

# SE274 Data Structure

## Lecture 5: Heaps, Priority Queues

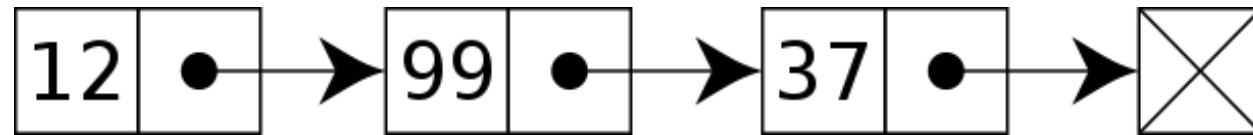
Mar 30, 2020

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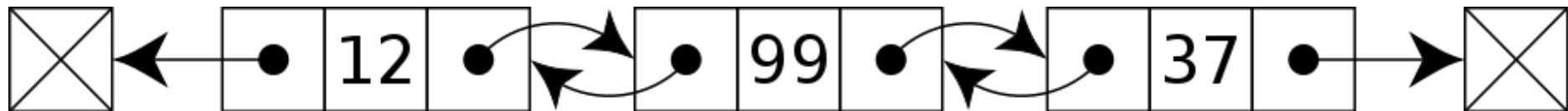
Information&Communication Engineering, DGIST

# Recap: Linked List

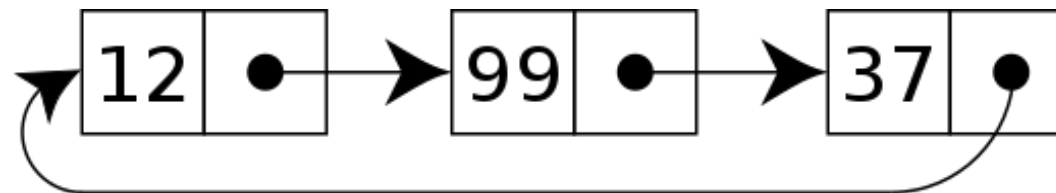
- Singly linked list



- Doubly linked list

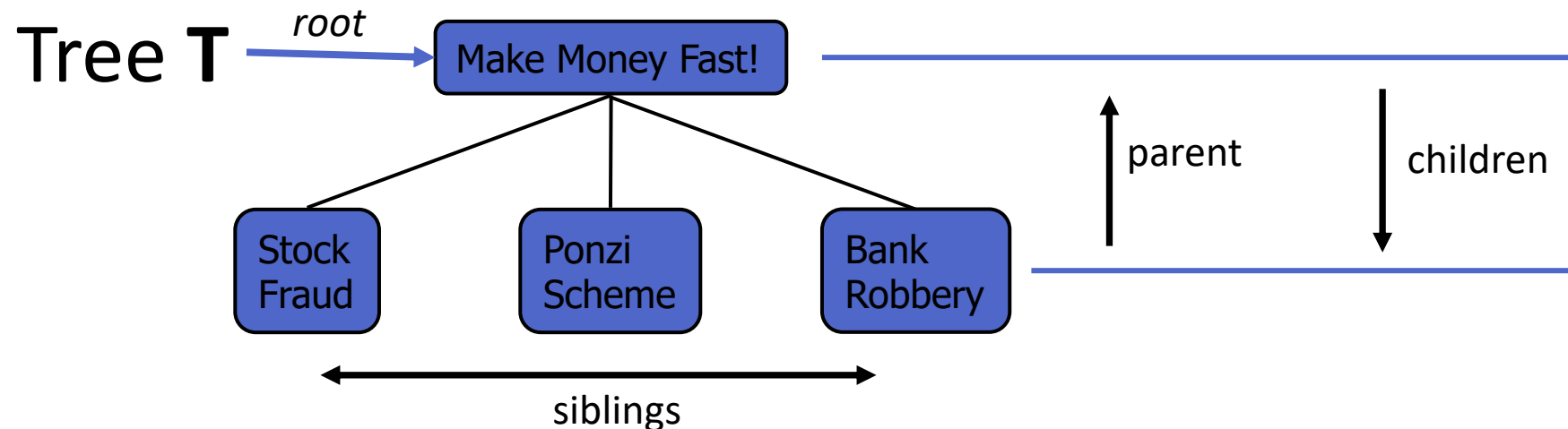


- Circular linked list



# Recap: Tree

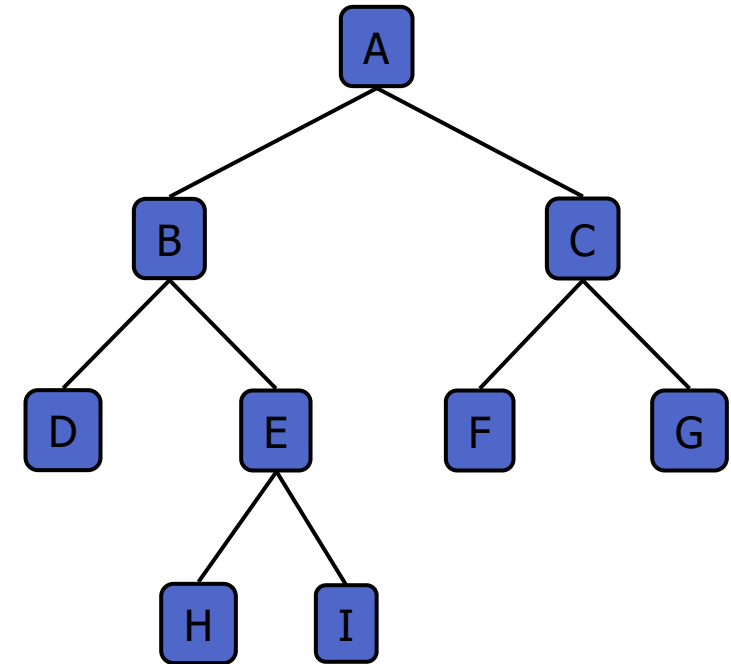
- We define a *tree* **T** as a set of **nodes** storing elements such that **Nodes** have a **parent-child** relationship, that satisfies:
  - If **T** is nonempty, it has a special node, called the **root** of **T**.
  - Each node **v** of **T** different from the root has a unique *parent node* **w**; every node with parent **w** is a child of **w**, nodes that share the same parent are called *siblings*.



# Recap: Binary Trees

- A binary tree is a tree with the following properties:
  - Each internal node has at most two children (exactly two for proper binary trees)
  - The children of a node are an ordered pair
- We call the children of an internal node **left child** and **right child**
- Alternative recursive definition: a binary tree is either
  - a tree consisting of a single node, or
  - a tree whose root has an ordered pair of children, each of which is a binary tree

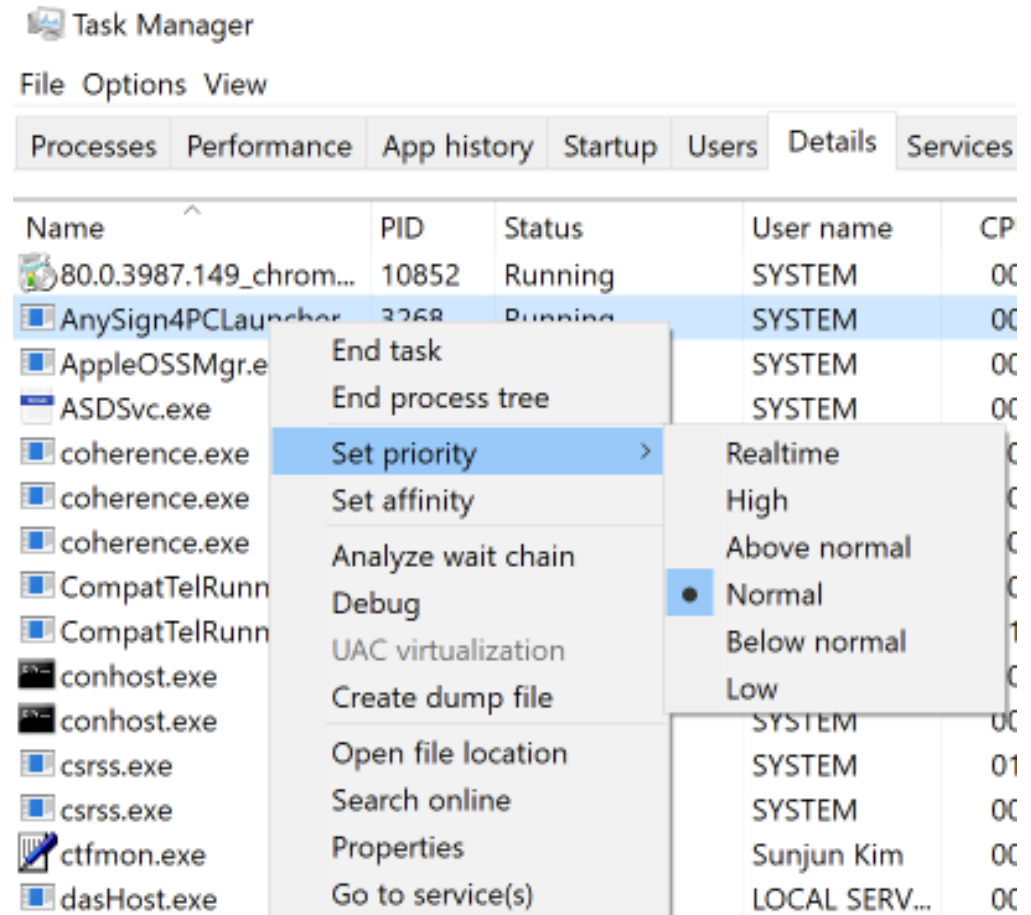
- Applications:
  - arithmetic expressions
  - decision processes
  - searching



# Priority Queues



# Priority Queue: practical examples



# Priority Queue: idea

- Assigning “priority” to Queue items.
- Each element is associated with a **key**, which represents the **priority** of the element.
  - Convention: lower key value = higher its priority.
    - $3 < 5$
    - $abc < fes$
    - 🐭 < 🐯 < 🐷
    - 나 < 너
  - Keys are not necessarily unique.

# Total Order Relations

- Keys in a priority queue can be arbitrary objects on which an order is defined as a key with a ***total order*** relation
- Two distinct entries in a priority queue can have the same key
- ***This prevents conflict between two items.***
- Mathematical concept of ***total order relation***  $\leq$ 
  - Reflexive property:  
 $x \leq x$
  - Antisymmetric property:  
 $x \leq y \wedge y \leq x \Rightarrow x = y$
  - Transitive property:  
 $x \leq y \wedge y \leq z \Rightarrow x \leq z$
- ***Example of total ordered set)***
  - The letters of Alphabet, in dictionary order.
  - Set of integers, real numbers, etc.
  - Ordered fields



# Priority Queue ADT

- A priority queue stores a collection of items
- Each item is a pair (key, value)
- Main methods of the Priority Queue ADT
  - **add** (k, x)  
inserts an item with key k and value x
  - **remove\_min**()  
removes and returns the item with smallest key

- Additional methods
  - **min**()  
returns, but does not remove, an item with smallest key
  - **len(P)**, **is\_empty**()
- Applications:
  - Standby flyers
  - Auctions
  - Stock market

# Priority Queue Example

Operation	Return Value	Priority Queue
P.add(5,A)		{(5,A)}
P.add(9,C)		{(5,A), (9,C)}
P.add(3,B)		{(3,B), (5,A), (9,C)}
P.add(7,D)		{(3,B), (5,A), (7,D), (9,C)}
P.min()	(3,B)	{(3,B), (5,A), (7,D), (9,C)}
P.remove_min()	(3,B)	{(5,A), (7,D), (9,C)}
P.remove_min()	(5,A)	{(7,D), (9,C)}
len(P)	2	{(7,D), (9,C)}
P.remove_min()	(7,D)	{(9,C)}
P.remove_min()	(9,C)	{ }
P.is_empty()	True	{ }
P.remove_min()	“error”	{ }

# Composition Design Pattern

- An `item` in a priority queue is simply a key-value pair
- Priority queues store items to allow for efficient insertion and removal based on keys

```
1 class PriorityQueueBase:
2     """Abstract base class for a priority queue."""
3
4     class _Item:
5         """Lightweight composite to store priority queue items."""
6         __slots__ = '_key', '_value'
7
8         def __init__(self, k, v):
9             self._key = k
10            self._value = v
11
12            def __lt__(self, other):
13                return self._key < other._key    # compare items based on their keys
14
15            def is_empty(self):
16                """Return True if the priority queue is empty."""
17                return len(self) == 0
```

`__lt__()` = larger than

→ Is called when ...

`A = _Item(aa, ab)`

`B = _Item(ba, bb)`

if `A < B`:

....

