30th September 2020

- Binary heaps
- Leftist heaps
- Binomial heaps
- Fibonacci heaps

Priority queues are important in, among other things, operating systems (process control in multitasking systems), search algorithms (A, A*, D*, etc.), and simulation.

Priority queues are data structures that hold elements with some kind of priority (*key*) in a queue-like structure, implementing the following operations:

- insert() Inserting an element into the queue.
- deleteMin() Removing the element with the highest priority.

And maybe also:

- buildHeap() Build a queue from a set (>1) of elements.
- increaseKey()/DecreaseKey() Change priority.
- delete() Removing an element from the queue.
- merge() Merge two queues.

An unsorted linked list can be used. insert() inserts an element at the head of the list (O(1)), and deleteMin() searches the list for the element with the highest priority and removes it (O(n)).

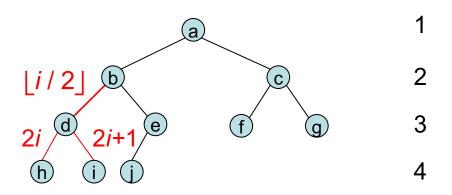
A sorted list can also be used (reversed running times).

Not very efficient implementations.

To make an efficient priority queue, it is enough to keeps the elements "almost sorted".

A binary heap is organized as a complete binary tree. (All levels are full, except possibly the last.)

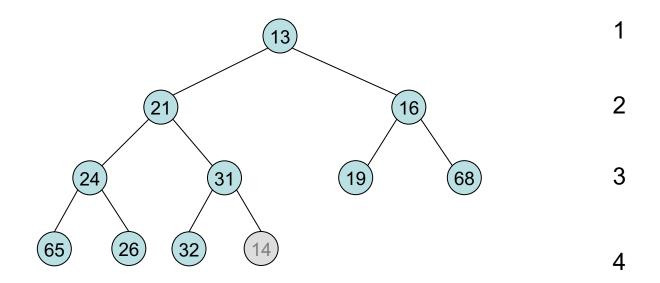
In a binary heap the element in the root must have a key less than or equal to the key of its children, in addition each sub-tree must be a binary heap.





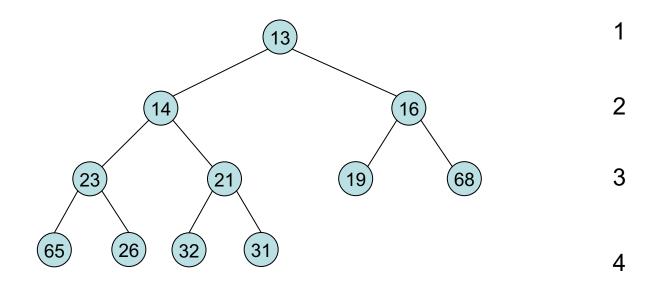
	а	b	С	d	е	f	g	h	i	j				
0	1	2	3	4	5	6	7	8	9	10	11	12	13	
	1	2 3						4						

insert(14)

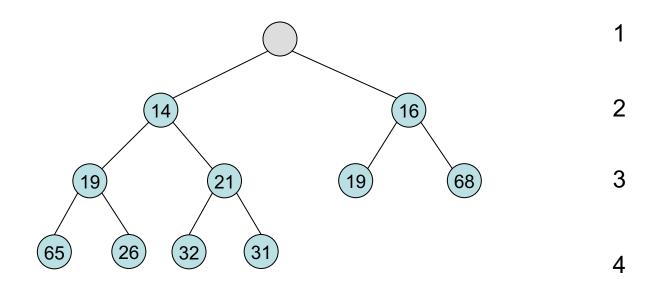


	13	21	16	24	31	19	68	65	26	32	14		
0	1	2	3	4	5	6	7	8	9	10	11	12	13
	1	2	2	3					4				

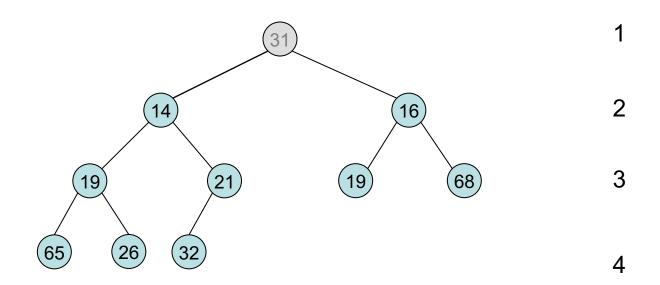
insert(14)



