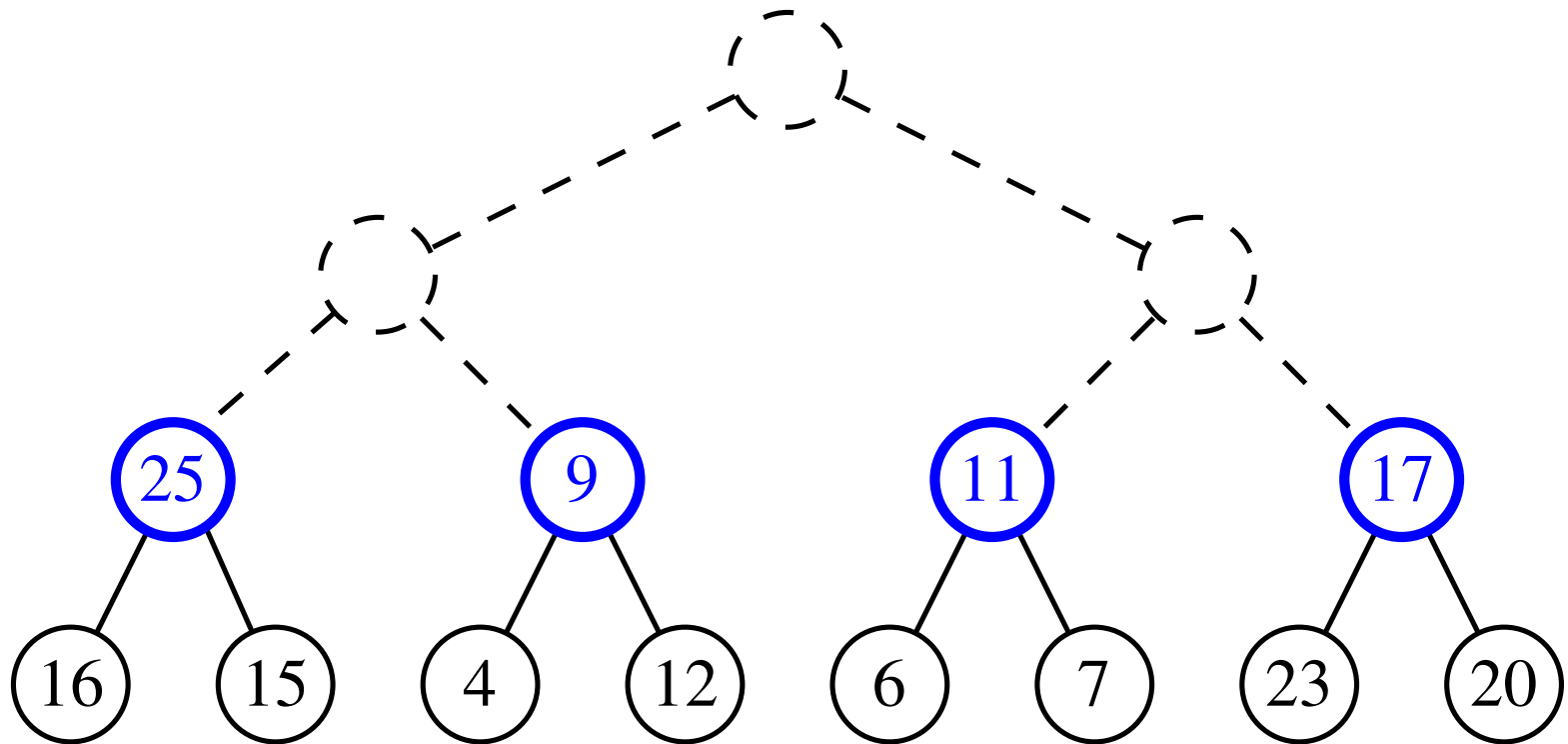
## 2. Bottom-up Heap Construction



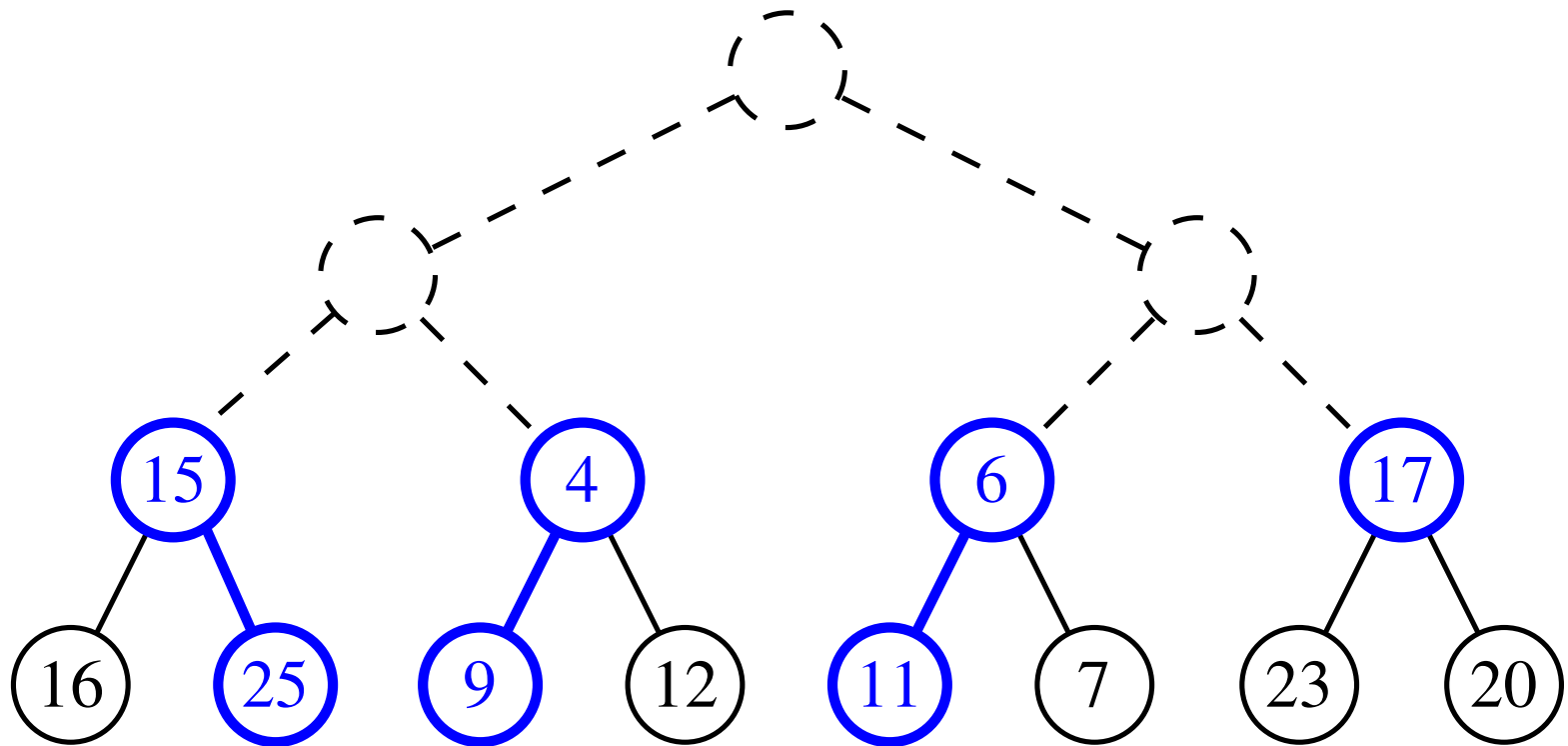
