

# DATA STRUCTURES AND ALGORITHMS

## LECTURE 7

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## In Lecture 6...

- Linked Lists on Arrays
- Stack, Queue, Deque

# Today

- 1 Binary Heap
- 2 Binomial heap

# ADT Priority Queue

- The ADT Priority Queue is a container in which each element has an associated *priority* (of type *TPriority*).
- In a Priority Queue access to the elements is restricted: we can access only the element with the highest priority.
- Because of this restricted access, we say that the Priority Queue works based on a **HPF - Highest Priority First** policy.

# Priority Queue - Representation

- What data structures can be used to implement a priority queue?
  - Dynamic Array
  - Linked List
  - (Binary) Heap - will be discussed later

# Priority Queue - Representation

- If the representation is a Dynamic Array or a Linked List we have to decide how we store the elements in the array/list:
  - we can keep the elements ordered by their priorities
    - Where would you put the element with the highest priority?
  - we can keep the elements in the order in which they were inserted

# Priority Queue - Representation

- Complexity of the main operations for the two representation options:

Operation	Sorted	Non-sorted
push	$O(n)$	$\Theta(1)$
pop	$\Theta(1)$	$\Theta(n)$
top	$\Theta(1)$	$\Theta(n)$

- What happens if we keep in a separate field the element with the highest priority?

# Binary Heap

- A binary heap is a data structure that can be used as an efficient representation for Priority Queues.
- A binary heap is a kind of hybrid between a dynamic array and a binary tree.
- The elements of the heap are actually stored in the dynamic array, but the array is visualized as a binary tree.



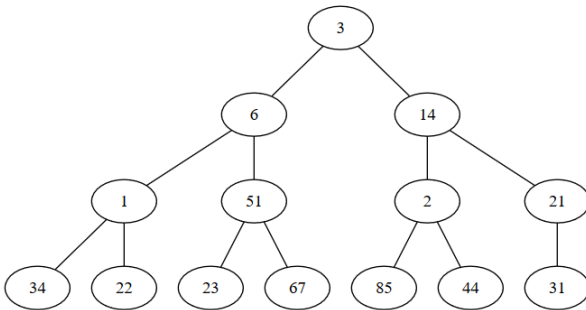
# Binary Heap

- Assume that we have the following array (upper row contains positions, lower row contains elements):

1	2	3	4	5	6	7	8	9	10	11	12	13	14
3	6	14	1	51	2	21	34	22	23	67	85	44	31

# Binary Heap

- We can visualize this array as a binary tree, where the root is the first element of the array, its children are the next two elements, and so on. Each node has exactly 2 children, except for the last two rows, but there the children of the nodes are completed from left to right.



# Binary Heap

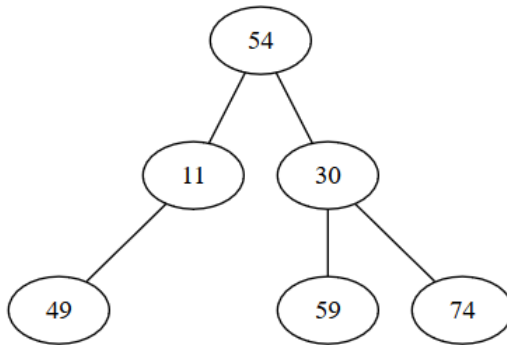
- If the elements of the array are:  $a_1, a_2, a_3, \dots, a_n$ , we know that:
  - $a_1$  is the root of the heap
  - for an element from position  $i$ , its children are on positions  $2 * i$  and  $2 * i + 1$  (if  $2 * i$  and  $2 * i + 1$  is less than or equal to  $n$ )
  - for an element from position  $i$  ( $i > 1$ ), the parent of the element is on position  $[i/2]$  (integer part of  $i/2$ )

# Binary Heap

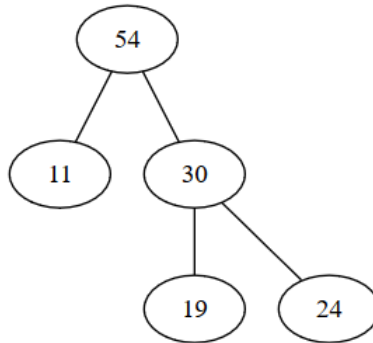
- A *binary heap* is an array that can be visualized as a binary tree having a *heap structure* and a *heap property*.
  - *Heap structure*: in the binary tree every node has exactly 2 children, except for the last two levels, where children are completed from left to right.
  - *Heap property*:  $a_i \geq a_{2*i}$  (if  $2 * i \leq n$ ) and  $a_i \geq a_{2*i+1}$  (if  $2 * i + 1 \leq n$ )
  - The  $\geq$  relation between a node and both its descendants can be generalized (other relations can be used as well).

# Binary Heap - Examples I

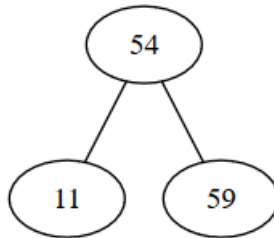
- Are the following binary trees heaps? If yes, specify the relation between a node and its children. If not, specify if the problem is with the structure, the property, or both.