# Sequence-based Priority Queue

- Implementation with an unsorted list



- Performance:
  - `add` takes  $O(1)$  time since we can insert the item at the beginning or end of the sequence
  - `Remove_min` and `min` take  $O(n)$  time since we have to traverse the entire sequence to find the smallest key

- Implementation with a sorted list



- Performance:
  - `add` takes  $O(n)$  time since we have to find the place where to insert the item
  - `remove_min` and `min` take  $O(1)$  time, since the smallest key is at the beginning

# Note: Positional List

- To provide for a general abstraction of a sequence of elements with the ability to identify the location of an element, we define a **positional list** ADT.
- A position acts as a marker or token within the broader positional list.
- A position  $p$  is unaffected by changes elsewhere in a list; the only way in which a position becomes invalid is if an explicit command is issued to delete it.
- A position instance is a simple object, supporting only the following method:
  - $p.\text{element}()$ : Return the element stored at position  $p$ .

# Note: Positional Accessor Operations

`L.first()`: Return the position of the first element of `L`, or `None` if `L` is empty.

`L.last()`: Return the position of the last element of `L`, or `None` if `L` is empty.

`L.before(p)`: Return the position of `L` immediately before position `p`, or `None` if `p` is the first position.

`L.after(p)`: Return the position of `L` immediately after position `p`, or `None` if `p` is the last position.

`L.is_empty()`: Return `True` if list `L` does not contain any elements.

`len(L)`: Return the number of elements in the list.

`iter(L)`: Return a forward iterator for the *elements* of the list. See Section 1.8 for discussion of iterators in Python.

# Note: Positional Update Operations

`L.add_first(e)`: Insert a new element `e` at the front of `L`, returning the position of the new element.

`L.add_last(e)`: Insert a new element `e` at the back of `L`, returning the position of the new element.

`L.add_before(p, e)`: Insert a new element `e` just before position `p` in `L`, returning the position of the new element.

`L.add_after(p, e)`: Insert a new element `e` just after position `p` in `L`, returning the position of the new element.

`L.replace(p, e)`: Replace the element at position `p` with element `e`, returning the element formerly at position `p`.

`L.delete(p)`: Remove and return the element at position `p` in `L`, invalidating the position.

# Unsorted List Implementation

```
1 class UnsortedPriorityQueue(PriorityQueueBase): # base class defines _Item
2     """ A min-oriented priority queue implemented with an unsorted list. """
3
4     def _find_min(self): # nonpublic utility
5         """ Return Position of item with minimum key. """
6         if self.is_empty(): # is_empty inherited from base class
7             raise Empty('Priority queue is empty')
8         small = self._data.first()
9         walk = self._data.after(small)
10        while walk is not None:
11            if walk.element() < small.element():
12                small = walk
13            walk = self._data.after(walk)
14        return small
15
16    def __init__(self):
17        """ Create a new empty Priority Queue. """
18        self._data = PositionalList()
19
20    def __len__(self):
21        """ Return the number of items in the priority queue. """
22        return len(self._data)
```

```
23
24    def add(self, key, value):
25        """ Add a key-value pair. """
26        self._data.add_last(self._Item(key, value))
27
28    def min(self):
29        """ Return but do not remove (k,v) tuple with minimum key. """
30        p = self._find_min()
31        item = p.element()
32        return (item._key, item._value)
33
34    def remove_min(self):
35        """ Remove and return (k,v) tuple with minimum key. """
36        p = self._find_min()
37        item = self._data.delete(p)
38        return (item._key, item._value)
```



# Sorted List Implementation

```
1 class SortedPriorityQueue(PriorityQueueBase): # base class defines _Item
2     """A min-oriented priority queue implemented with a sorted list."""
3
4     def __init__(self):
5         """Create a new empty Priority Queue."""
6         self._data = PositionalList()
7
8     def __len__(self):
9         """Return the number of items in the priority queue."""
10        return len(self._data)
11
12    def add(self, key, value):
13        """Add a key-value pair."""
14        newest = self._Item(key, value) # make new item instance
15        walk = self._data.last( ) # walk backward looking for smaller key
16        while walk is not None and newest < walk.element():
17            walk = self._data.before(walk)
18        if walk is None:
19            self._data.add_first(newest) # new key is smallest
20        else:
21            self._data.add_after(walk, newest) # newest goes after walk
22
```

```
23    def min(self):
24        """Return but do not remove (k,v) tuple with minimum key."""
25        if self.is_empty():
26            raise Empty('Priority queue is empty.')
27        p = self._data.first()
28        item = p.element()
29        return (item._key, item._value)
30
31    def remove_min(self):
32        """Remove and return (k,v) tuple with minimum key."""
33        if self.is_empty():
34            raise Empty('Priority queue is empty.')
35        item = self._data.delete(self._data.first())
36        return (item._key, item._value)
```

# Priority Queue Sorting

- We can use a priority queue to sort a set of comparable elements
  1. Insert the elements one by one with a series of add operations
  2. Remove the elements in sorted order with a series of remove\_min operations
- The running time of this sorting method depends on the priority queue implementation

## Algorithm *PQ-Sort*(*S*, *C*)

**Input** sequence *S*, comparator *C* for the elements of *S*

**Output** sequence *S* sorted in increasing order according to *C*

*P* ← priority queue with comparator *C*

**while**  $\neg S.is\_empty()$

$e \leftarrow S.remove\_first()$

*P.add*(*e*,  $\emptyset$ )

**while**  $\neg P.is\_empty()$

$e \leftarrow P.removeMin().key()$

*S.add\_last*(*e*)

# Selection-Sort

- Selection-sort is the variation of PQ-sort where the priority queue is implemented with an unsorted sequence
- Running time of Selection-sort:
  1. Inserting the elements into the priority queue with  $n$  insert operations takes  $O(n)$  time
  2. Removing the elements in sorted order from the priority queue with  $n$  removeMin operations takes time proportional to
$$1 + 2 + \dots + n$$
- Selection-sort runs in  $O(n^2)$  time

# Selection-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)	(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)		()

# Insertion-Sort

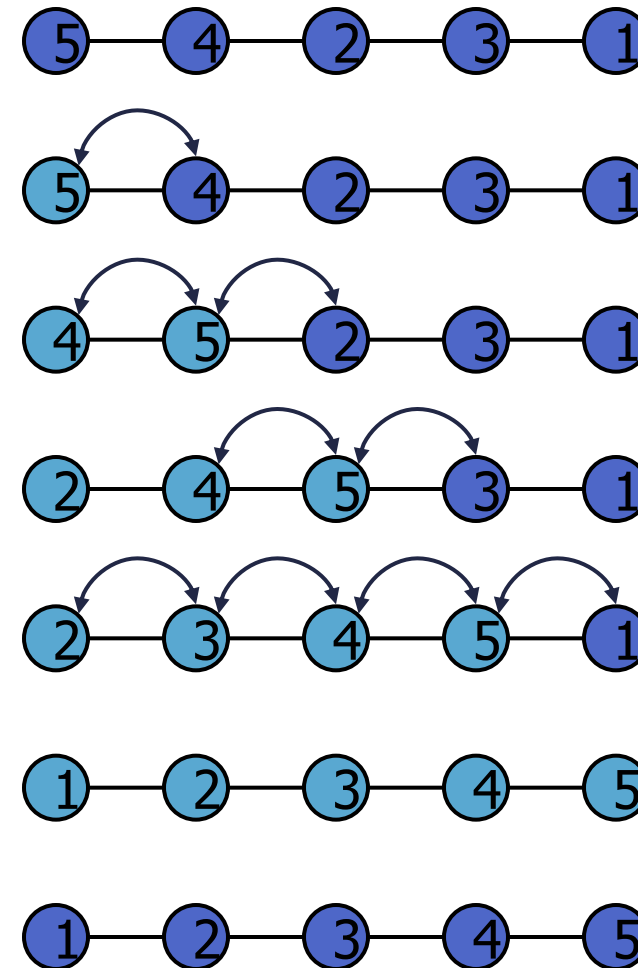
- Insertion-sort is the variation of PQ-sort where the priority queue is implemented with a sorted sequence
- Running time of Insertion-sort:
  1. Inserting the elements into the priority queue with  $n$  insert operations takes time proportional to
$$1 + 2 + \dots + n$$
  2. Removing the elements in sorted order from the priority queue with a series of  $n$  removeMin operations takes  $O(n)$  time
- Insertion-sort runs in  $O(n^2)$  time

# Insertion-Sort Example

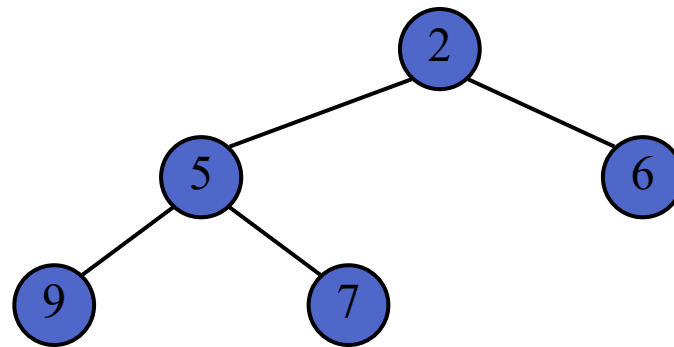
	Sequence S	Priority queue P
Input:	(7,4,8,2,5,3,9)	()
Phase 1		
(a)	(4,8,2,5,3,9)	(7)
(b)	(8,2,5,3,9)	(4,7)
(c)	(2,5,3,9)	(4,7,8)
(d)	(5,3,9)	(2,4,7,8)
(e)	(3,9)	(2,4,5,7,8)
(f)	(9)	(2,3,4,5,7,8)
(g)	()	(2,3,4,5,7,8,9)
Phase 2		
(a)	(2)	(3,4,5,7,8,9)
(b)	(2,3)	(4,5,7,8,9)
..	..	..
(g)	(2,3,4,5,7,8,9)	()

# In-place Insertion-Sort

- Instead of using an external data structure, we can implement selection-sort and insertion-sort in-place
- A portion of the input sequence itself serves as the priority queue
- For in-place insertion-sort
  - We keep sorted the initial portion of the sequence
  - We can use *swaps* instead of modifying the sequence



# Heaps





# Recall Priority Queue ADT

- A priority queue stores a collection of items
- Each item is a pair (key, value)
- Main methods of the Priority Queue ADT
  - `add(k, x)`  
inserts an item with key `k` and value `x`
  - `remove_min()`  
removes and returns the item with smallest key
- Additional methods
  - `min()`  
returns, but does not remove, an item with smallest key
  - `len()`, `is_empty()`
- Applications:
  - Standby flyers
  - Auctions
  - Stock market

# Recall PQ Sorting



- We use a priority queue
  - Insert the elements with a series of `add` operations
  - Remove the elements in sorted order with a series of `remove_min` operations
- The running time depends on the priority queue implementation:
  - Unsorted sequence gives selection-sort:  $O(n^2)$  time
  - Sorted sequence gives insertion-sort:  $O(n^2)$  time
- Can we do better?

## Algorithm *PQ-Sort*(*S*, *C*)

**Input** sequence *S*, comparator *C*  
for the elements of *S*

**Output** sequence *S* sorted in  
increasing order according to *C*

*P* ← priority queue with  
comparator *C*

**while**  $\neg S.is\_empty()$

*e* ← *S.remove*(*S.first*())

*P.add*(*e*, *e*)

**while**  $\neg P.is\_empty()$

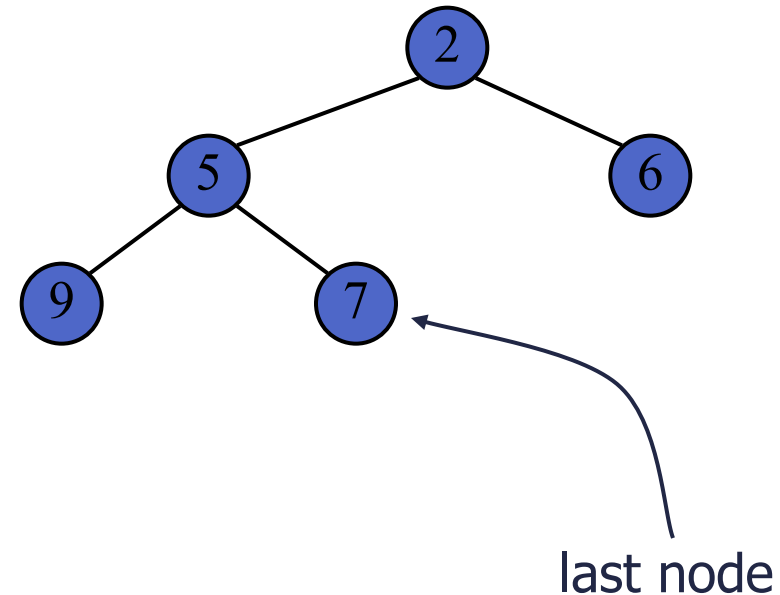
*e* ← *P.remove\_min*().key()

*S.add\_last*(*e*)

# Heaps

- A heap is a binary tree storing keys at its nodes and satisfying the following properties:
- **Heap-Order:** for every internal node  $v$  other than the root,  $key(v) \geq key(parent(v))$
- **Complete Binary Tree:** let  $h$  be the height of the heap
  - for  $i = 0, \dots, h - 1$ , there are  $2^i$  nodes of depth  $i$  (**fully filled**)
  - at depth  $h$ , the remaining nodes are residing in the leftmost possible positions at that level.