# Building a Heap – Bottomup



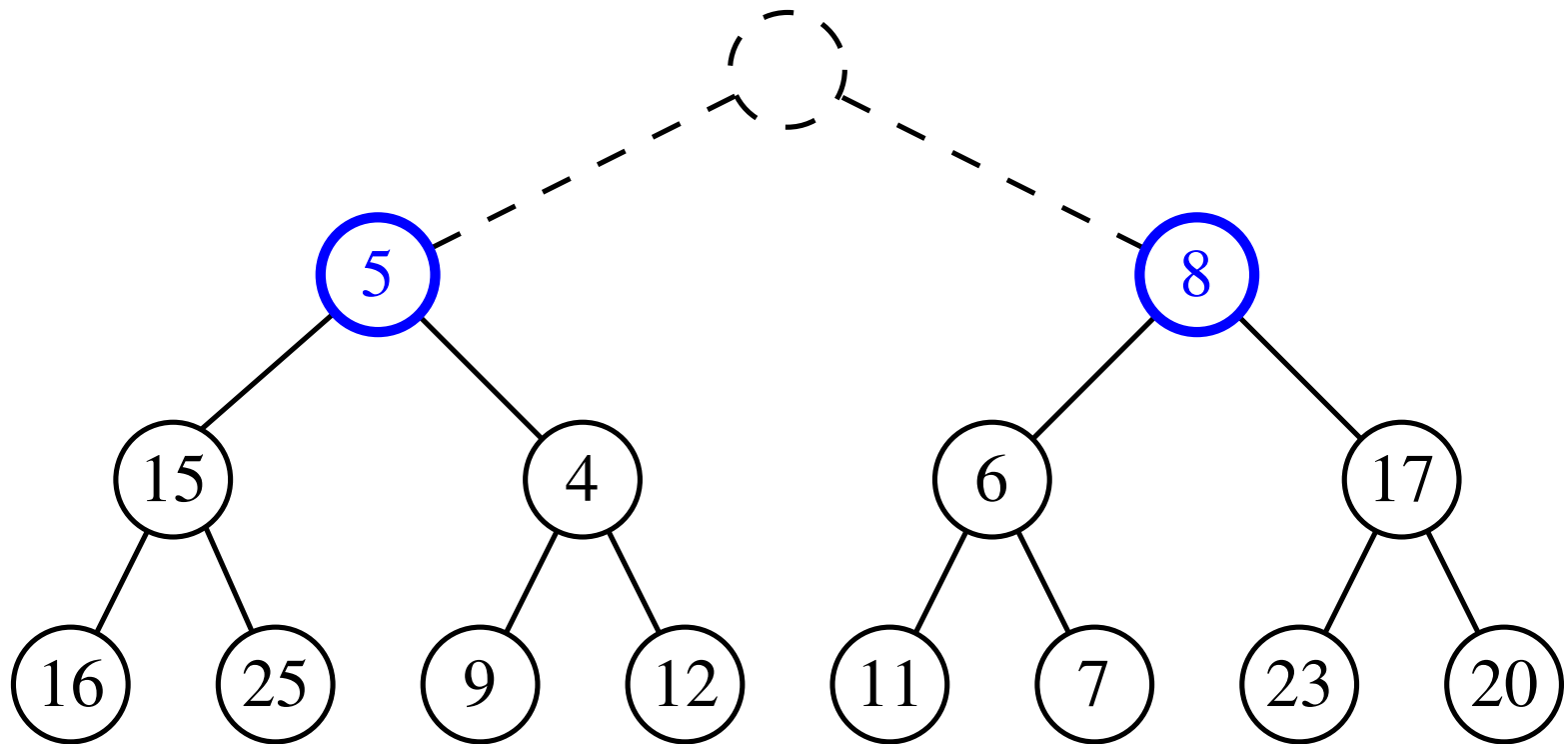
# Building a Heap – Bottomup



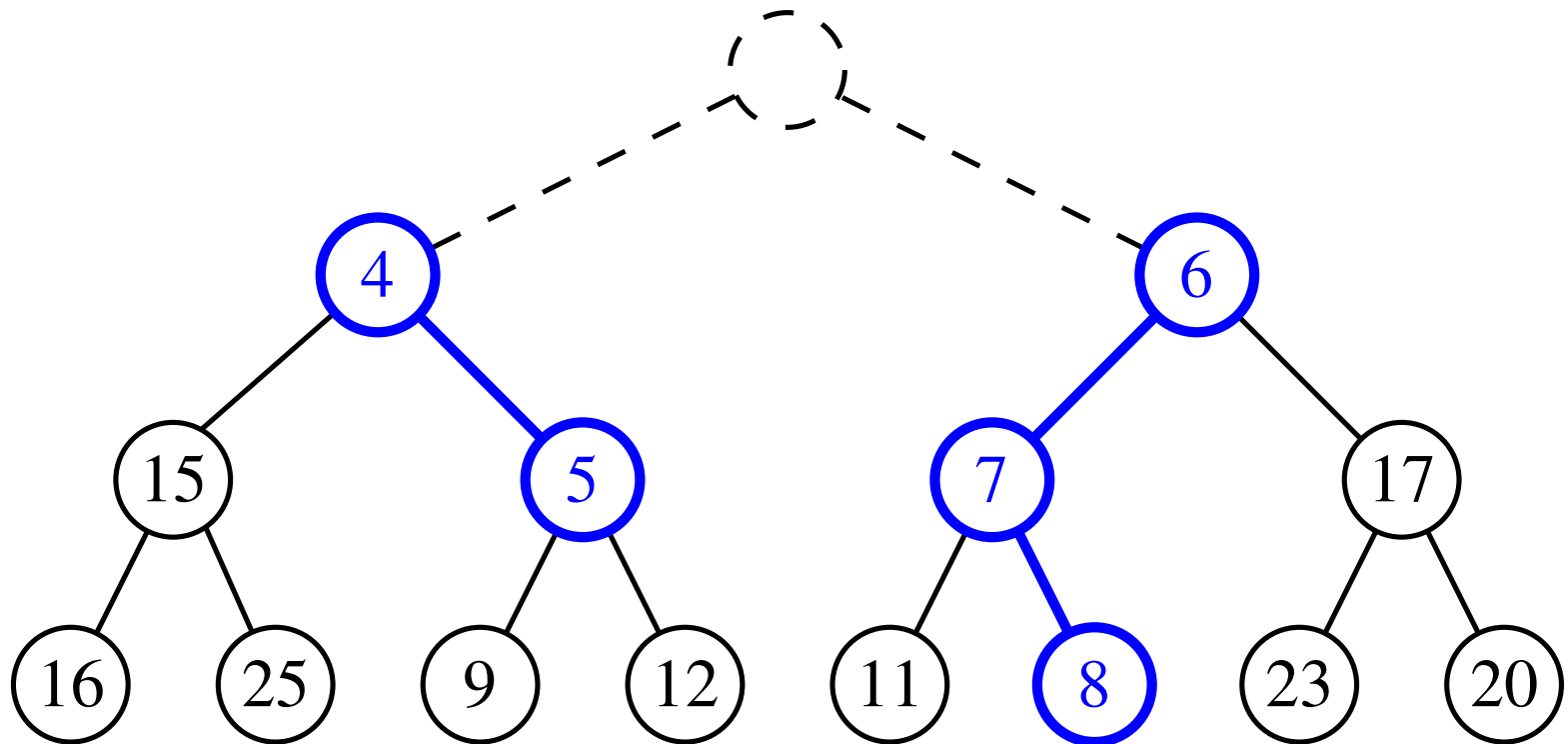