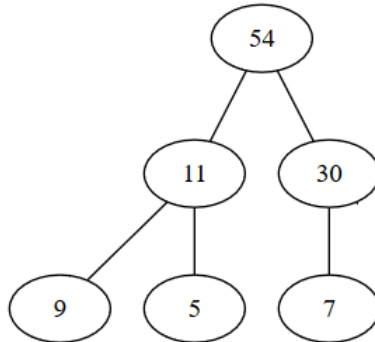
# Binary Heap - Examples II



# Binary Heap - Examples III

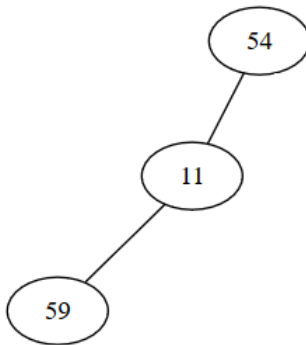


# Binary Heap - Examples IV

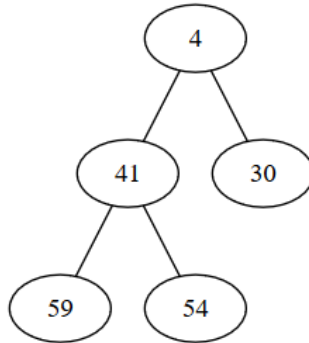




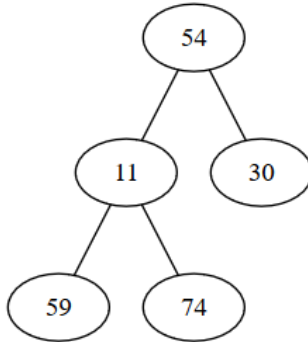
# Binary Heap - Examples V



# Binary Heap - Examples VI



# Binary Heap - Examples VII



# Binary Heap - Examples VIII

- Are the following arrays valid heaps? If not, transform them into a valid heap by swapping two elements.
  - 1 [70, 10, 50, 7, 1, 33, 3, 8]
  - 2 [1, 2, 4, 8, 16, 32, 64, 65, 10]
  - 3 [10, 12, 100, 60, 13, 102, 101, 80, 90, 14, 15, 16]

# Binary Heap - Notes

- If we use the  $\geq$  relation, we will have a *MAX-HEAP*. Do you know why?

# Binary Heap - Notes

- If we use the  $\geq$  relation, we will have a *MAX-HEAP*. Do you know why?
- If we use the  $\leq$  relation, we will have a *MIN-HEAP*. Do you know why?

# Binary Heap - Notes

- If we use the  $\geq$  relation, we will have a *MAX-HEAP*. Do you know why?
- If we use the  $\leq$  relation, we will have a *MIN-HEAP*. Do you know why?
- The height of a heap with  $n$  elements is  $\log_2 n$ .

# Binary Heap - operations

- A heap can be used as representation for a Priority Queue and it has two specific operations:
  - add a new element in the heap (in such a way that we keep both the heap structure and the heap property).
  - remove (we always remove the root of the heap - no other element can be removed).



# Binary Heap - representation

Heap:

cap: Integer

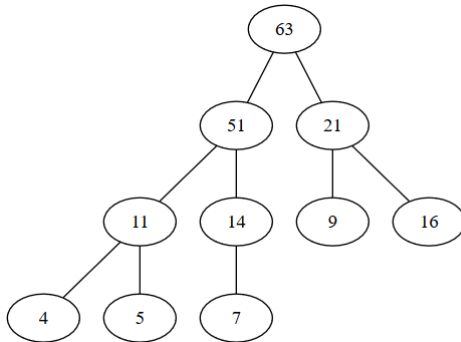
len: Integer

elems: TElem[]

- For the implementation we will assume that we have a MAX-HEAP.

# Binary Heap - Add - example

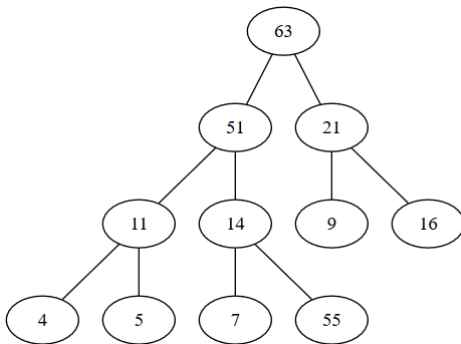
- Consider the following (MAX) heap:



- Let's add the number 55 to the heap.

# Binary Heap - Add - example

- In order to keep the *heap structure*, we will add the new node as the right child of the node 14 (and as the last element of the array in which the elements are kept).

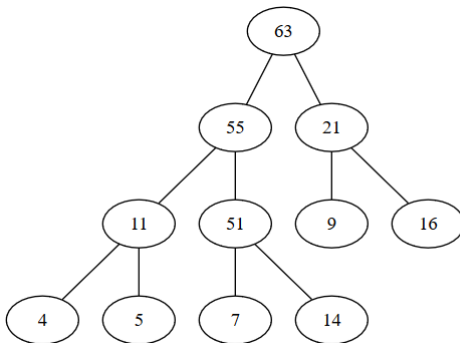


# Binary Heap - Add - example

- *Heap property* is not kept: 14 has as child node 55 (since it is a MAX-heap, each node has to be greater than or equal to its descendants).
- In order to restore the heap property, we will start a *bubble-up* process: we will keep swapping the value of the new node with the value of its parent node, until it gets to its final place. No other node from the heap is changed.