- The last node of a heap is the rightmost node of maximum depth



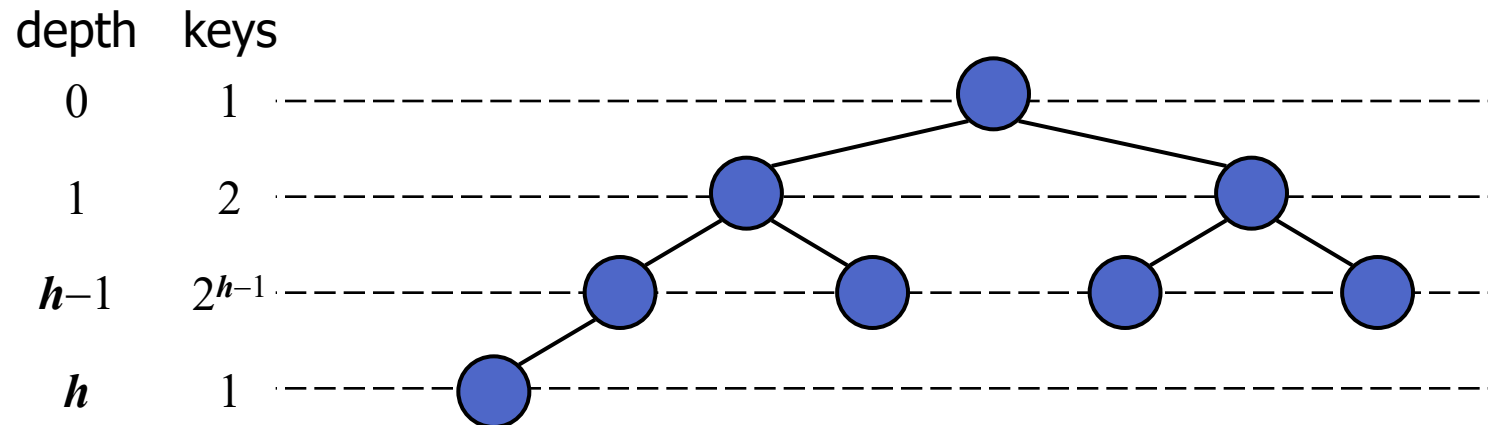
# Height of a Heap



- Theorem: A heap storing  $n$  keys has height  $O(\log n)$

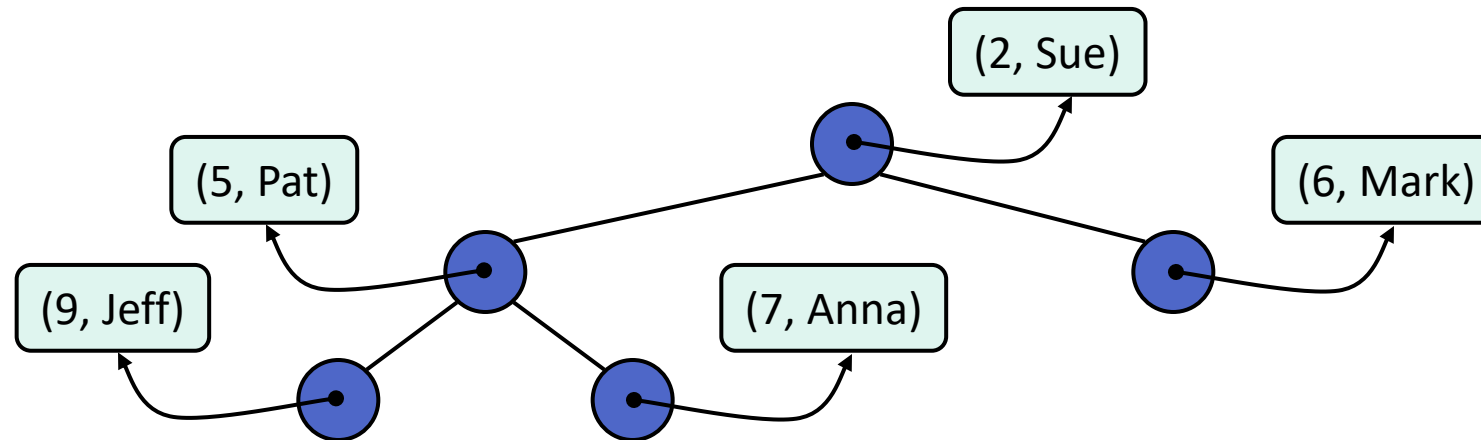
Proof: (we apply the complete binary tree property)

- Let  $h$  be the height of a heap storing  $n$  keys
- Since there are  $2^i$  keys at depth  $i = 0, \dots, h-1$  and at least one key at depth  $h$ , we have  $n \geq 1 + 2 + 4 + \dots + 2^{h-1} + 1$
- Thus,  $n \geq 2^h$ , i.e.,  $h \leq \log n$



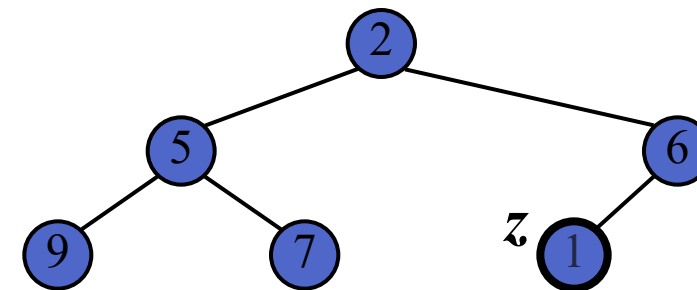
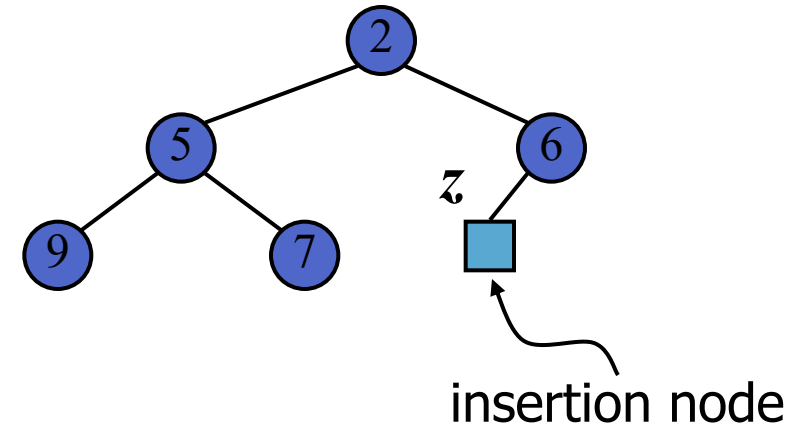
# Heaps and Priority Queues

- We can use a heap to implement a priority queue
- We store a (key, element) item at each node
- We keep track of the **position of the last node**



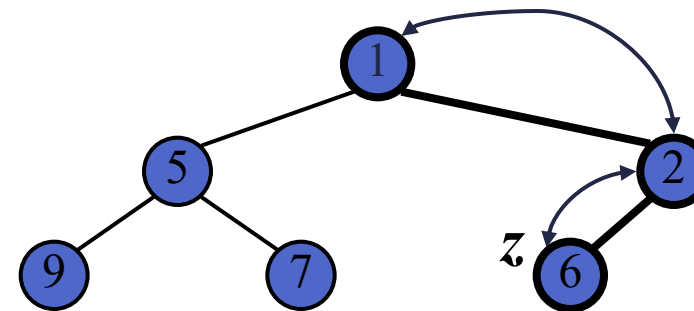
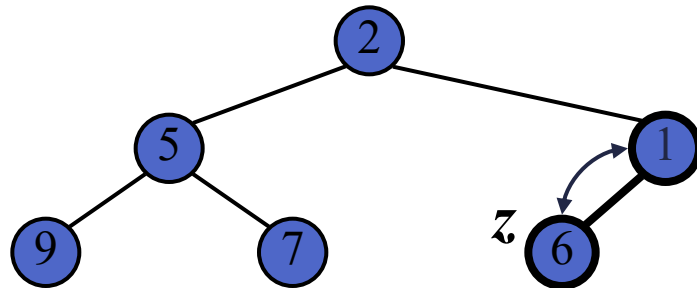
# Insertion into a Heap

- Method add of the priority queue ADT corresponds to the insertion of a key  $k$  to the heap
- The insertion algorithm consists of three steps
  - Find the insertion node  $z$  (the new last node)
  - Store  $k$  at  $z$
  - Restore the heap-order property (discussed next)