# Building a Heap – Bottomup



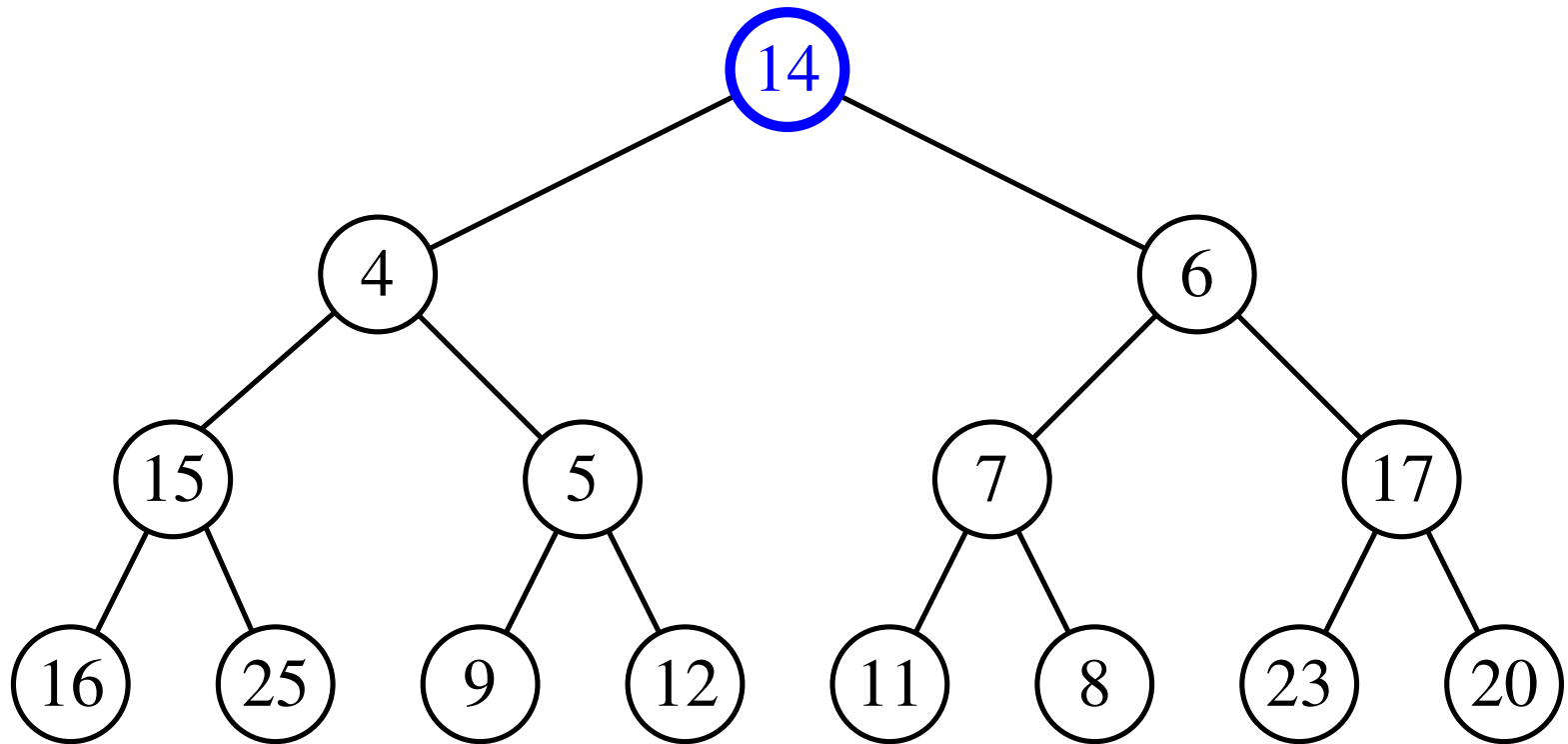
# Building a Heap – Bottomup



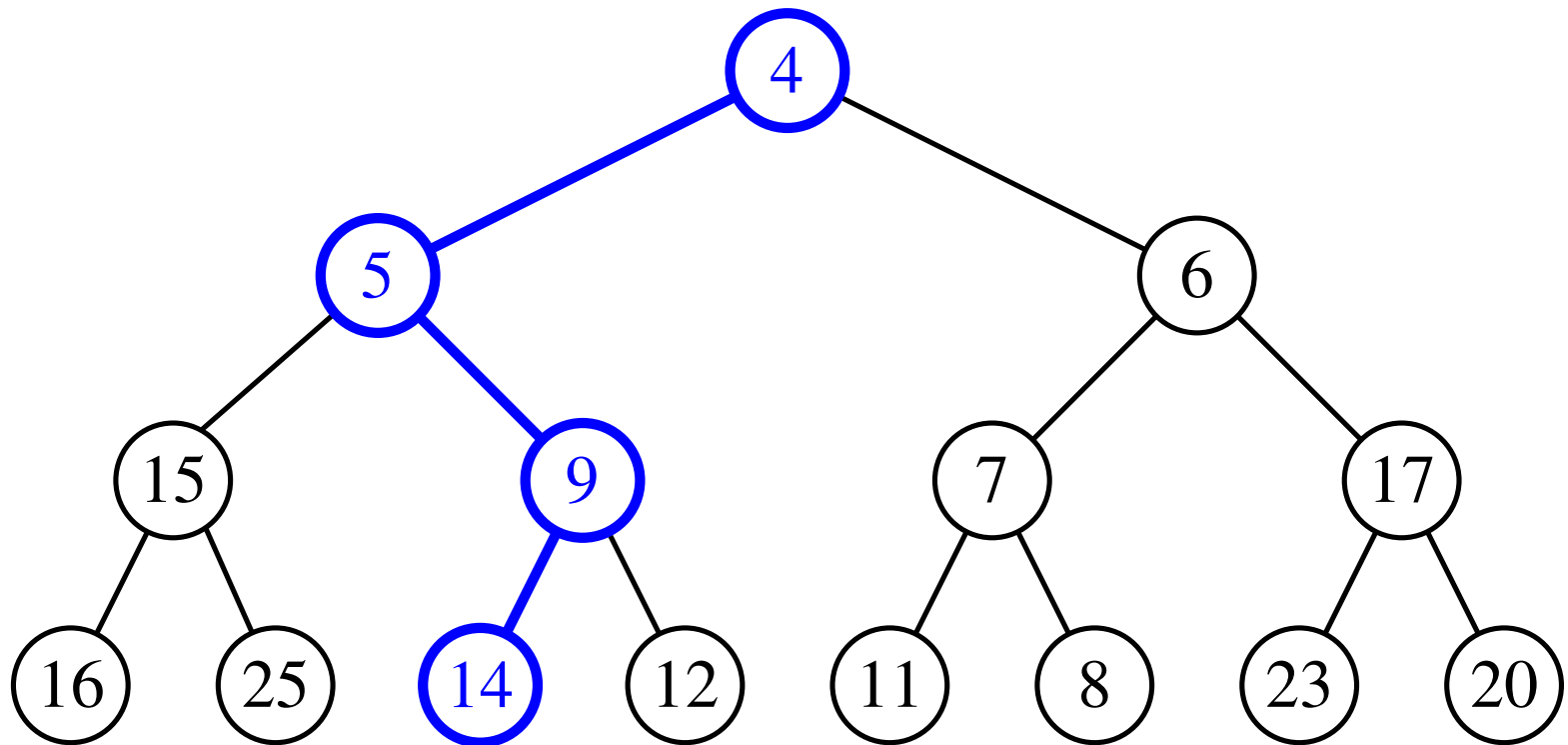