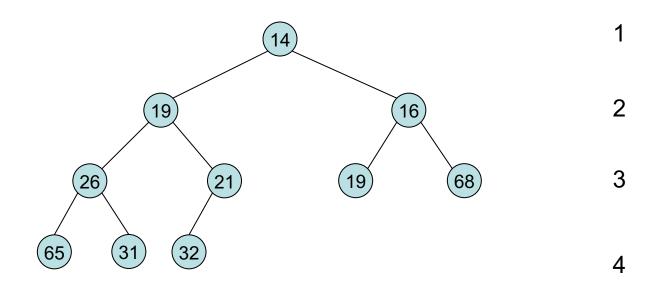
	13	14	16	24	21	19	68	65	26	32	31		
0	1	2	3	4	5	6	7	8	9	10	11	12	13
	1	2	2	3					4				



		14	16	19	21	19	68	65	26	32	31		
0	1	2	3	4	5	6	7	8	9	10	11	12	13
	1	2	2	3				4					



	31	14	16	19	21	19	68	65	26	32			
0	1	2	3	4	5	6	7	8	9	10	11	12	13
	1	2	2	3					4				



	14	19	16	19	21	26	68	65	31	32			
0	1	2	3	4	5	6	7	8	9	10	11	12	13
	1	2	2	3				4					

worst case average

insert() $O(\log N)$ O(1)

 $deleteMin() O(\log N) O(\log N)$

buildHeap() O(N)

(Insert elements into the array unsorted, and run percolateDown() on each root in the resulting heap (the tree), bottom up)

(The sum of the heights of a binary tree with N nodes is O(N).)

merge() O(N)

(N = number of elements)

To implement an efficient merge(), we move away from arrays, and implement so-called *leftist heaps* as pure trees.

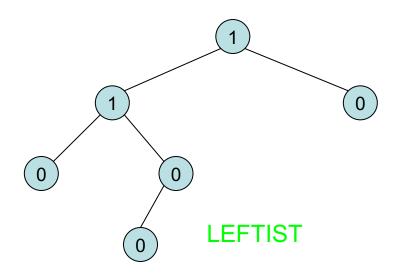
The idea is to make the heap (the tree) as skewed as possible, and do all the work on a short (right) branch, leaving the long (left) branch untouched.

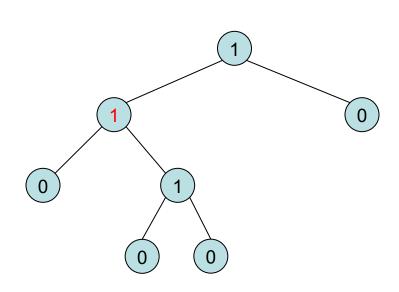
A *leftist heap* is still a binary tree with the heap structure (key in root is lower than key in children), but with an extra skewness requirement.

For all nodes X in our tree, we define the *null-path-length*(X) as the distance from X to a descendant with less than two children (*i.e.* 0 or 1).

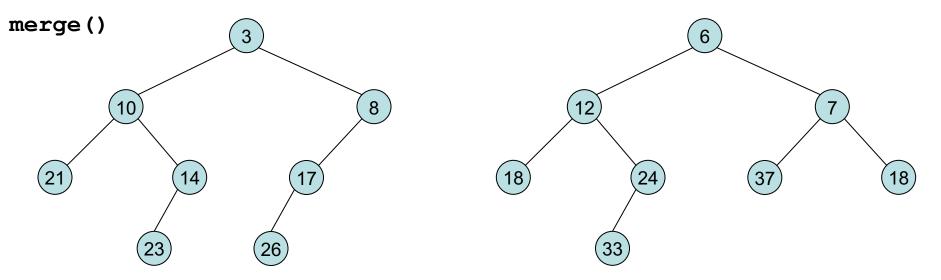
The skewness requirement is that for every node the null path length of its left child be at least as large as the null path length of the right child.

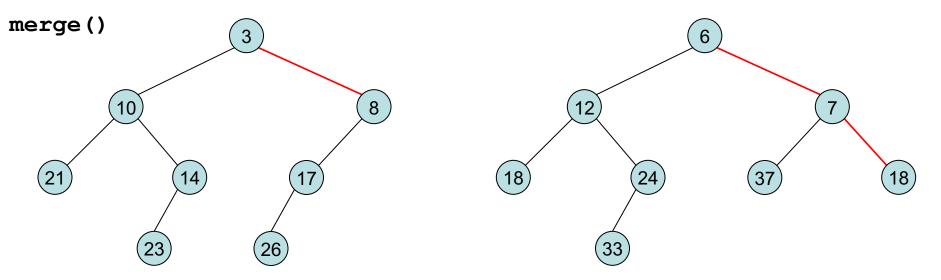
For the empty tree we define the *null-path-length* to be -1, as a special case.