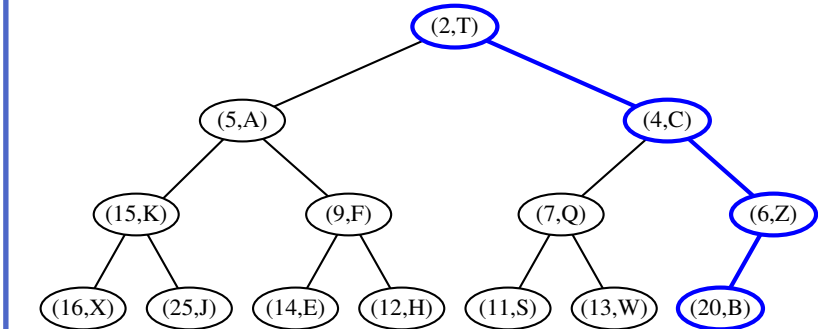
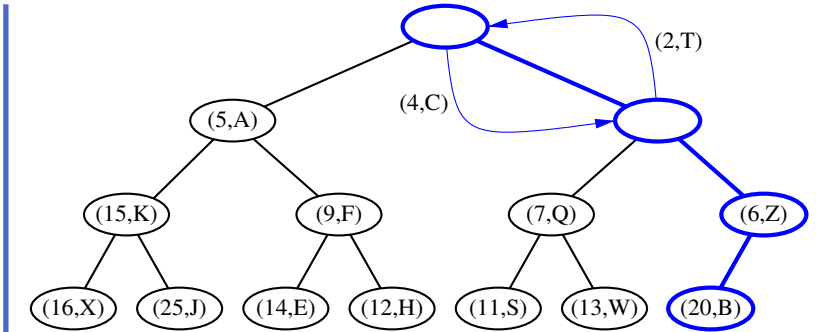
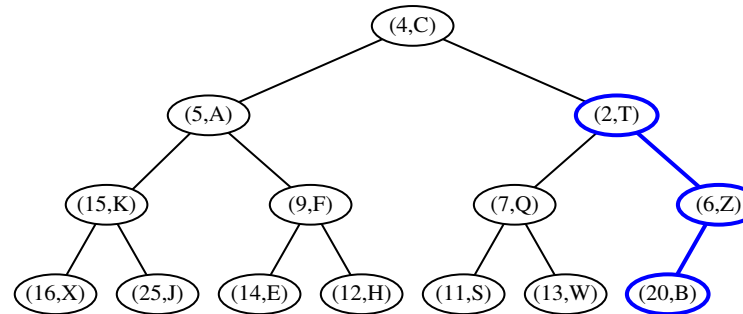
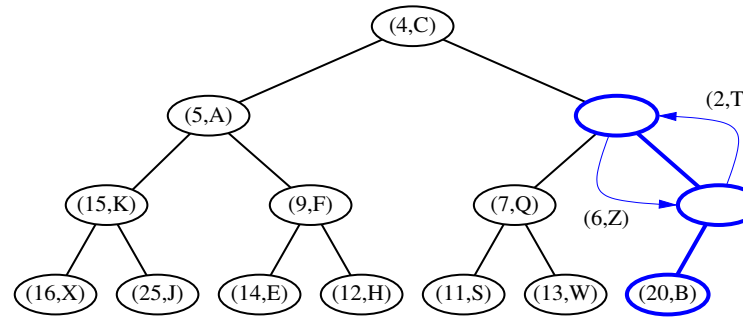
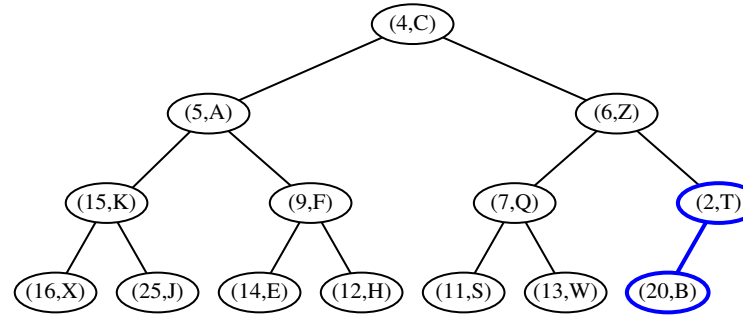
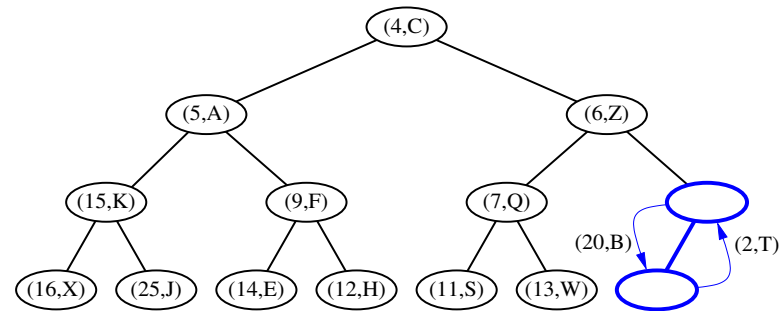
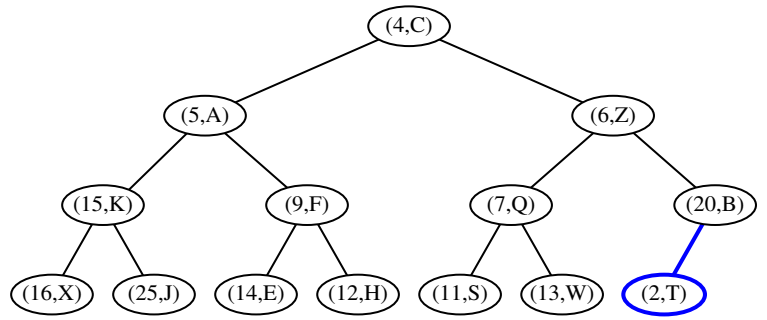
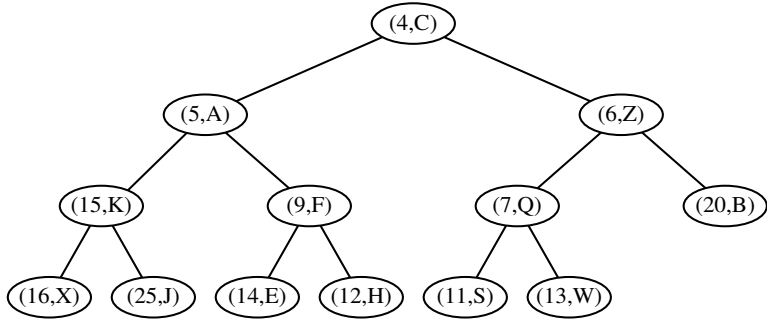


# Upheap

- After the insertion of a new key  $k$ , the heap-order property may be violated
- Algorithm upheap restores the heap-order property by swapping  $k$  along an upward path from the insertion node
- Upheap terminates when the key  $k$  reaches the root or a node whose parent has a key smaller than or equal to  $k$
- Since a heap has height  $O(\log n)$ , upheap runs in  $O(\log n)$  time



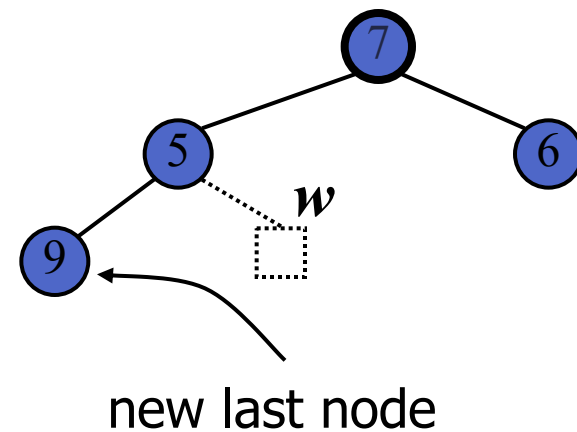
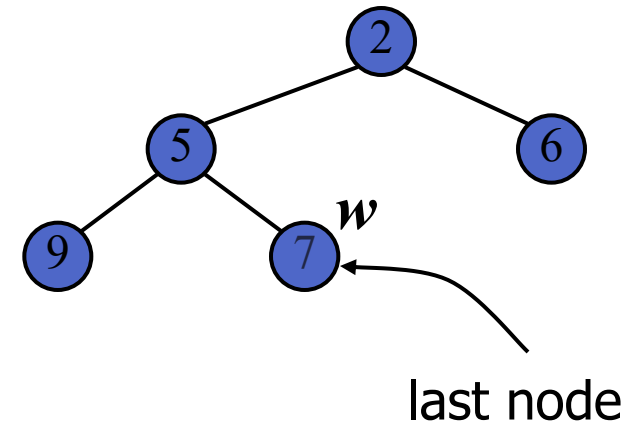
# Upheap example





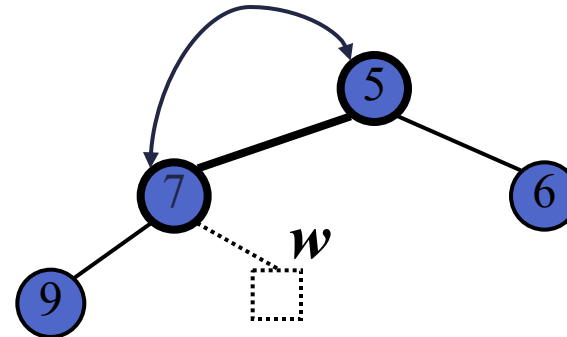
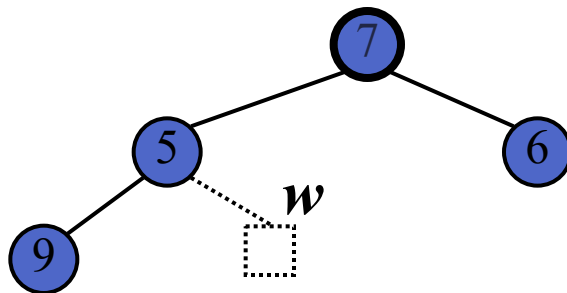
# Removal from a Heap

- Method `remove_min` of the priority queue ADT corresponds to the removal of the root key from the heap
- The removal algorithm consists of three steps
  - Replace the root key with the key of the last node  $w$
  - Remove  $w$
  - Restore the heap-order property (discussed next)

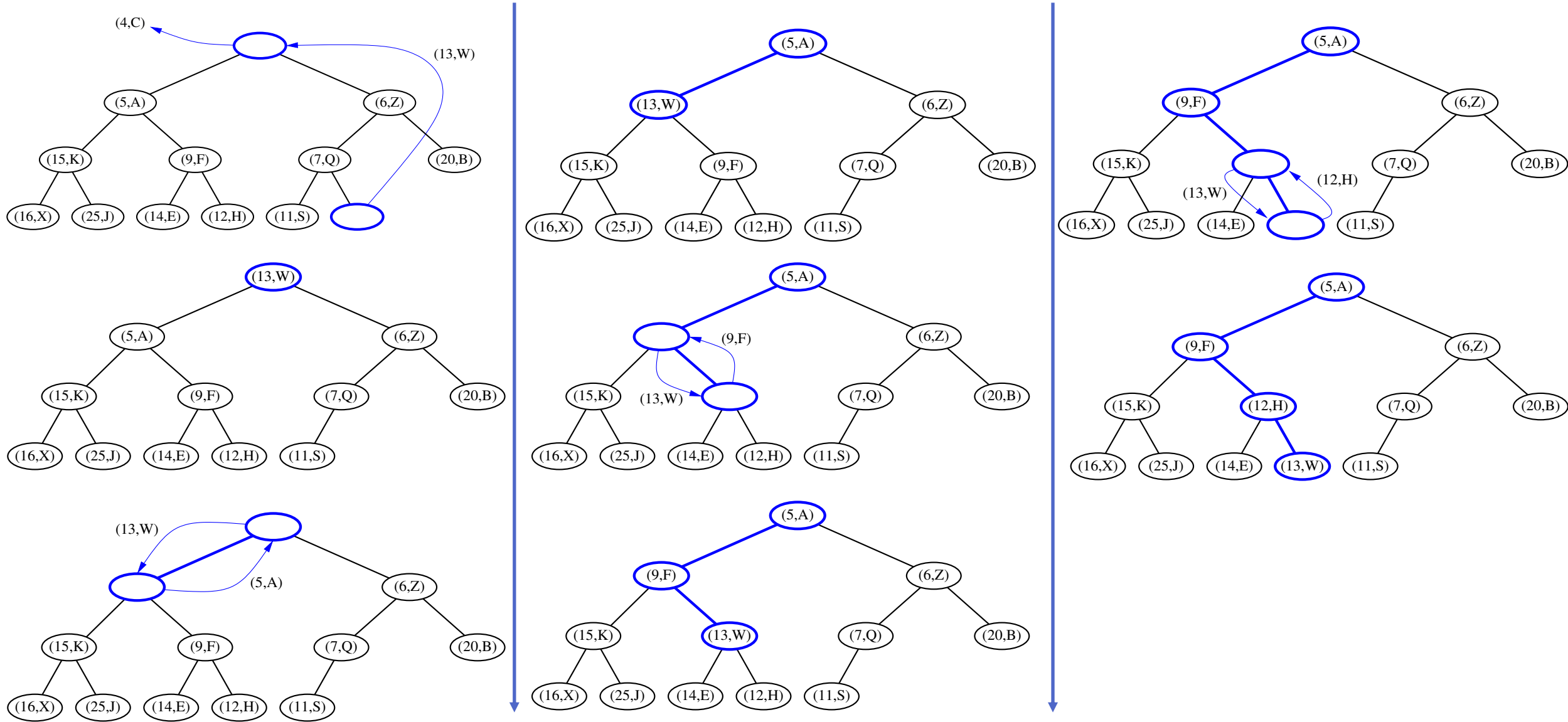


# Downheap

- After replacing the root key with the key  $k$  of the last node, the heap-order property may be violated
- Algorithm downheap restores the heap-order property by swapping key  $k$  along a downward path from the root
  - For swapping, select a child node with the minimal key (=higher priority)
- Downheap terminates when key  $k$  reaches a leaf or a node whose children have keys greater than or equal to  $k$
- Since a heap has height  $O(\log n)$ , downheap runs in  $O(\log n)$  time

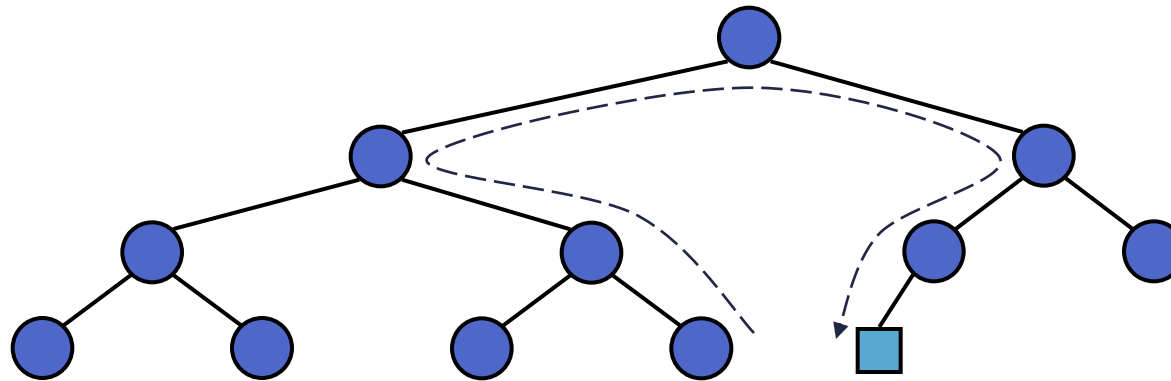


# Downheap example

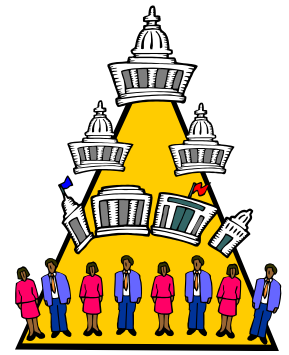


# Updating the Last Node

- The insertion node can be found by traversing a path of  $O(\log n)$  nodes
  - Go up until a left child or the root is reached
    - If a left child is reached, go to the right child
    - If a root is reached, stay there.
  - Go down left until a leaf is reached
- Similar algorithm for updating the last node after a removal