# Building a Heap – Bottomup



# Building a Heap – Bottomup



# Building a Heap – Bottomup

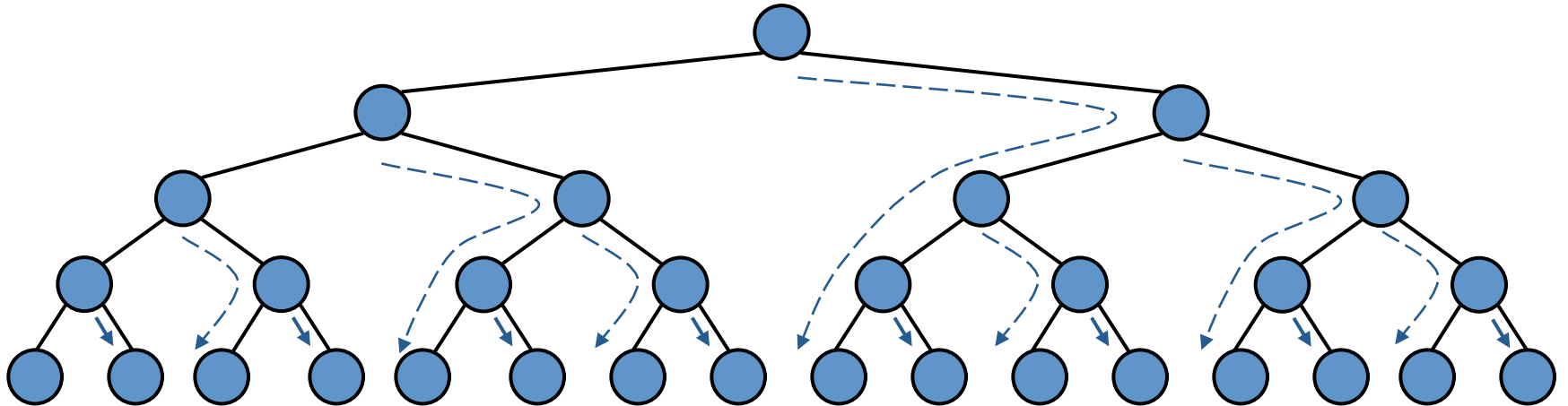




in phase  $i$ , pairs of heaps  
with  $2^i - 1$  keys are merged  
into heaps with  $2^{i+1} - 1$  keys