# Binary Heap - Add - example

- When *bubble-up* ends:



# Binary Heap - add

**subalgorithm** add(heap, e) **is:**

*//heap - a heap*

*//e - the element to be added*

**if** heap.len = heap.cap **then**

    @ resize

**end-if**

heap.ellems[heap.len+1]  $\leftarrow$  e

heap.len  $\leftarrow$  heap.len + 1

bubble-up(heap, heap.len)

**end-subalgorithm**

# Binary Heap - add

**subalgorithm** bubble-up (heap, p) **is:**

*//heap - a heap*

*//p - position from which we bubble the new node up*

poz  $\leftarrow$  p

elem  $\leftarrow$  heap.elems[p]

parent  $\leftarrow$  p / 2

**while** poz > 1 **and** elem > heap.elems[parent] **execute**

*//move parent down*

heap.elems[poz]  $\leftarrow$  heap.elems[parent]

poz  $\leftarrow$  parent

parent  $\leftarrow$  poz / 2

**end-while**

heap.elems[poz]  $\leftarrow$  elem

**end-subalgorithm**

- Complexity:

# Binary Heap - add

**subalgorithm** bubble-up (heap, p) **is:**

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**end-while**

heap.elems[poz]  $\leftarrow$  elem

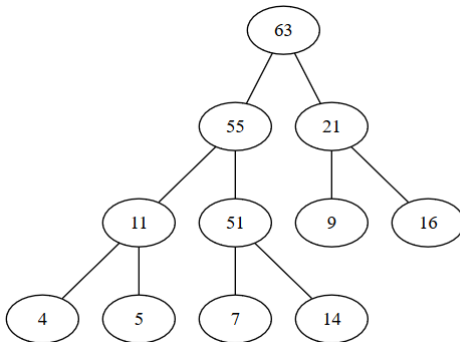
**end-subalgorithm**

- Complexity:  $O(\log_2 n)$
- Can you give an example when the complexity of the algorithm is less than  $\log_2 n$  (best case scenario)?



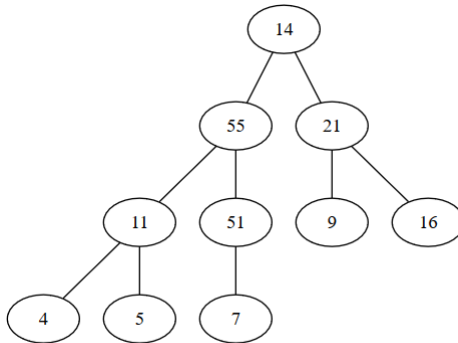
# Binary Heap - Remove - example

- From a heap we can only remove the root element.



# Binary Heap - Remove - example

- In order to keep the *heap structure*, when we remove the root, we are going to move the last element from the array to be the root.

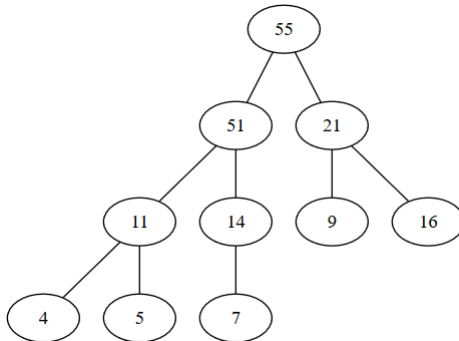


# Binary Heap - Remove - example

- *Heap property* is not kept: the root is no longer the maximum element.
- In order to restore the heap property, we will start a *bubble-down* process, where the new node will be swapped with its maximum child, until it becomes a leaf, or until it will be greater than both children.

# Binary Heap - Remove - example

- When the bubble-down process ends:



# Binary Heap - remove

```
function remove(heap) is:  
  //heap - is a heap  
  if heap.len = 0 then  
    @ error - empty heap  
  end-if  
  deletedElem  $\leftarrow$  heap.elems[1]  
  heap.elems[1]  $\leftarrow$  heap.elems[heap.len]  
  heap.len  $\leftarrow$  heap.len - 1  
  bubble-down(heap, 1)  
  remove  $\leftarrow$  deletedElem  
end-function
```

# Binary Heap - remove

**subalgorithm** bubble-down(heap, p) **is:**

*//heap - is a heap*

*//p - position from which we move down the element*

poz  $\leftarrow$  p

elem  $\leftarrow$  heap.elems[p]

**while** poz < heap.len **execute**

maxChild  $\leftarrow$  -1

**if** poz \* 2  $\leq$  heap.len **then**

*//it has a left child, assume it is the maximum*

maxChild  $\leftarrow$  poz\*2

**end-if**

**if** poz\*2+1  $\leq$  heap.len **and** heap.elems[2\*poz+1] > heap.elems[2\*poz] **th**

*//it has two children and the right is greater*

maxChild  $\leftarrow$  poz\*2 + 1

**end-if**

*//continued on the next slide...*

# Binary Heap - remove

```
if maxChild  $\neq$  -1 and heap.elems[maxChild] > elem then  
    heap.elems[poz]  $\leftarrow$  heap.elems[maxChild]  
    poz  $\leftarrow$  maxChild  
else  
    poz  $\leftarrow$  heap.len + 1  
    //to stop the while loop  
end-if  
end-while  
end-subalgorithm
```

- Complexity:

# Binary Heap - remove

```
if maxChild  $\neq$  -1 and heap.elems[maxChild] > elem then  
    heap.elems[poz]  $\leftarrow$  heap.elems[maxChild]  
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- Complexity:  $O(\log_2 n)$
- Can you give an example when the complexity of the algorithm is less than  $\log_2 n$  (best case scenario)?



# Questions

- In a max-heap where can we find the:
  - maximum element of the array?

# Questions

- In a max-heap where can we find the:
  - maximum element of the array?
  - minimum element of the array?

# Questions

- In a max-heap where can we find the:
  - maximum element of the array?
  - minimum element of the array?
- Assume you have a MAX-HEAP and you need to add an operation that returns the minimum element of the heap. How would you implement this operation, using constant time and space? (Note: we only want to return the minimum, we do not want to be able to remove it).

# Exercises

- Consider an initially empty Binary MAX-HEAP and insert the elements 8, 27, 13, 15\*, 32, 20, 12, 50\*, 29, 11\* in it. Draw the heap in the tree form after the insertion of the elements marked with a \* (3 drawings). Remove 3 elements from the heap and draw the tree form after every removal (3 drawings).
- Insert the following elements, in this order, into an initially empty MIN-HEAP: 15, 17, 9, 11, 5, 19, 7. Remove all the elements, one by one, in order from the resulting MIN HEAP. Draw the heap after every second operation (after adding 17, 11, 19, etc.)

# Heap-sort

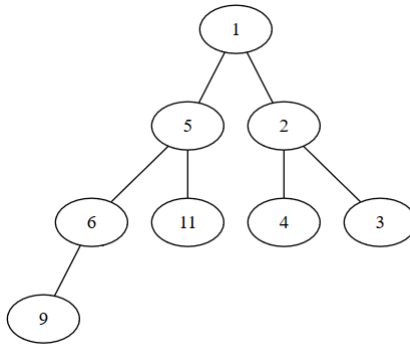
- There is a sorting algorithm, called *Heap-sort*, that is based on the use of a heap.
- In the following we are going to assume that we want to sort a sequence in ascending order.
- Let's sort the following sequence: [6, 1, 3, 9, 11, 4, 2, 5]

# Heap-sort - Naive approach

- Based on what we know so far, we can guess how heap-sort works:
  - Build a min-heap adding elements one-by-one to it.
  - Start removing elements from the min-heap: they will be removed in the sorted order.

# Heap-sort - Naive approach

- The heap when all the elements were added:



- When we remove the elements one-by-one we will have: 1, 2, 3, 4, 5, 6, 9, 11.

# Heap-sort - Naive approach

- What is the time complexity of the heap-sort algorithm described above?



# Heap-sort - Naive approach

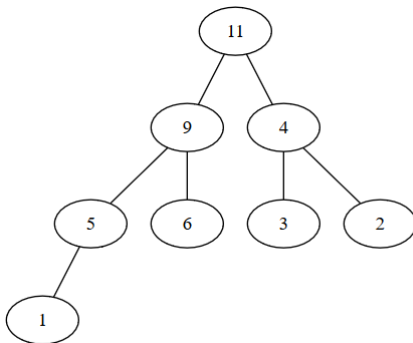
- What is the time complexity of the heap-sort algorithm described above?
- The time complexity of the algorithm is  $O(n \log_2 n)$
- What is the extra space complexity of the heap-sort algorithm described above (do we need an extra array)?

# Heap-sort - Naive approach

- What is the time complexity of the heap-sort algorithm described above?
- The time complexity of the algorithm is  $O(n \log_2 n)$
- What is the extra space complexity of the heap-sort algorithm described above (do we need an extra array)?
- The extra space complexity of the algorithm is  $\Theta(n)$  - we need an extra array.

# Heap-sort - Better approach

- If instead of building a min-heap, we build a max-heap (even if we want to do ascending sorting), we do not need the extra array.



# Heap-sort - Better approach

- We can improve the time complexity of building the heap as well.

# Heap-sort - Better approach

- We can improve the time complexity of building the heap as well.
  - If we have an unsorted array, we can transform it easier into a heap: the second half of the array will contain leaves, they can be left where they are.
  - Starting from the first non-leaf element (and going towards the beginning of the array), we will just call *bubble-down* for every element.
  - Time complexity of this approach:  $O(n)$  (but removing the elements from the heap is still  $O(n \log_2 n)$ )