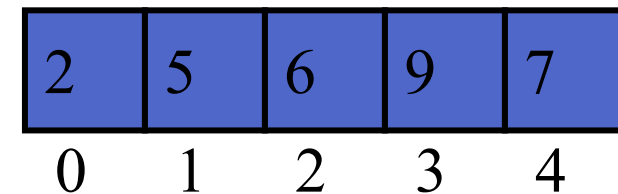
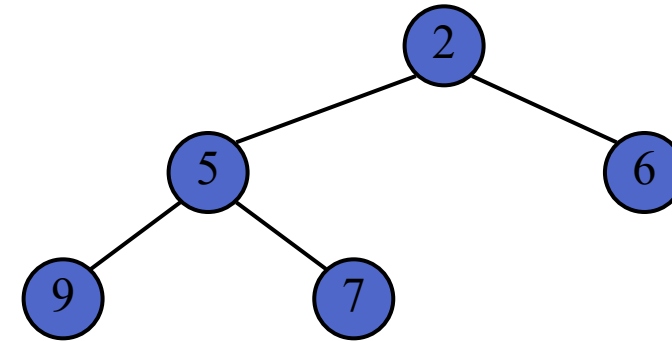
# Heap-Sort



- Consider a priority queue with  $n$  items implemented by means of a heap
  - the space used is  $O(n)$
  - methods `add` and `remove_min` take  $O(\log n)$  time
  - methods `len`, `is_empty`, and `min` take time  $O(1)$  time
- Using a heap-based priority queue, we can sort a sequence of  $n$  elements in  $O(n \log n)$  time
- The resulting algorithm is called heap-sort
- Heap-sort is much faster than quadratic sorting algorithms, such as insertion-sort and selection-sort

# Array-based Heap Implementation

- We can represent a heap with  $n$  keys by means of an array of length  $n$
- For the node at rank  $i$ 
  - the left child is at rank  $2i + 1$
  - the right child is at rank  $2i + 2$
- Links between nodes are not explicitly stored
- Operation add corresponds to inserting at rank  $n + 1$
- Operation remove\_min corresponds to removing at rank  $n$
- Yields in-place heap-sort

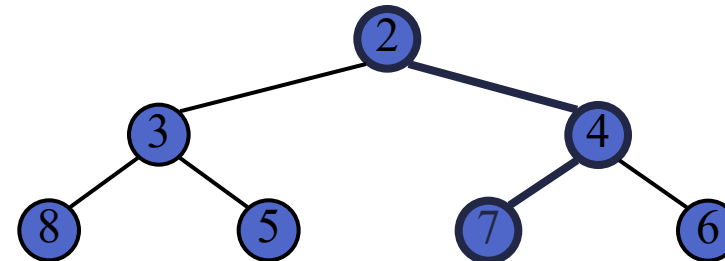
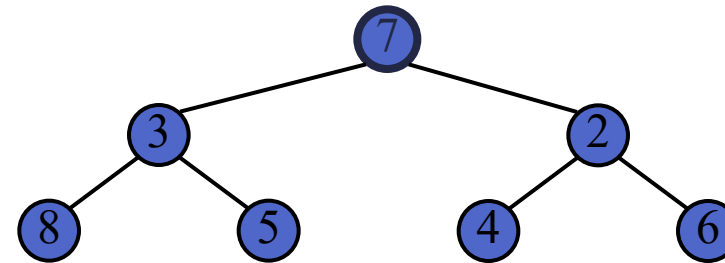
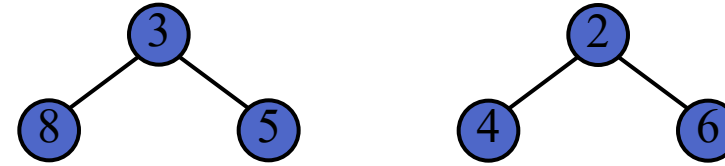


# Python Heap Implementation

```
1 class HeapPriorityQueue(PriorityQueueBase): # base class defines _Item
2     """A min-oriented priority queue implemented with a binary heap."""
3     #----- nonpublic behaviors -----
4     def _parent(self, j):
5         return (j-1) // 2
6
7     def _left(self, j):
8         return 2*j + 1
9
10    def _right(self, j):
11        return 2*j + 2
12
13    def _has_left(self, j):
14        return self._left(j) < len(self._data) # index beyond end of list?
15
16    def _has_right(self, j):
17        return self._right(j) < len(self._data) # index beyond end of list?
18
19    def _swap(self, i, j):
20        """Swap the elements at indices i and j of array."""
21        self._data[i], self._data[j] = self._data[j], self._data[i]
22
23    def _upheap(self, j):
24        parent = self._parent(j)
25        if j > 0 and self._data[j] < self._data[parent]:
26            self._swap(j, parent)
27            self._upheap(parent) # recur at position of parent
28
29    def _downheap(self, j):
30        if self._has_left(j):
31            left = self._left(j)
32            small_child = left # although right may be smaller
33            if self._has_right(j):
34                right = self._right(j)
35                if self._data[right] < self._data[left]:
36                    small_child = right
37            if self._data[small_child] < self._data[j]:
38                self._swap(j, small_child)
39                self._downheap(small_child) # recur at position of small child
40
41    #----- public behaviors -----
42    def __init__(self):
43        """Create a new empty Priority Queue."""
44        self._data = []
45
46    def __len__(self):
47        """Return the number of items in the priority queue."""
48        return len(self._data)
49
50    def add(self, key, value):
51        """Add a key-value pair to the priority queue."""
52        self._data.append(self._Item(key, value))
53        self._upheap(len(self._data) - 1) # upheap newly added position
54
55    def min(self):
56        """Return but do not remove (k,v) tuple with minimum key.
57
58        Raise Empty exception if empty.
59        """
60        if self.is_empty():
61            raise Empty('Priority queue is empty.')
62        item = self._data[0]
63        return (item._key, item._value)
64
65    def remove_min(self):
66        """Remove and return (k,v) tuple with minimum key.
67
68        Raise Empty exception if empty.
69        """
70        if self.is_empty():
71            raise Empty('Priority queue is empty.')
72        self._swap(0, len(self._data) - 1) # put minimum item at the end
73        item = self._data.pop() # and remove it from the list;
74        self._downheap(0) # then fix new root
75        return (item._key, item._value)
```

# Merging Two Heaps

- We are given two heaps and a key  $k$
- We create a new heap with the root node storing  $k$  and with the two heaps as subtrees
- We perform downheap to restore the heap-order property

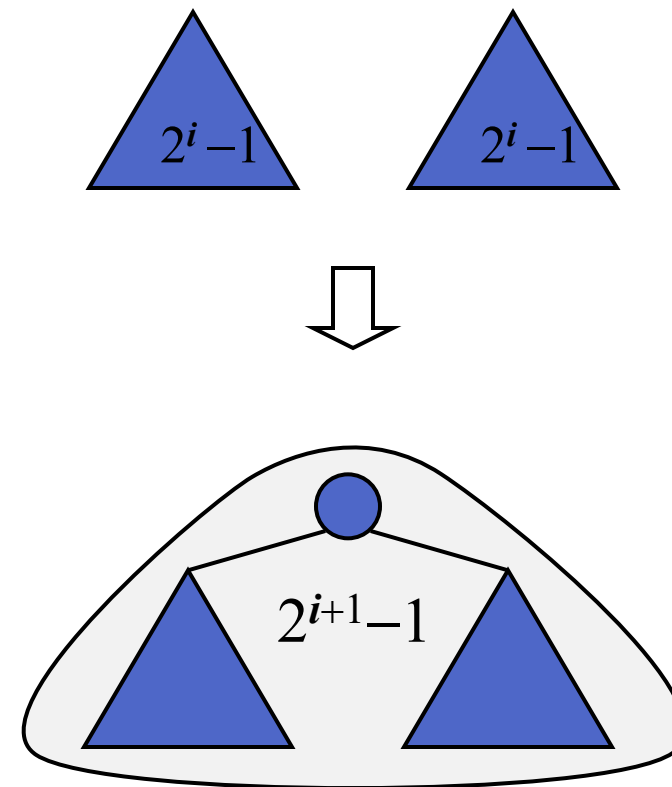




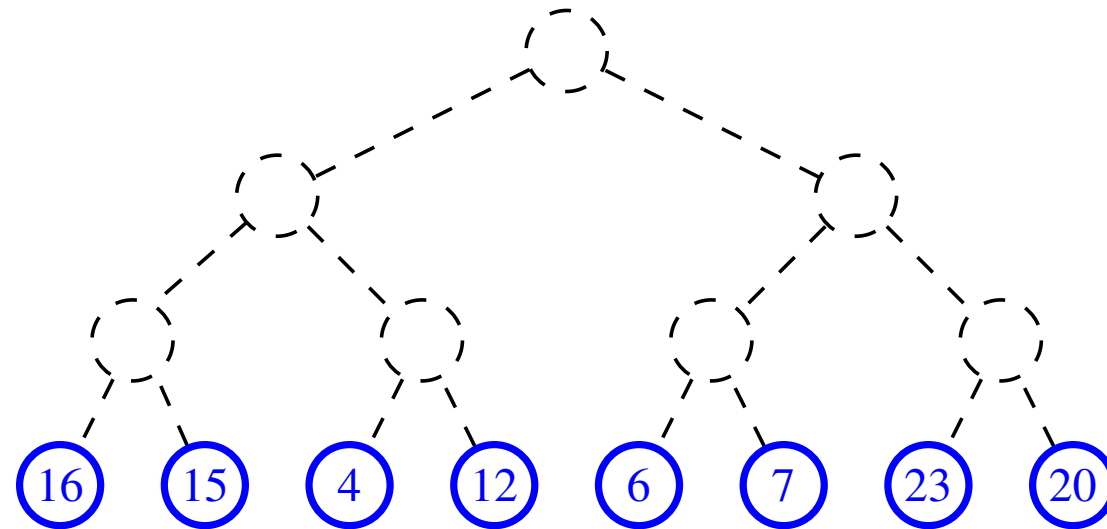
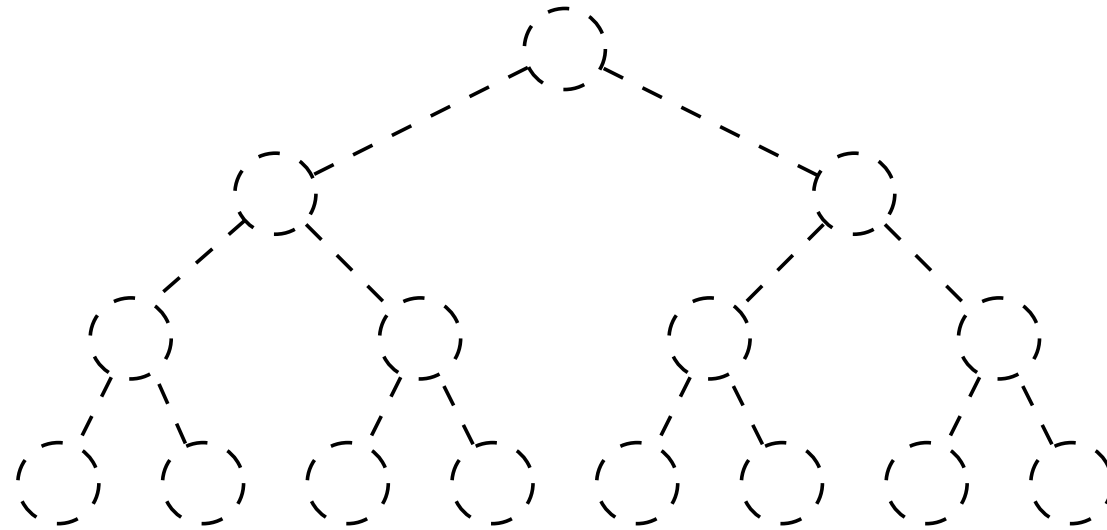
# Bottom-up Heap Construction