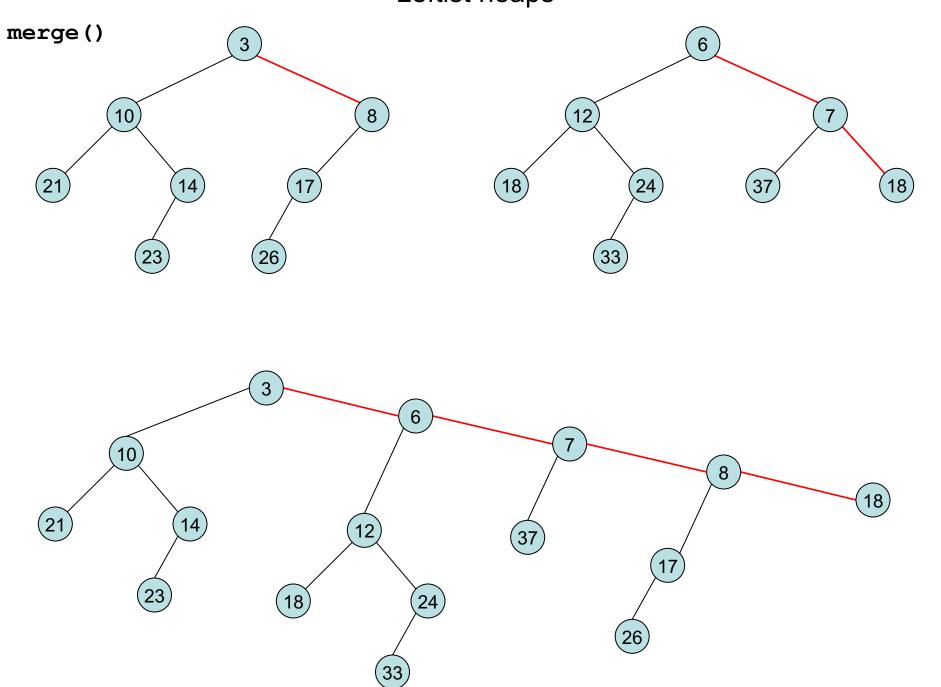


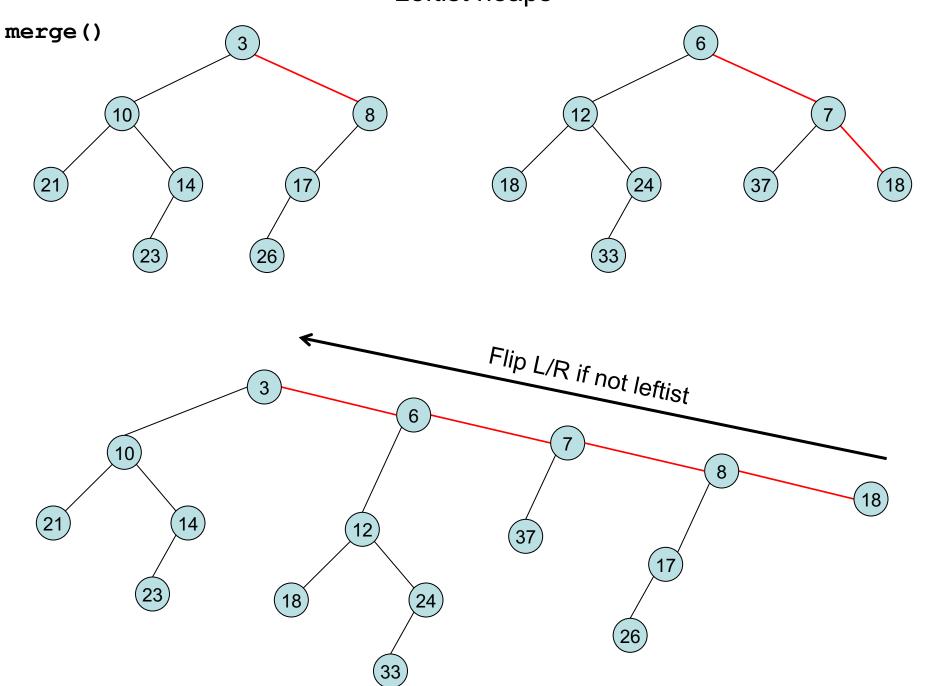


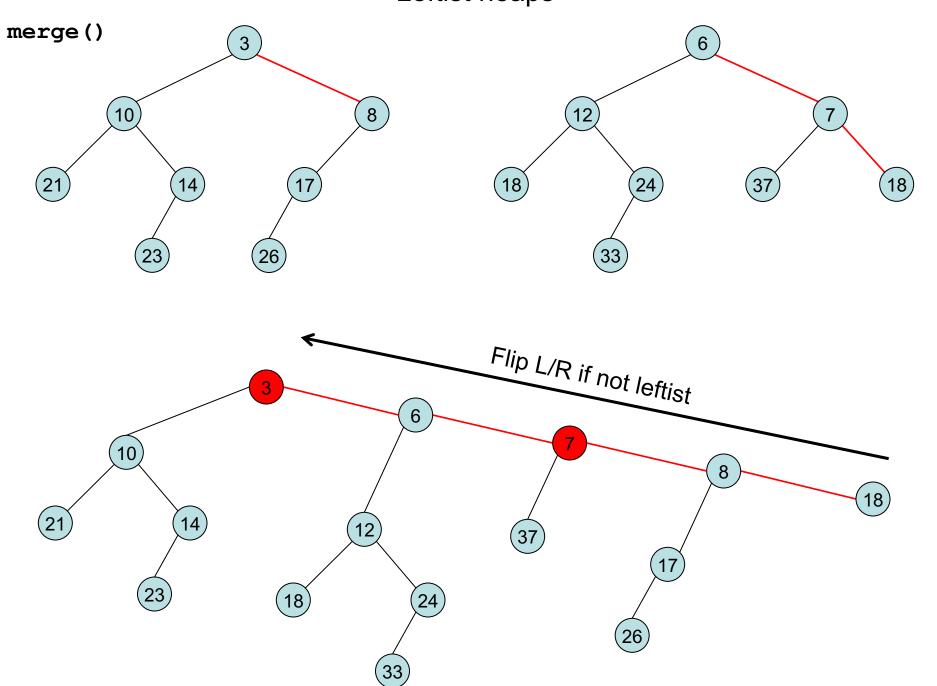
NOT LEFTIST

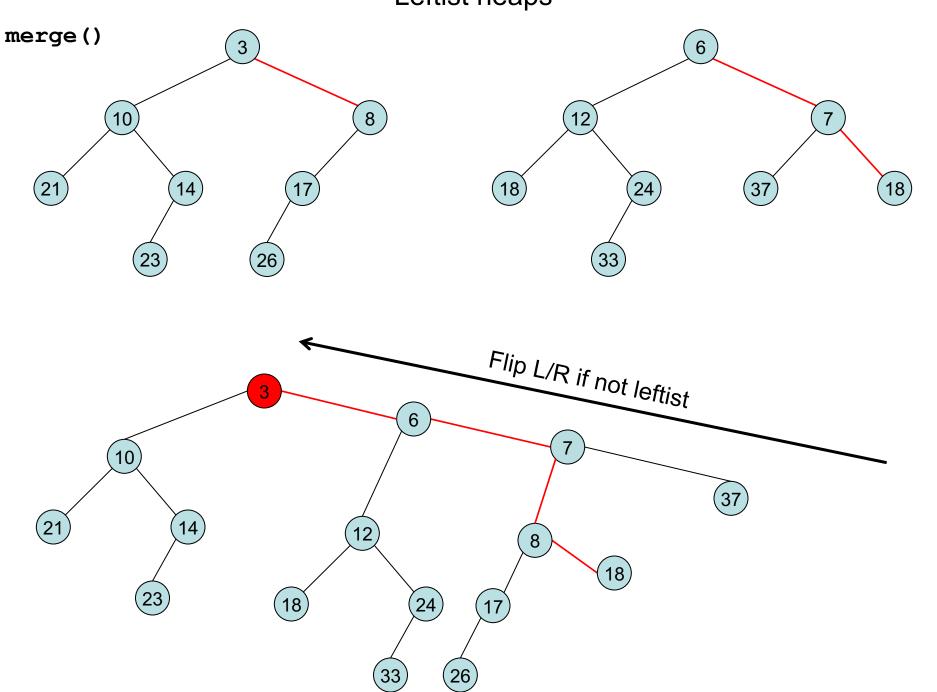


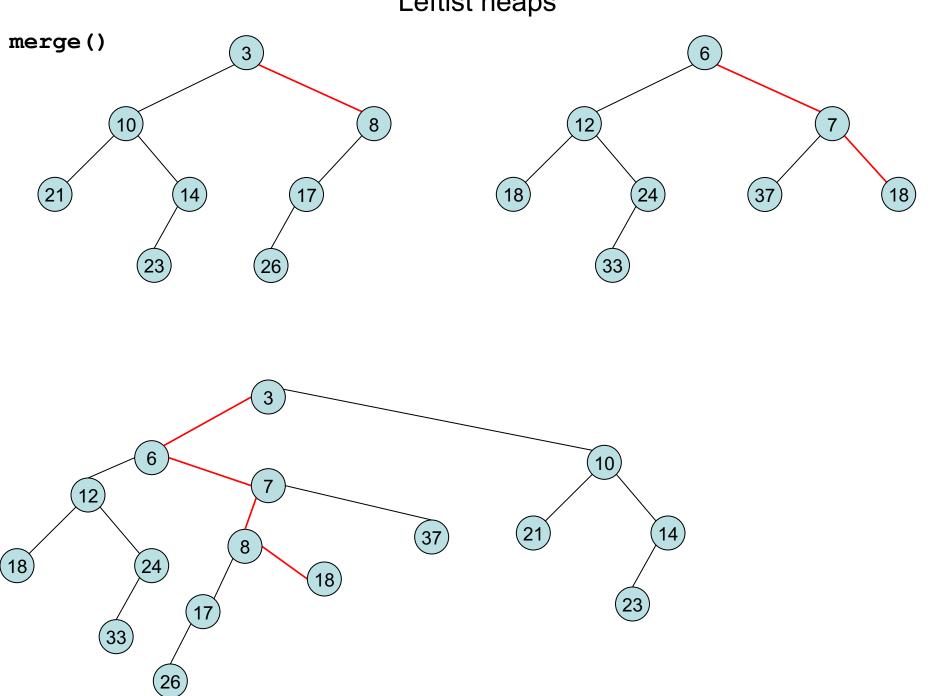




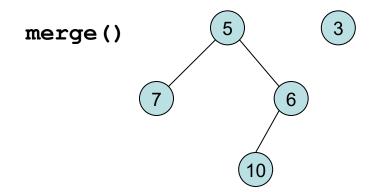


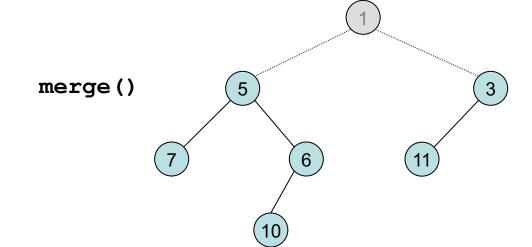






insert(3)





worst case

merge() $O(\log N)$

insert() $O(\log N)$

deleteMin() O(log N)

buildHeap() O(N)

(N = number of elements)

In a leftist heap with N nodes, the right path is at most $\lfloor \log (N+1) \rfloor \log$.

Leftist heaps:

merge(), insert() and deleteMin() in $O(\log N)$ time w.c.

Binary heaps:

insert() in O(1) time on average.

Binomial heaps

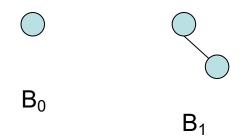
merge(), insert() og deleteMin() in $O(\log N)$ time w.c. insert() O(1) time on average

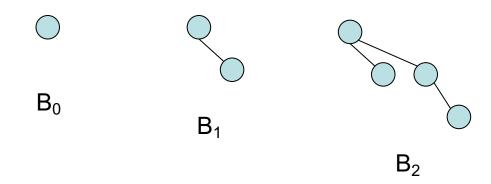
Binomial heaps are collections of trees (sometimes called a forest), each tree a heap.