# Priority Queue - Representation on a binary heap

- When an element is pushed to the priority queue, it is simply added to the heap (and bubbled-up if needed)
- When an element is popped from the priority queue, the root is removed from the heap (and bubble-down is performed if needed)
- Top simply returns the root of the heap.

# Priority Queue - Representation

- Let's complete our table with the complexity of the operations if we use a heap as representation:

Operation	Sorted	Non-sorted	Heap
push	$O(n)$	$\Theta(1)$	$O(\log_2 n)$
pop	$\Theta(1)$	$\Theta(n)$	$O(\log_2 n)$
top	$\Theta(1)$	$\Theta(n)$	$\Theta(1)$

- Consider the total complexity of the following sequence of operations:
  - start with an empty priority queue
  - push  $n$  random elements to the priority queue
  - perform pop  $n$  times

# Priority Queue - Application

- Print all numbers of the form  $a^3 + b^3$  in ascending order where  $0 \leq a \leq N$  and  $0 \leq b \leq N$ .



# Priority Queue - Application

- Print all numbers of the form  $a^3 + b^3$  in ascending order where  $0 \leq a \leq N$  and  $0 \leq b \leq N$ .
  - Option 1: Generate all possible numbers, sort them and print the sorted array
  - We can use less memory if we use the fact that  $a^3 + b^3$  is always less than  $(a + 1)^3 + b^3$  and  $a^3 + (b + 1)^3$ . But if we push those two numbers when popping  $a^3 + b^3$ , we will have many duplicates in the priority queue ( $a^3 + b^3$  will be pushed both for  $(a - 1)^3 + b^3$  and  $a^3 + (b - 1)^3$ ).
  - Generate all numbers of the form  $0^3 + i^3$ , and add them in a priority queue. When a number is popped, we push  $(a + 1)^3 + b^3$ .

# Priority Queue - Extension

- We have discussed the *standard* interface of a Priority Queue, the one that contains the following operations:
  - push
  - pop
  - top
  - isEmpty
  - init
- Sometimes, depending on the problem to be solved, it can be useful to have the following three operations as well:
  - increase the priority of an existing element
  - delete an arbitrary element
  - merge two priority queues

# Priority Queue - Extension

- What is the complexity of these three extra operations if we use as representation a binary heap?
  - Increasing the priority of an existing element is  $O(\log_2 n)$  if we know the position where the element is.
  - Deleting an arbitrary element is  $O(\log_2 n)$  if we know the position where the element is.
  - Merging two priority queues has complexity  $\Theta(n)$  (assume both priority queues have  $n$  elements).

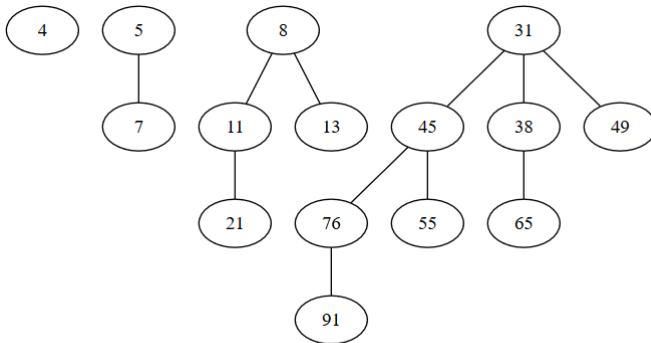
# Priority Queue - Other representations

- If we do not want to merge priority queues, a binary heap is a good representation. If we need the merge operation, there are other heap data structures that can be used, which offer a better complexity.
- Out of these data structures we are going to discuss one: the *binomial heap*.

# Binomial heap

- A *binomial heap* is a collection of *binomial trees*.
- A *binomial tree* can be defined in a recursive manner:
  - A *binomial tree of order 0* is a single node.
  - A *binomial tree of order  $k$*  is a tree which has a root and  $k$  children, each being the root of a binomial tree of order  $k - 1$ ,  $k - 2$ , ..., 2, 1, 0 (in this order).

# Binomial tree - Example

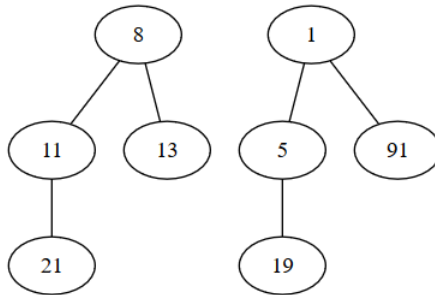


Binomial trees of order 0, 1, 2 and 3

# Binomial tree

- A binomial tree of order  $k$  has exactly  $2^k$  nodes.
- The height of a binomial tree of order  $k$  is  $k$ .
- If we delete the root of a binomial tree of order  $k$ , we will get  $k$  binomial trees, of orders  $k - 1, k - 2, \dots, 2, 1, 0$ .
- Two binomial trees of the same order  $k$  can be merged into a binomial tree of order  $k + 1$  by setting one of them to be the leftmost child of the other.

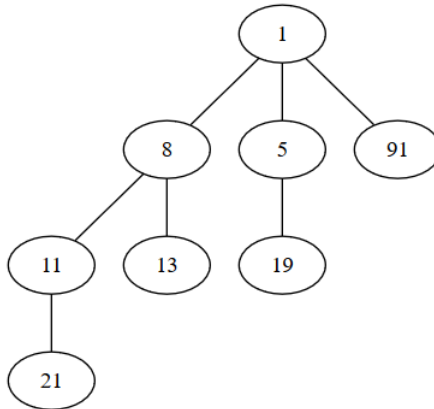