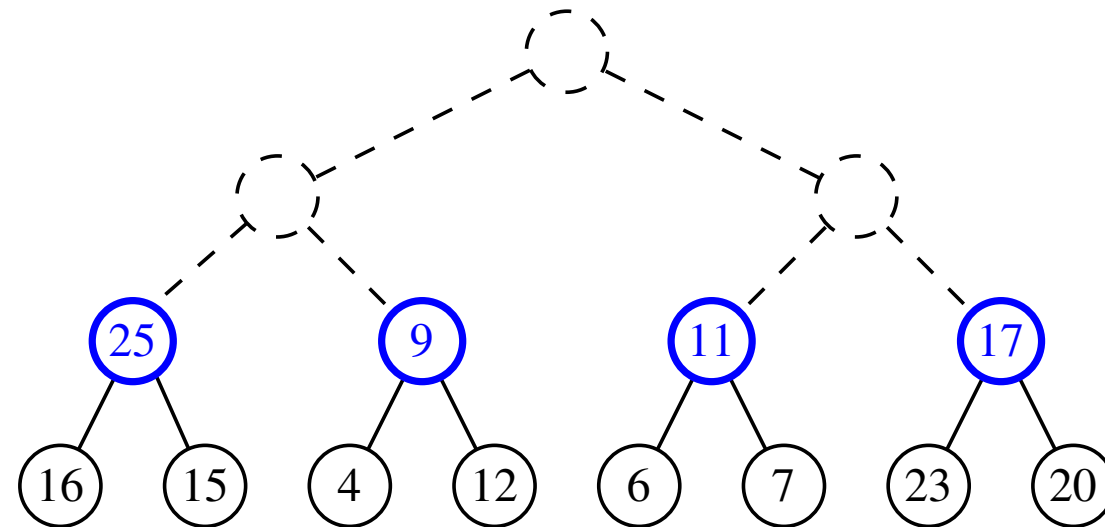
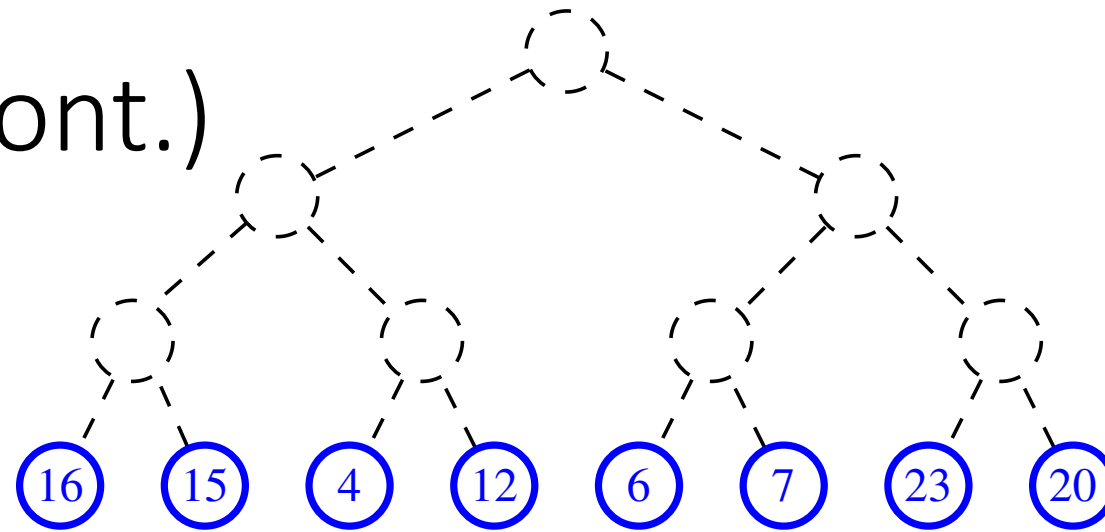
- We can construct a heap storing  $n$  given keys in using a bottom-up construction with  $\log n$  phases
- In phase  $i$ , pairs of heaps with  $2^i - 1$  keys are merged into heaps with  $2^{i+1} - 1$  keys



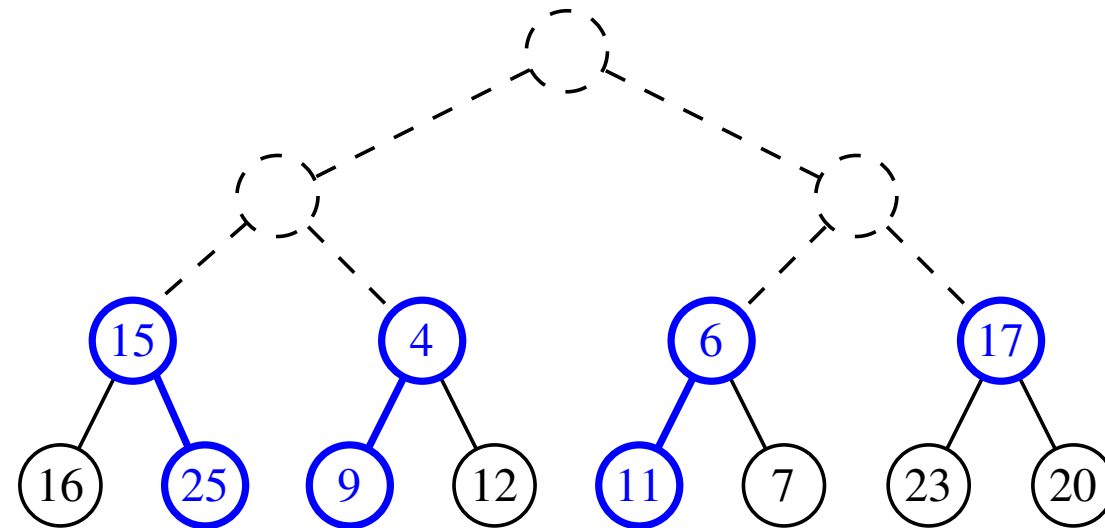
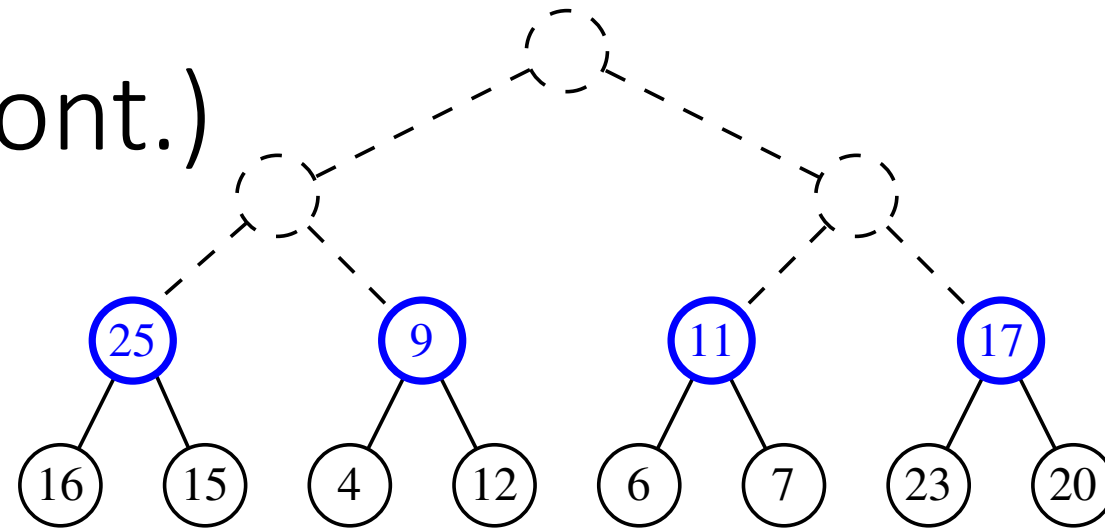
# Example



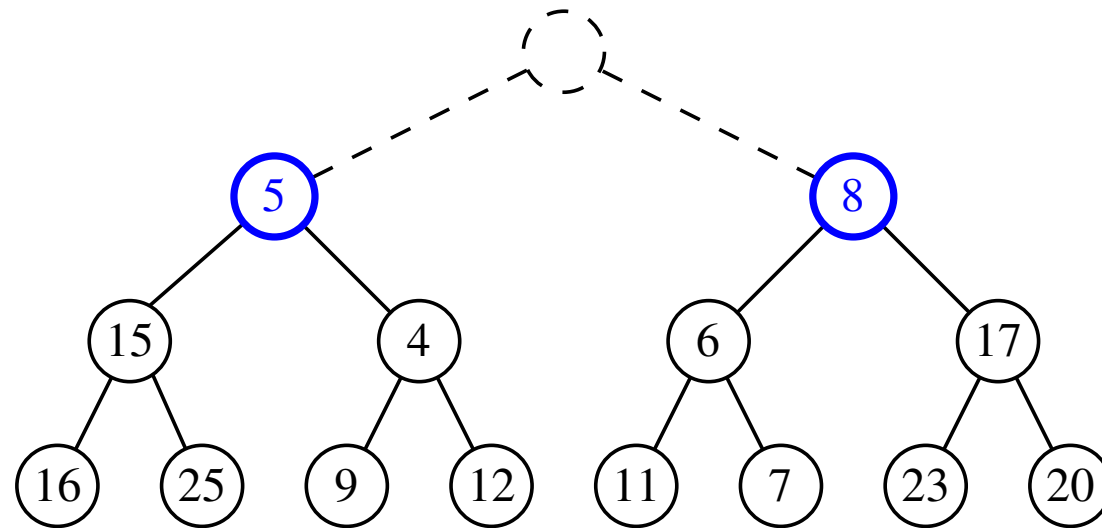
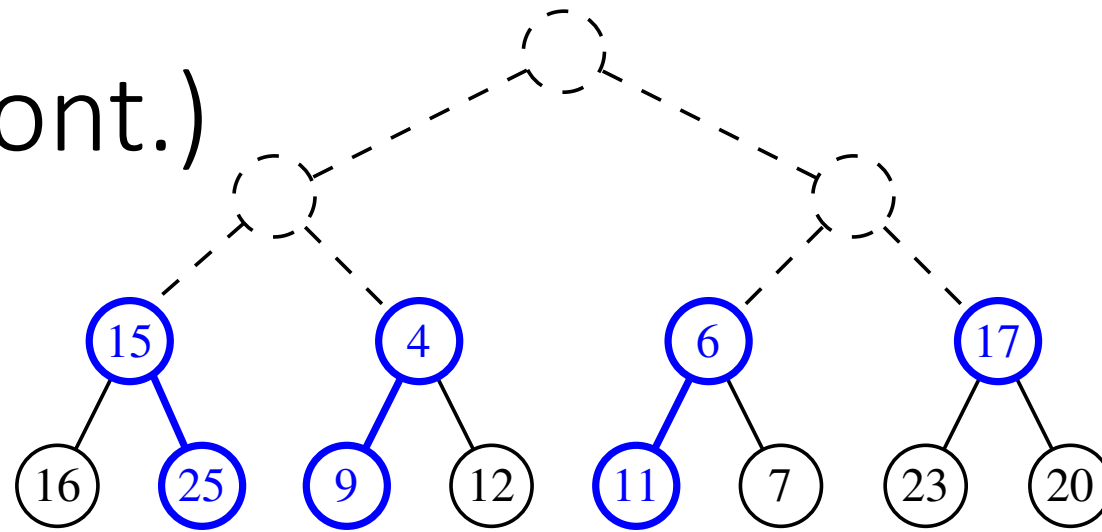
## Example (cont.)