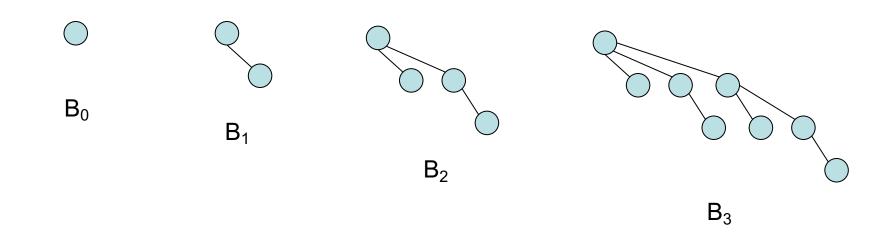
Binomial trees

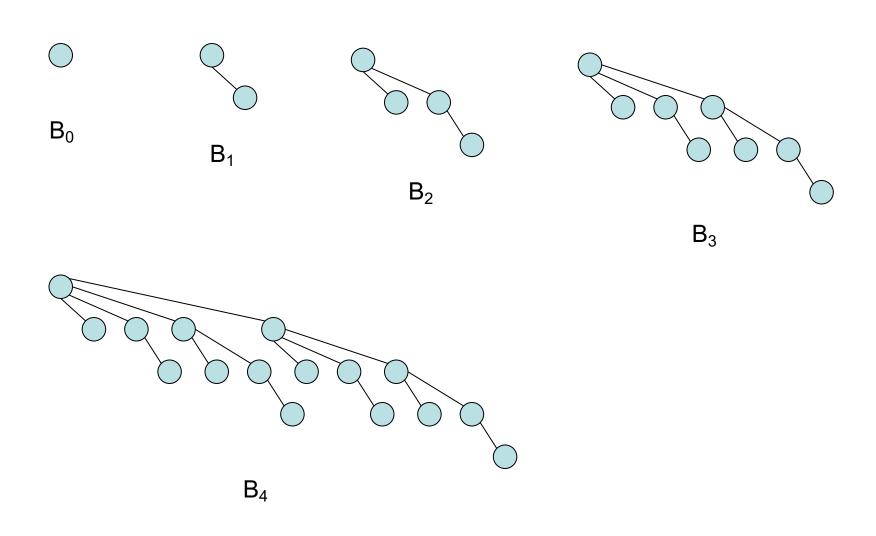


 B_0

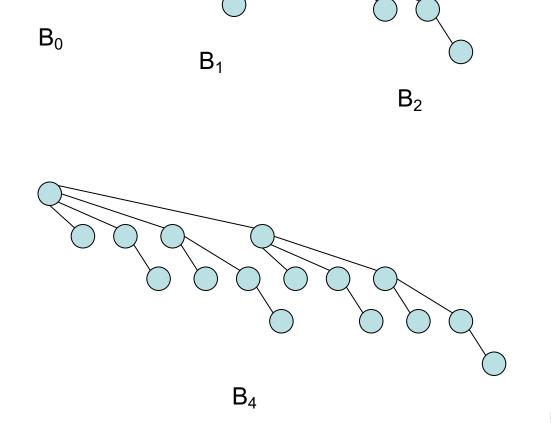


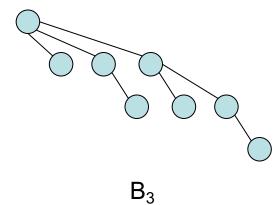






Binomial trees





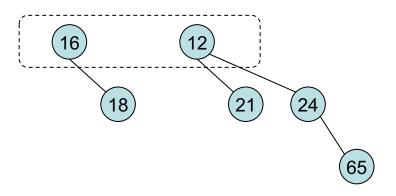
 $B_i = 2 \times B_{i-1}$, root of one tree connected as a child of the root of the other tree.

A tree of height *k* has:

 2^k nodes in total,

 $\binom{k}{d}$ nodes on level d.

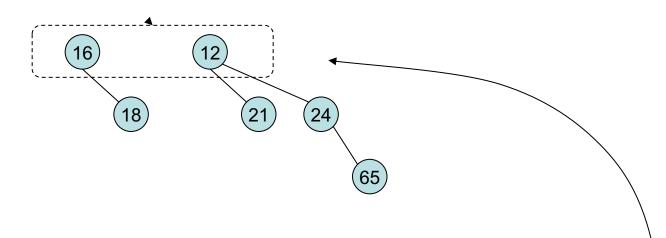
Binomial heap



Maximum one tree of each size:

6 elements: 6 binary = 011 (0+2+4) $B_0 B_1 B_2$

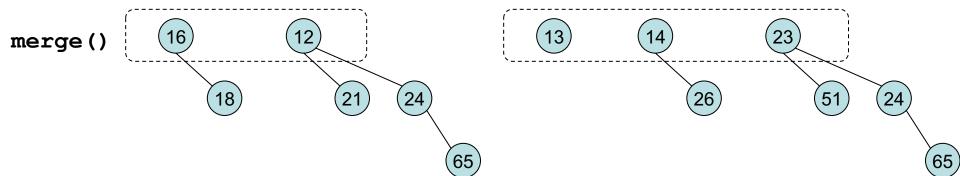
Binomial heap

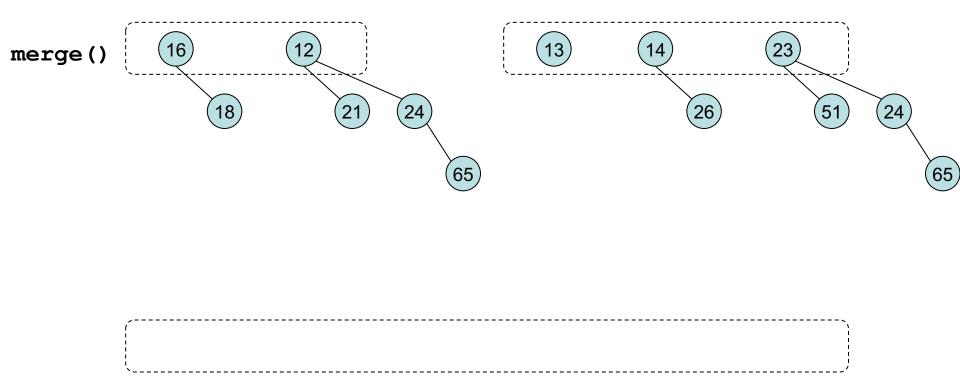


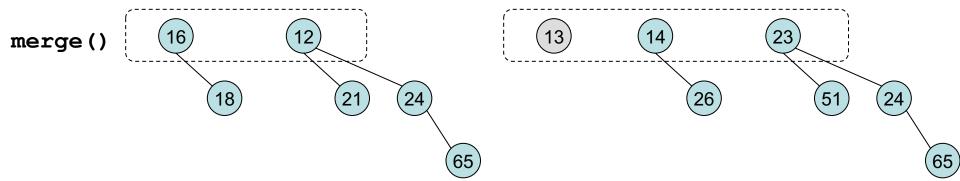
Maximum one tree of each size:

6 elements: 6 binary = 011 (0+2+4)
$$\mathbb{E}_0 B_1 B_2$$

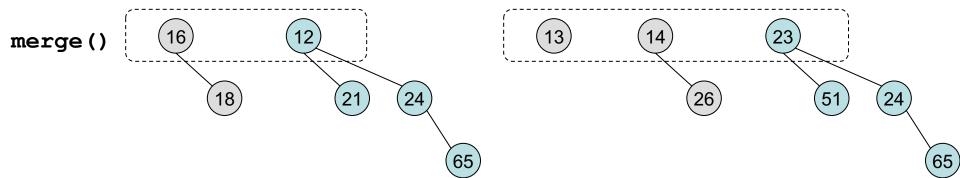
The length of the root list in a heap of *N* elements is $O(\log N)$. (Doubly linked, circular list.)

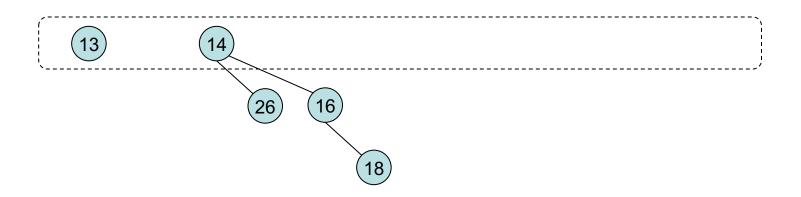


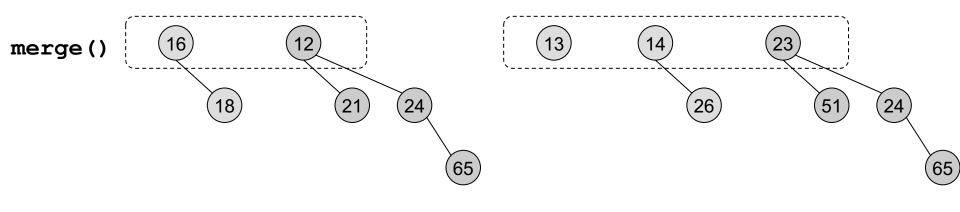


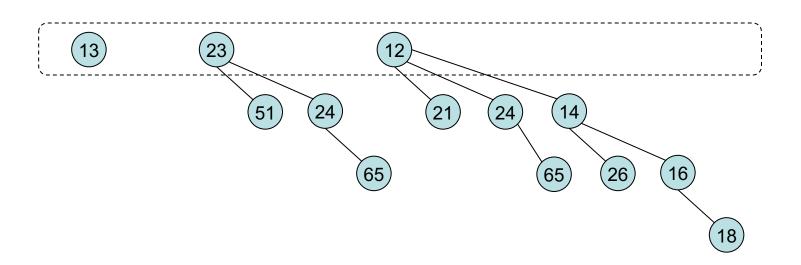


13)