# Binomial tree - Merge I



Before merge we have two binomial trees of order 2

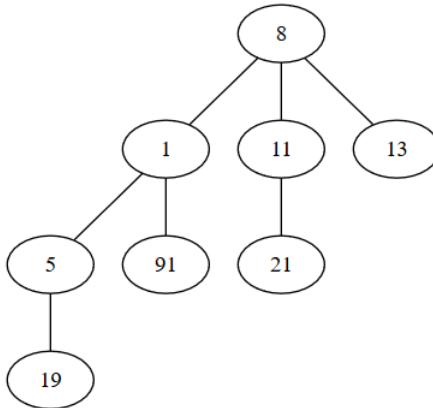


# Binomial tree - Merge II



One way of merging the two binomial trees into one of order 3

# Binomial tree - Merge III



Another way of merging the two binomial trees into one of order 3

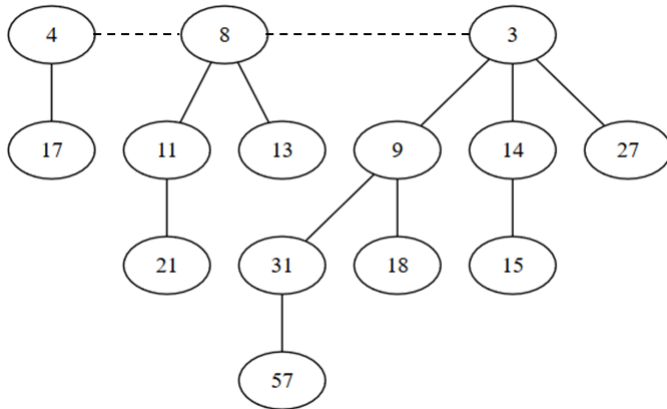
# Binomial tree - representation

- If we want to implement a binomial tree, we can use the following representation:
  - We need a structure for nodes, and for each node we keep the following:
    - The information from the node
    - The address of the parent node
    - The address of the first child node
    - The address of the next sibling node
  - For the tree we will keep the address of the root node (and probably the order of the tree)

# Binomial heap

- A binomial heap is made of a collection/sequence of binomial trees with the following property:
  - Each binomial tree respects the heap-property: for every node, the value from the node is less than the value of its children (assume MIN\_HEAPS).
  - There can be at most one binomial tree of a given order  $k$ .
  - As representation, a binomial heap is usually a sorted linked list, where each node contains a binomial tree, and the list is sorted by the order of the trees.

# Binomial tree - Example



Binomial heap with 14 nodes, made of 3 binomial trees of orders 1, 2 and 3

# Binomial tree

- For a given number of elements,  $n$ , the structure of a binomial heap (i.e. the number of binomial trees and their orders) is unique.
- The structure of the binomial heap is determined by the binary representation of the number  $n$ .
- For example  $14 = 1110$  (in binary)  $= 2^3 + 2^2 + 2^1$ , so a binomial heap with 14 nodes contains binomial trees of orders 3, 2, 1 (but they are stored in the reverse order: 1, 2, 3).
- For example  $21 = 10101 = 2^4 + 2^2 + 2^0$ , so a binomial heap with 21 nodes contains binomial trees of orders 4, 2, 0.

# Binomial heap

- A binomial heap with  $n$  elements contains at most  $\log_2 n$  binomial trees.
- The height of the binomial heap is at most  $\log_2 n$ .

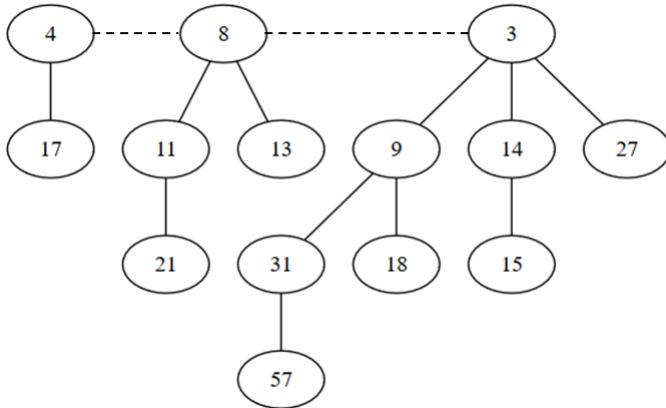
# Binomial heap - merge

- The most interesting operation for two binomial heaps is the merge operation, which is used by other operations as well. After the merge operation the two previous binomial heaps will no longer exist, we will only have the result.
- Since both binomial heaps are sorted linked lists, the first step is to *merge* the two linked lists (standard merge algorithm for two sorted linked lists).
- The result of the merge can contain two binomial trees of the same order, so we have to iterate over the resulting list and transform binomial trees of the same order  $k$  into a binomial tree of order  $k + 1$ . When we merge the two binomial trees we must keep the heap property.

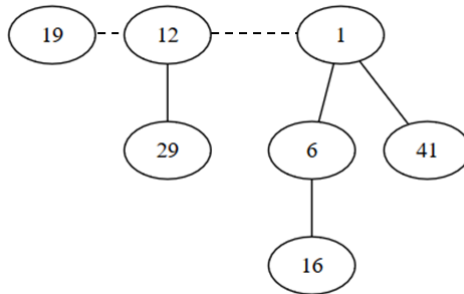


# Binomial heap - merge - example I

- Let's merge the following two binomial heaps:

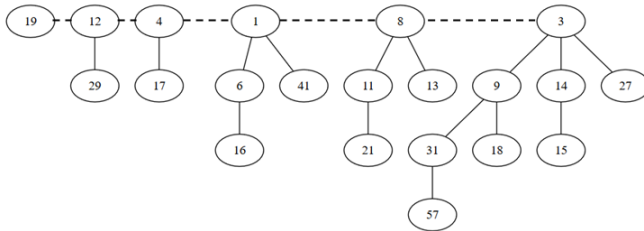


# Binomial heap - merge - example II



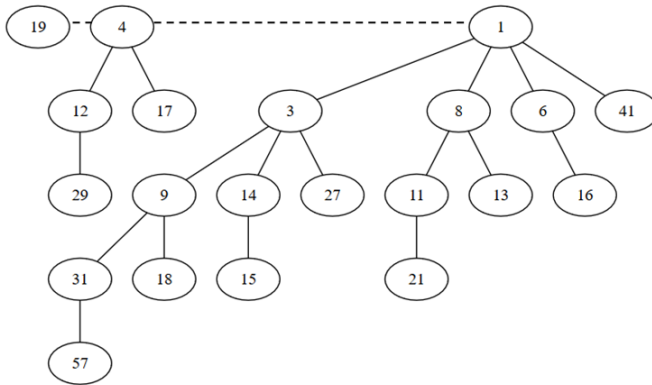
# Binomial heap - merge - example III

- After merging the two linked lists of binomial trees:



# Binomial heap - merge - example IV

- After transforming the trees of the same order (final result of the merge operation).



# Binomial heap - Merge operation

- If both binomial heaps have  $n$  elements, merging them will have  $O(\log_2 n)$  complexity (the maximum number of binomial trees for a binomial heap with  $n$  elements is  $\log_2 n$ ).

# Binomial heap - other operations I

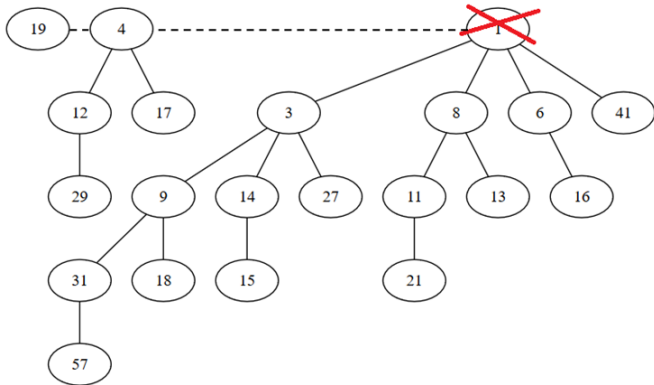
- Most of the other operations that we have for the binomial heap (because we need them for the priority queue) will use the merge operation presented above.