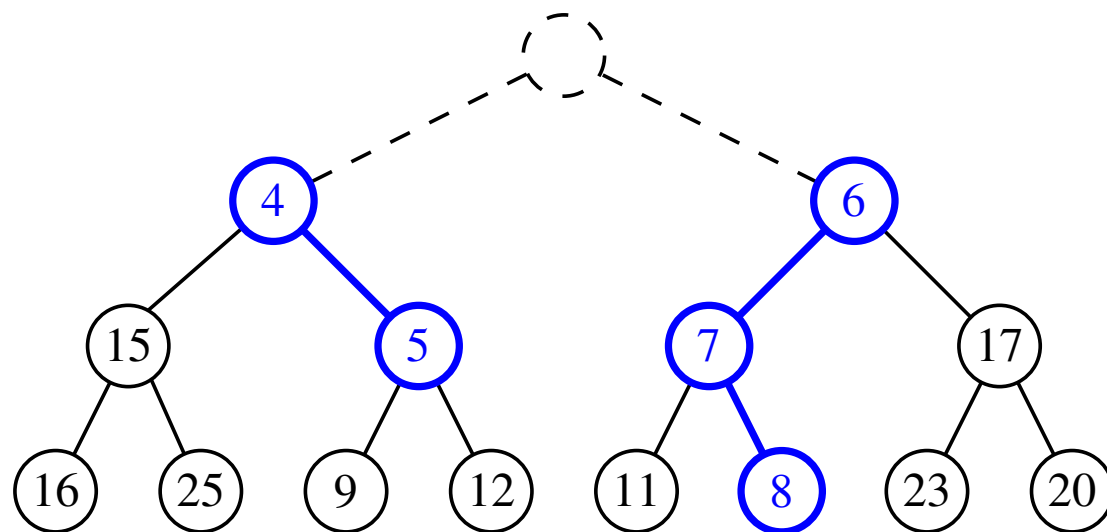
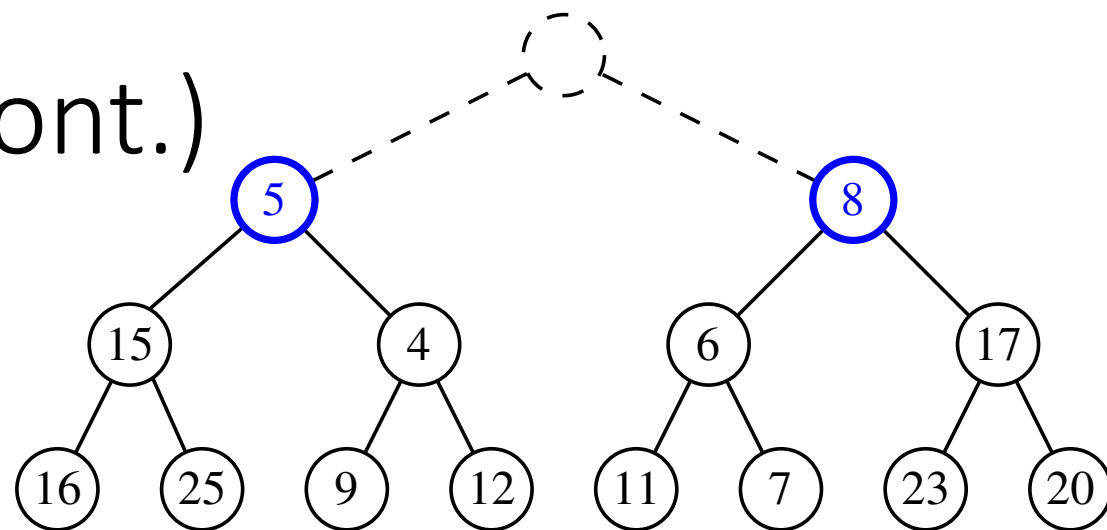
## Example (cont.)



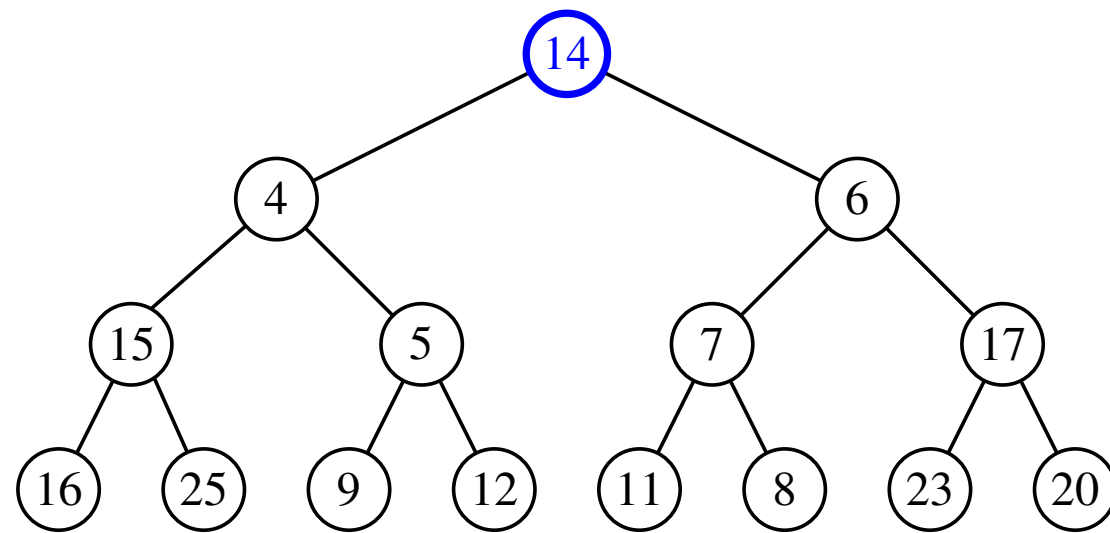
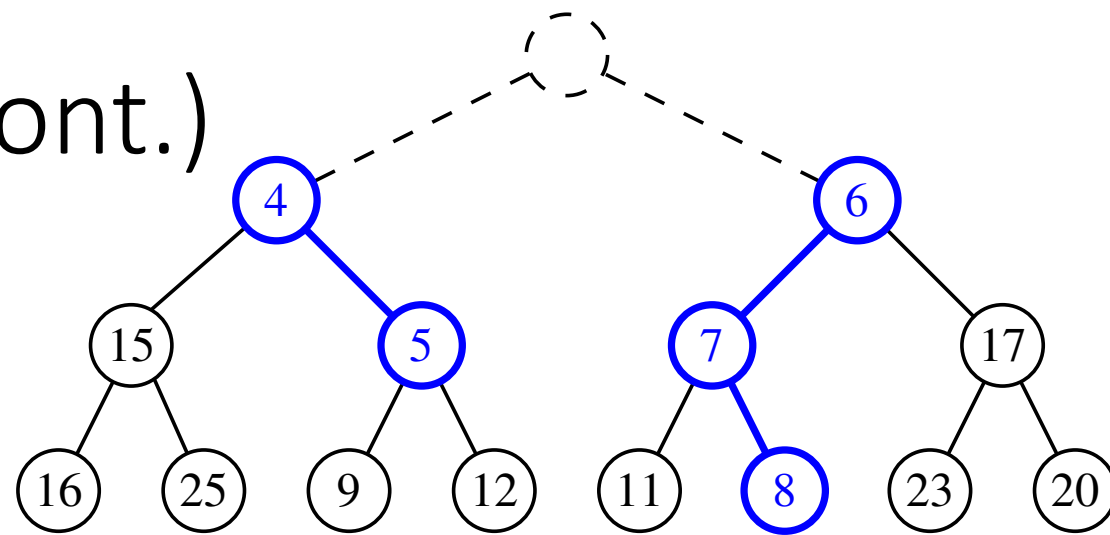
## Example (cont.)



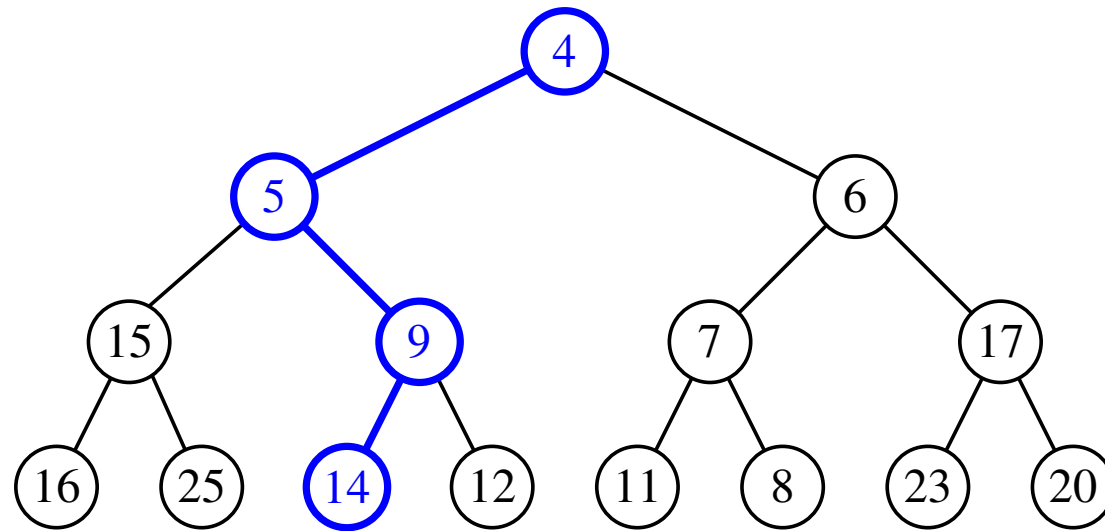
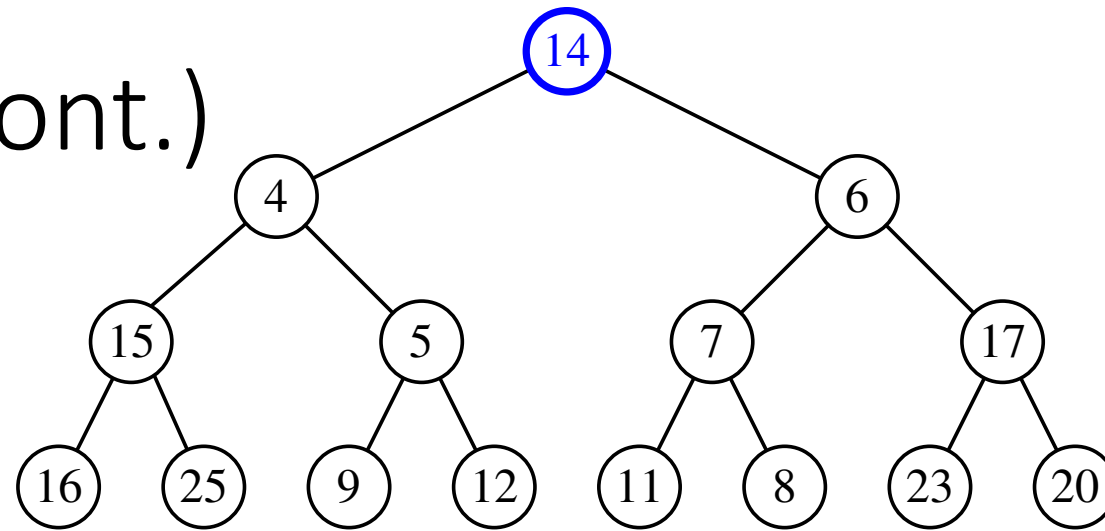
# Example (cont.)



# Example (cont.)



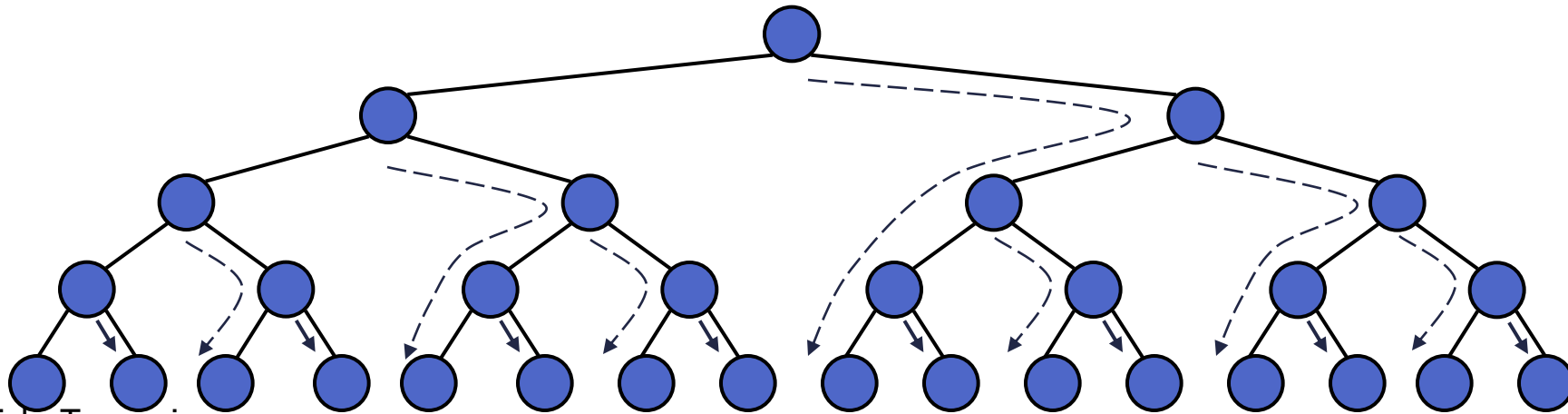
Example (cont.)