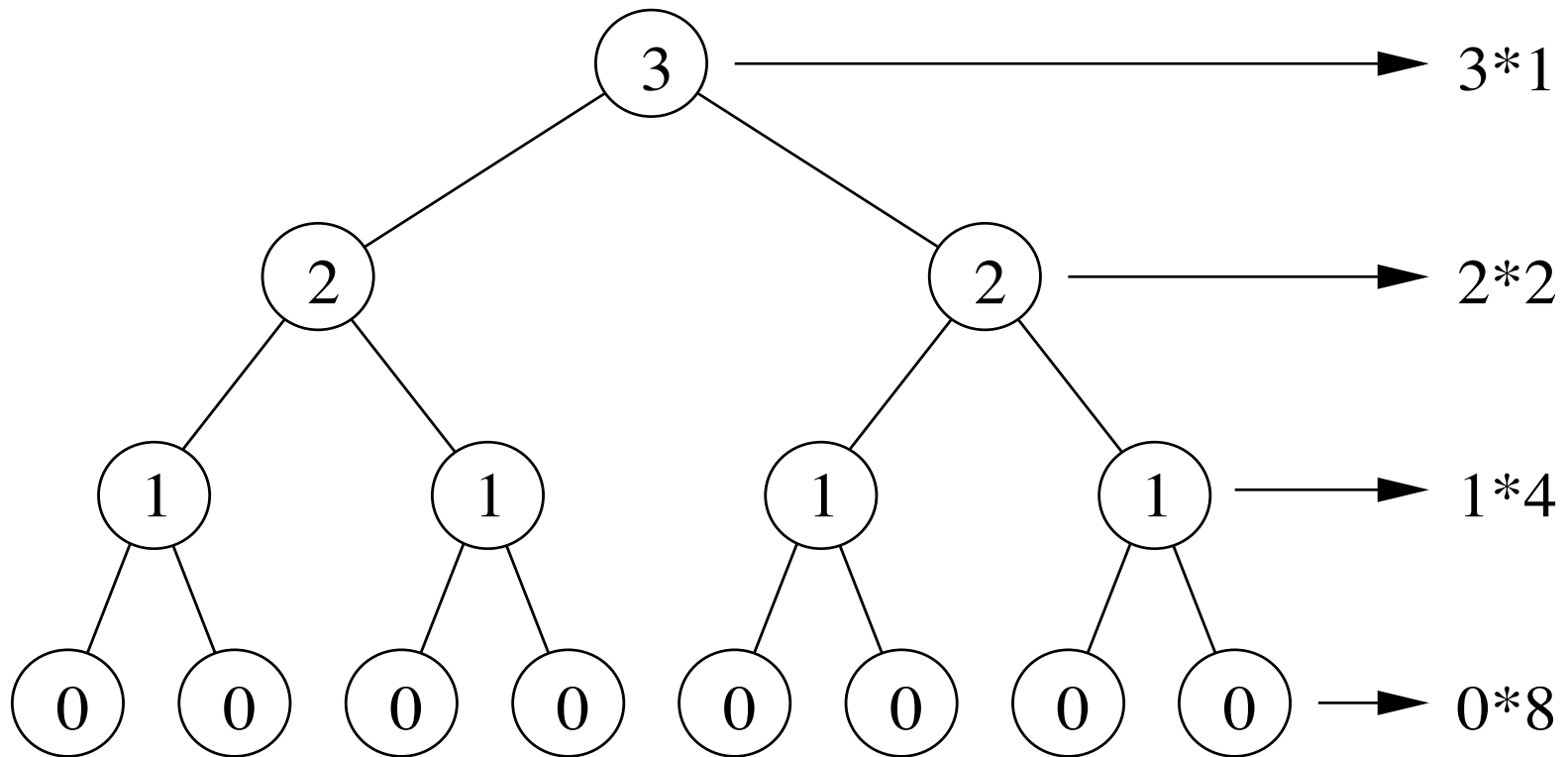
# Analysis 1

-first right, then left until leaf

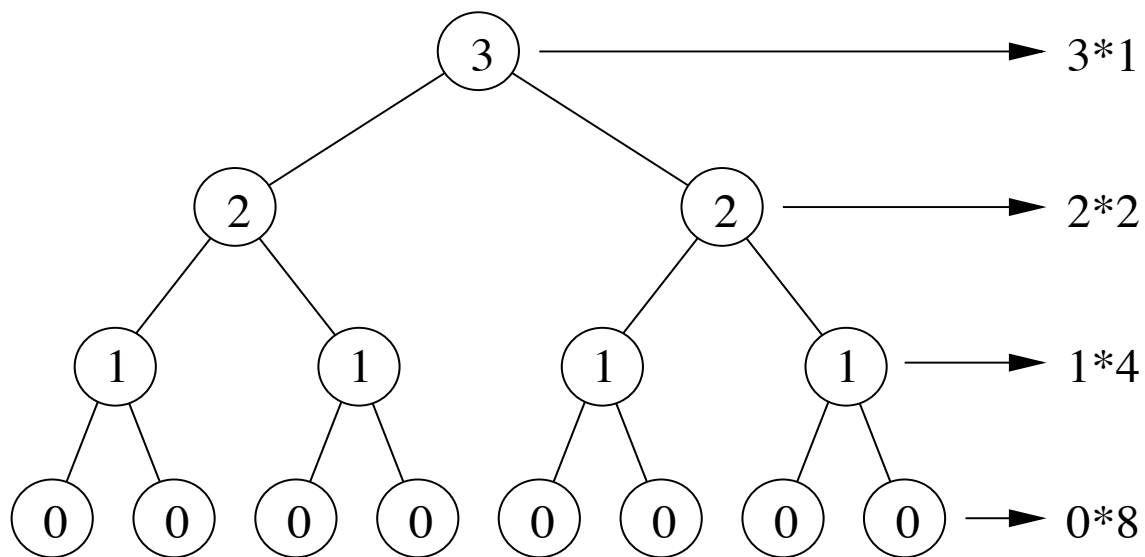


# Analysis 2

-sum of heights of all nodes



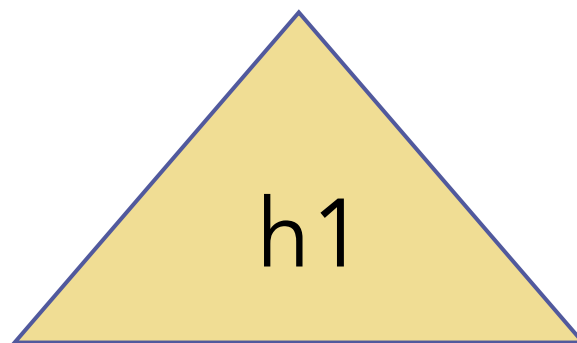
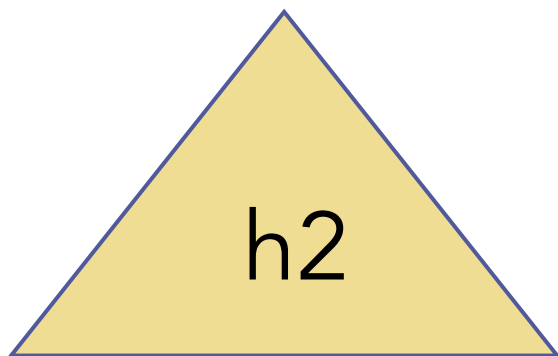