- *Push operation:* Inserting a new element means creating a binomial heap with just that element and merging it with the existing one. Complexity of insert is  $O(\log_2 n)$  in worst case ( $\Theta(1)$  amortized).
- *Top operation:* The minimum element of a binomial heap (the element with the highest priority) is the root of one of the binomial trees. Returning the minimum means checking every root, so it has complexity  $O(\log_2 n)$ .

# Binomial heap - other operations II

- *Pop operation:* Removing the minimum element means removing the root of one of the binomial trees. If we delete the root of a binomial tree, we will get a sequence of binomial trees. These trees are transformed into a binomial heap (just reverse their order), and a merge is performed between this new binomial heap and the one formed by the remaining elements of the original binomial heap.

# Binomial heap - other operations III

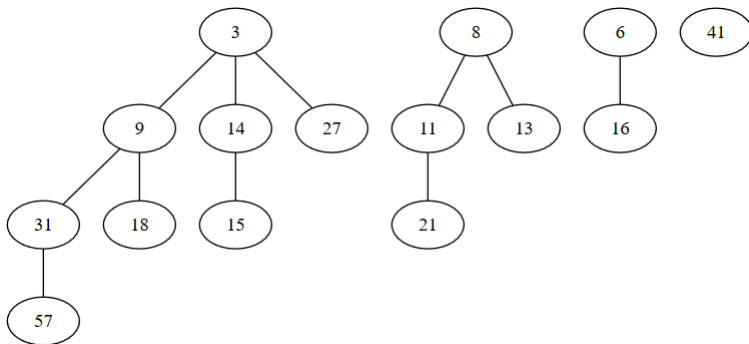
- The minimum is one of the roots.





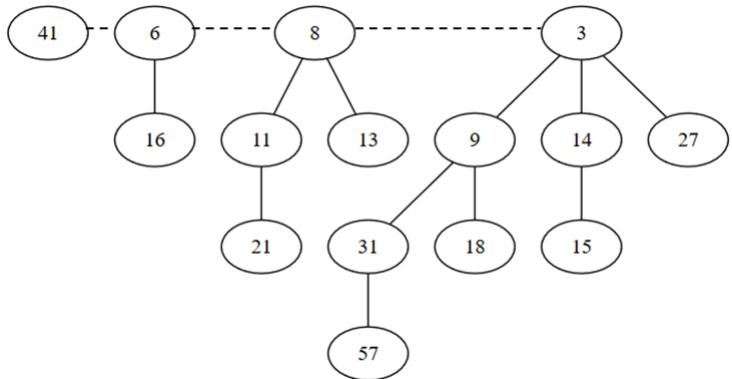
# Binomial heap - other operations IV

- Break the corresponding tree into  $k$  binomial trees



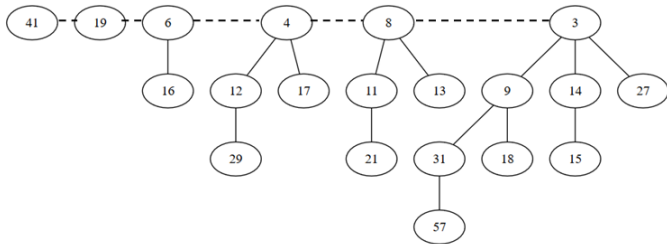
# Binomial heap - other operations V

- Create a binomial heap of these trees



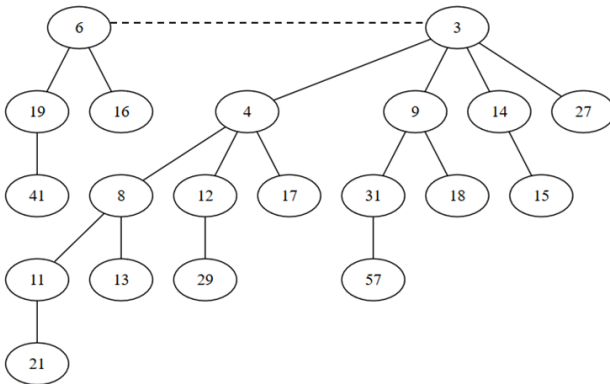
# Binomial heap - other operations VI

- Merge it with the existing one (after the merge algorithm)



# Binomial heap - other operations VII

- After the transformation



- The complexity of the remove-minimum operation is  $O(\log_2 n)$

# Binomial heap - other operations VIII

- Assuming that we have a pointer to the element whose priority has to be increased (in our figures lower number means higher priority), we can just change the priority and bubble-up the node if its priority is greater than the priority of its parent. Complexity of the operation is:  $O(\log_2 n)$
- Assuming that we have a pointer to the element that we want to delete, we can first decrease its priority to  $-\infty$  (this will move it to the root of the corresponding binomial tree) and remove it. Complexity of the operation is:  $O(\log_2 n)$

# Problems with stacks, queues and priority queues I

- Red-Black Card Game:
  - Statement: Two players each receive  $\frac{n}{2}$  cards, where each card can be red or black. The two players take turns; at every turn the current player puts the card from the upper part of his/her deck on the table. If a player puts a red card on the table, the other player has to take all cards from the table and place them at the bottom of his/her deck. The winner is the player that has all the cards.
  - Requirement: Given the number  $n$  of cards, simulate the game and determine the winner.
  - Hint: use stack(s) and queue(s)

# Problems with stacks, queues and priority queues II

- Robot in a maze:
  - Statement: There is a rectangular maze, composed of occupied cells (X) and free cells (\*). There is a robot (R) in this maze and it can move in 4 directions: N, S, E, V.
  - Requirements:
    - Check whether the robot can get out of the maze (get to the first or last line or the first or last column).
    - Find a path that will take the robot out of the maze (if exists).

X	*	*	X	X	X	*	*
X	*	X	*	*	*	*	*
X	*	*	*	*	*	X	*
X	X	X	*	*	*	X	*
*	X	*	*	R	X	X	*
*	*	*	X	X	X	X	*
*	*	*	*	*	*	*	X
X	X	X	X	X	X	X	X

# Problems with stacks, queues and priority queues III

- Hint:
  - Let  $T$  be the set of positions where the robot can get from the starting position.
  - Let  $s$  be the set of positions to which the robot can get at a given moment and from which it could continue going to other positions.
  - A possible way of determining the sets  $T$  and  $S$  could be the following:



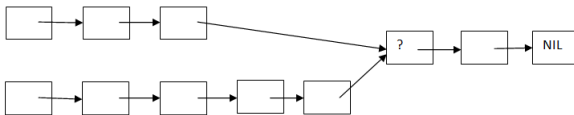
# Problems with stacks, queues and priority queues IV