# Analysis of Heap Construction

- We visualize the worst-case time of a downheap with a proxy path that goes first right and then repeatedly goes left until the bottom of the heap (this path may differ from the actual downheap path)
- Since each node is traversed by at most two proxy paths, the total number of nodes of the proxy paths is  $O(n)$
- Thus, bottom-up heap construction runs in  $O(n)$  time
- Bottom-up heap construction is faster than  $n$  successive insertions and speeds up the first phase of heap-sort



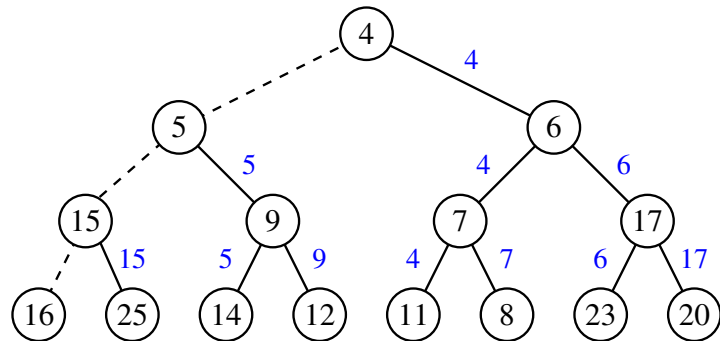
# Exercise (optional) (textbook p383)

## **Asymptotic Analysis of Bottom-Up Heap Construction**

Bottom-up heap construction is asymptotically faster than incrementally inserting  $n$  keys into an initially empty heap. Intuitively, we are performing a single down-heap operation at each position in the tree, rather than a single up-heap operation from each. Since more nodes are closer to the bottom of a tree than the top, the sum of the downward paths is linear, as shown in the following proposition.

**Proposition 9.3:** *Bottom-up construction of a heap with  $n$  entries takes  $O(n)$  time, assuming two keys can be compared in  $O(1)$  time.*

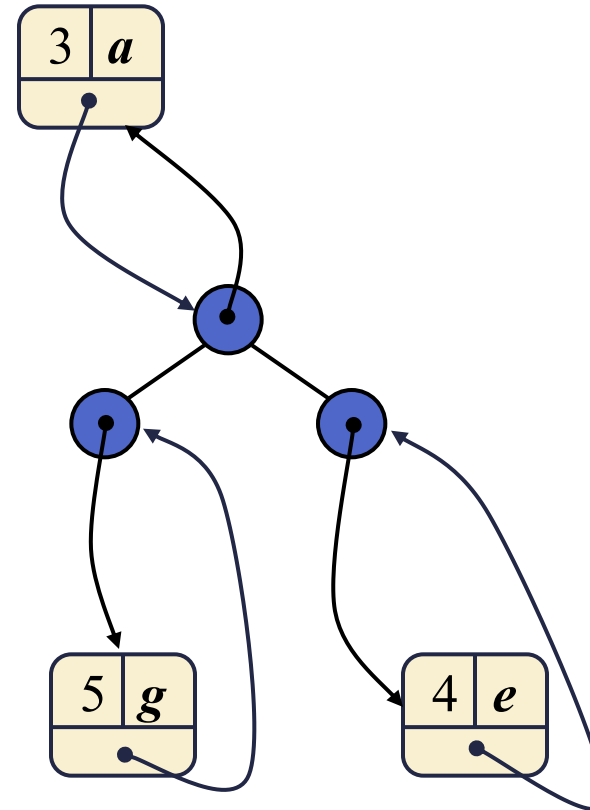
# Exercise : answer



**Justification:** The primary cost of the construction is due to the down-heap steps performed at each nonleaf position. Let  $\pi_v$  denote the path of  $T$  from nonleaf node  $v$  to its “inorder successor” leaf, that is, the path that starts at  $v$ , goes to the right child of  $v$ , and then goes down leftward until it reaches a leaf. Although,  $\pi_v$  is not necessarily the path followed by the down-heap bubbling step from  $v$ , the length  $\|\pi_v\|$  (its number of edges) is proportional to the height of the subtree rooted at  $v$ , and thus a bound on the complexity of the down-heap operation at  $v$ . We can bound the total running time of the bottom-up heap construction algorithm based on the sum of the sizes of paths,  $\sum_v \|\pi_v\|$ . For intuition, Figure 9.6 illustrates the justification “visually,” marking each edge with the label of the nonleaf node  $v$  whose path  $\pi_v$  contains that edge.

We claim that the paths  $\pi_v$  for all nonleaf  $v$  are edge-disjoint, and thus the sum of the path lengths is bounded by the number of total edges in the tree, hence  $O(n)$ . To show this, we consider what we term “right-leaning” and “left-leaning” edges (i.e., those going from a parent to a right, respectively left, child). A particular right-leaning edge  $e$  can only be part of the path  $\pi_v$  for node  $v$  that is the parent in the relationship represented by  $e$ . Left-leaning edges can be partitioned by considering the leaf that is reached if continuing down leftward until reaching a leaf. Each nonleaf node only uses left-leaning edges in the group leading to that nonleaf node’s inorder successor. Since each nonleaf node must have a different inorder successor, no two such paths can contain the same left-leaning edge. We conclude that the bottom-up construction of heap  $T$  takes  $O(n)$  time. ■

# Adaptable Priority Queues



# Items and Priority Queues

- An **item** stores a (key, value) pair
- Item fields:
  - **\_key**: the key associated with this item
  - **\_value**: the value paired with the key associated with this item
- Priority Queue ADT:
  - **add(k, x)**  
inserts an item with key k and value x
  - **remove\_min()**  
removes and returns the item with smallest key
  - **min()**  
returns, but does not remove, an item with smallest key
  - **len(P), is\_empty()**



# Example

- Online trading system where orders to purchase and sell a stock are stored in two priority queues (one for sell orders and one for buy orders) as  $(p,s)$  entries:
  - The key,  $p$ , of an order is the price
  - The value,  $s$ , for an entry is the number of shares
  - A buy order  $(p,s)$  is executed when a sell order  $(p',s')$  with price  $p' \leq p$  is added (the execution is complete if  $s' \geq s$ )
  - A sell order  $(p,s)$  is executed when a buy order  $(p',s')$  with price  $p' \geq p$  is added (the execution is complete if  $s' \geq s$ )
- What if someone wishes to cancel their order before it executes?
- What if someone wishes to update the price or number of shares for their order?

# Methods of the Adaptable Priority Queue ADT

- **remove(loc)**: Remove from P and return item e for locator loc.
- **update(loc,k,v)**: Replace the key-value pair for locator, loc, with (k,v).



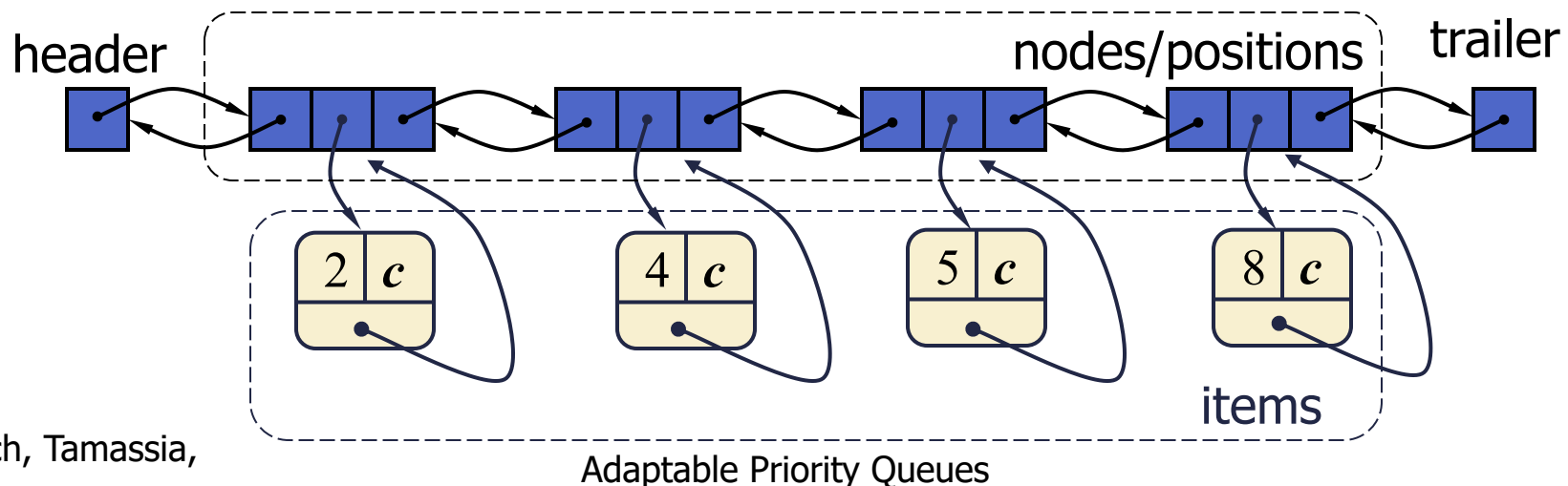
# Locators

- A locator-aware item identifies and tracks the location of its (key, value) object within a data structure
- Intuitive notion:
  - Coat claim check
  - Valet claim ticket
  - Reservation number
- Main idea:
  - Since items are created and returned from the data structure itself, it can return location-aware items, thereby making future updates easier



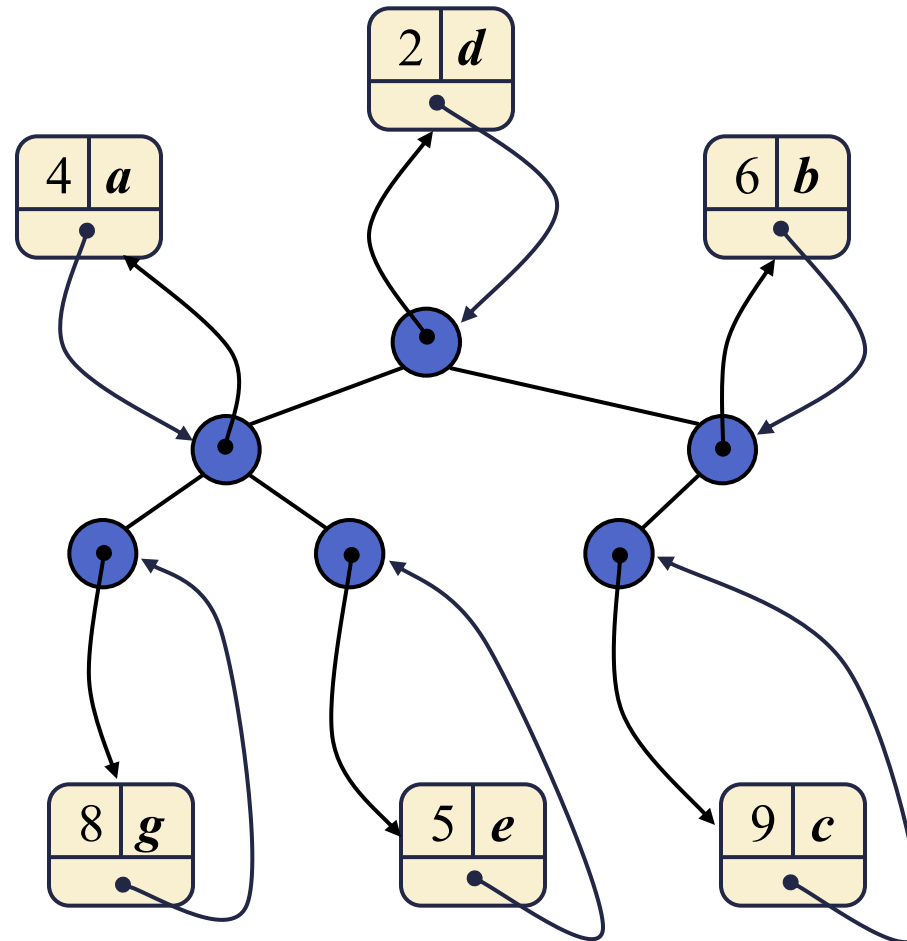
# List Implementation

- A location-aware list item is an object storing
  - key
  - value
  - position (or rank) of the item in the list
- In turn, the position (or array cell) stores the entry
- Back pointers (or ranks) are updated during swaps



# Heap Implementation

- A location-aware heap item is an object storing
  - key
  - value
  - position of the item in the underlying heap
- In turn, each heap position stores an item
- Back pointers are updated during item swaps



# Performance

- Improved times thanks to location-aware items are highlighted in red

Method	Unsorted List	Sorted List	Heap
len, is_empty	$O(1)$	$O(1)$	$O(1)$
add	$O(1)$	$O(n)$	$O(\log n)$
min	$O(n)$	$O(1)$	$O(1)$
remove_min	$O(n)$	$O(1)$	$O(\log n)$
remove	<b><math>O(1)</math></b>	<b><math>O(1)</math></b>	<b><math>O(\log n)</math></b>
update	<b><math>O(1)</math></b>	$O(n)$	<b><math>O(\log n)</math></b>