$$\sum_{j=0}^h j 2^{h-j}$$

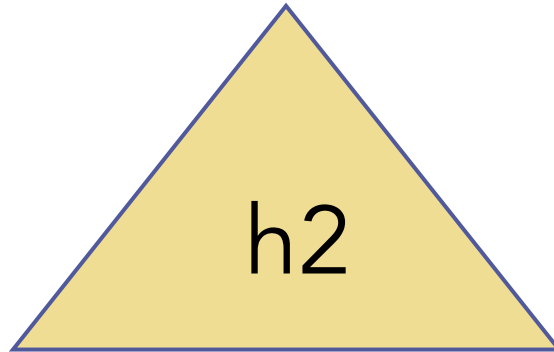
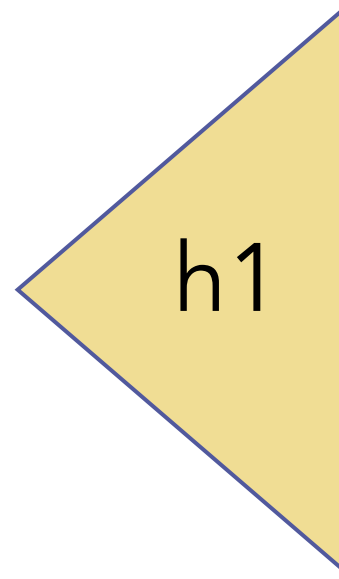


sum of heights

$O(n)$

How to merge two  
heaps?



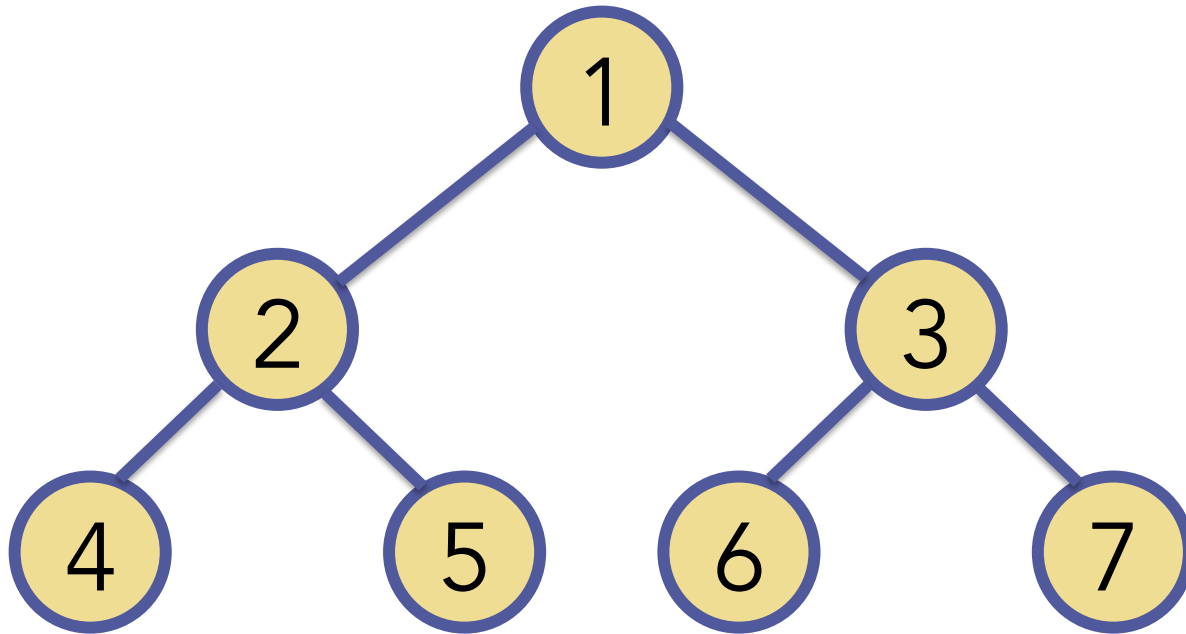


Time Complexity??