```
T ← {initial position}
S ← {initial position}
while S ≠ ∅ execute
    Let  $p$  be one element of S
    S ← S \ { $p$ }
    for each valid position  $q$  where we can get from  $p$  and which is not in  $T$  do
        T ← T ∪ { $q$ }
        S ← S ∪ { $q$ }
    end-for
end-while
```

- T can be a list, a vector or a matrix associated to the maze
- S can be a stack or a queue (or even a priority queue, depending on what we want)

# Think about it - Linked Lists

- Write a non-recursive algorithm to reverse a singly linked list with  $\Theta(n)$  time complexity, using constant space/memory.
- Suppose there are two singly linked lists both of which intersect at some point and become a single linked list (see the image below). The number of nodes in the two list before the intersection is not known and may be different in each list. Give an algorithm for finding the merging point (hint - use a Stack)



# Think about it - Stacks and Queues I

- How can we implement a Stack using two Queues? What will be the complexity of the operations?
- How can we implement a Queue using two Stacks? What will be the complexity of the operation?
- How can we implement two Stacks using only one array? The stack operations should throw an exception only if the total number of elements in the two Stacks is equal to the size of the array.

# Think about it - Stacks and Queues II

- Given a string of lower-case characters, recursively remove adjacent duplicate characters from the string. For example, for the word "mississippi" the result should be "m".
- Given an integer  $k$  and a queue of integer numbers, how can we reverse the order of the first  $k$  elements from the queue? For example, if  $k=4$  and the queue has the elements [10, 20, 30, 40, 50, 60, 70, 80, 90], the output should be [40, 30, 20, 10, 50, 60, 70, 80, 90].

# Think about it - Priority Queues

- How can we implement a stack using a Priority Queue?
- How can we implement a queue using a Priority Queue?