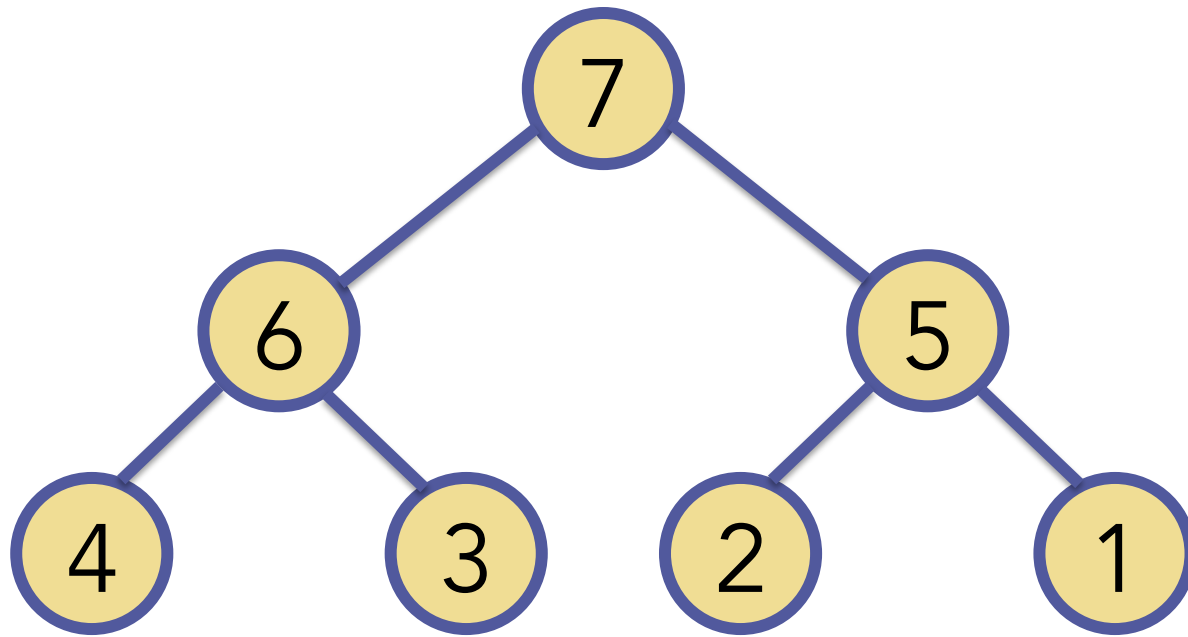


Merge two heaps in  
 $O(n)$

# Min Heap



# Max Heap



# Recall Priority Queue ADT

- a collection of entries
- `entry = (key, value)`
- main methods
  - `insert(k, v)`
  - `removeMin()`
- additional methods
  - `min()`
  - `size()`,
  - `isEmpty()`

# Heap-Sort

```
while !L.empty ()  
    e ← L.front();  
    L.eraseFront()  
    P.insert (e)  
while !P.empty()  
    e ← P.removeMin()  
    L.insertBack(e)
```

Check if a given  
Binary Tree is Heap

Converting min-heap  
into max-heap

# Implementing Heap-Sort In-Place

- Use left side array for heap and right side array for list
- move from left to right, and then right to left



| 4 7 2 1 3 6 5  
4 | 7 2 1 3 6 5  
4 7 | 2 1 3 6 5  
7 4 | 2 1 3 6 5  
⋮  
7 4 6 1 3 2 5|

What should be the next step?

What is the time complexity?

$$n \log n + n \log n$$

Can we make it faster?

$$n + n \log n$$

# Bottom-up Heap Construction for a Linked List implementation

- BottomUpHeap( $L$ ):
- if  $L.empty()$  then return an empty heap
- $e \leftarrow L.front()$
- $L.pop\ front()$
- Split  $L$  into two lists,  $L1$  and  $L2$ , each of size  $(n - 1)/2$
- $T1 \leftarrow \text{BottomUpHeap}(L1)$
- $T2 \leftarrow \text{BottomUpHeap}(L2)$
- Create binary tree  $T$  with root  $r$  storing  $e$ , left subtree  $T1$ , and right subtree  $T2$
- Perform a down-heap bubbling from the root  $r$  of  $T$ , if necessary
- return  $T$

# How to find top k students who will be allowed to go for 6 month internship?

E.g.  $A = [8.1, 7.2, 7.5, 9, 9.8, 10, 5.4]$ ,  $k = 3$

Output =  $[10, 9.8, 9]$

- Solution 1: Sort the numbers  $\Rightarrow O(n \log n)$
- Solution 2: Select min k times  $\Rightarrow O(nk)$
- Solution 3: Build min-heap  $\Rightarrow O(n + k \log n)$
- Solution 4: Build min-heap of first k elements, for rest:
  - if  $<$  root then ignore else replace root with the number and heapify
  - $O(k + (n-k) \log k)$
- More?

A company has  $n$  items ( $w_1..w_n$ ) and wants to make packages of minimum weight  $W$ . Give a fast algorithms to do this in minimum number of steps.

- Sort, add first two, sort again, stop when first element is greater than  $W$ 
  - $O(n \log n)$
- Create min heap, compare  $W$  with the root, if smaller, remove two elements, add them, and insert into the heap, repeat the procedure
  - $O(n + x \log n)$