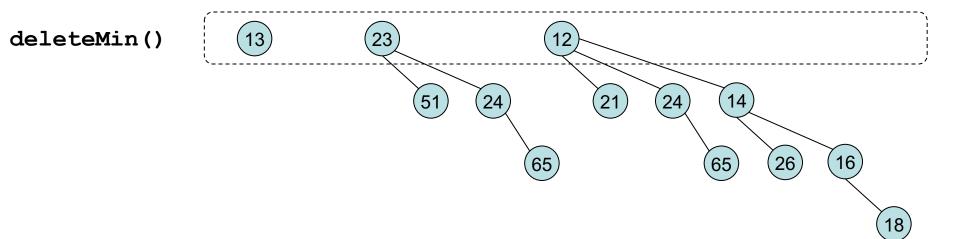


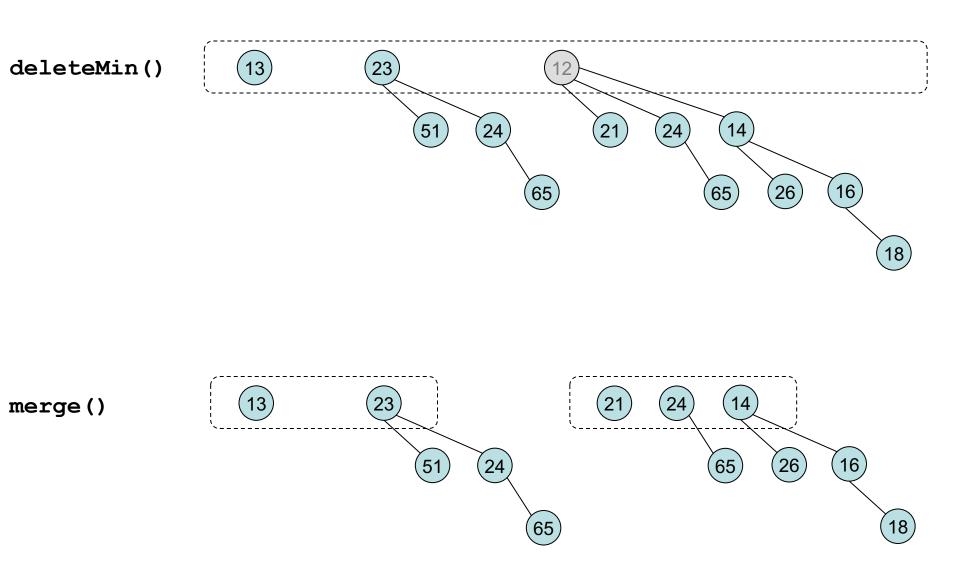






The trees (the root list) is kept sorted on height.





worst case average case

merge() $O(\log N)$ $O(\log N)$

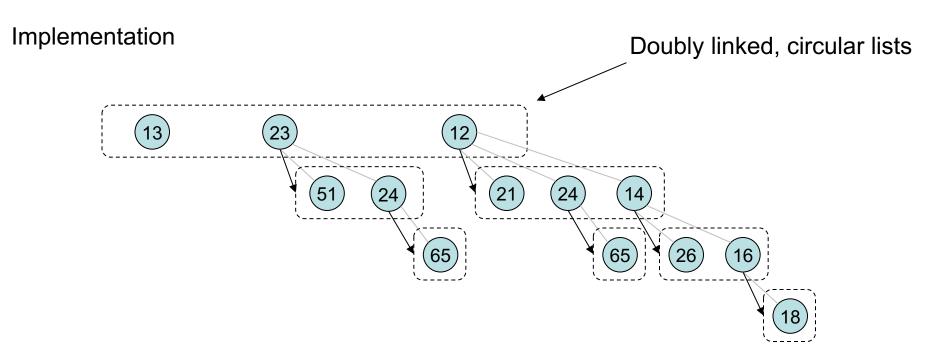
insert() $O(\log N)$ O(1)

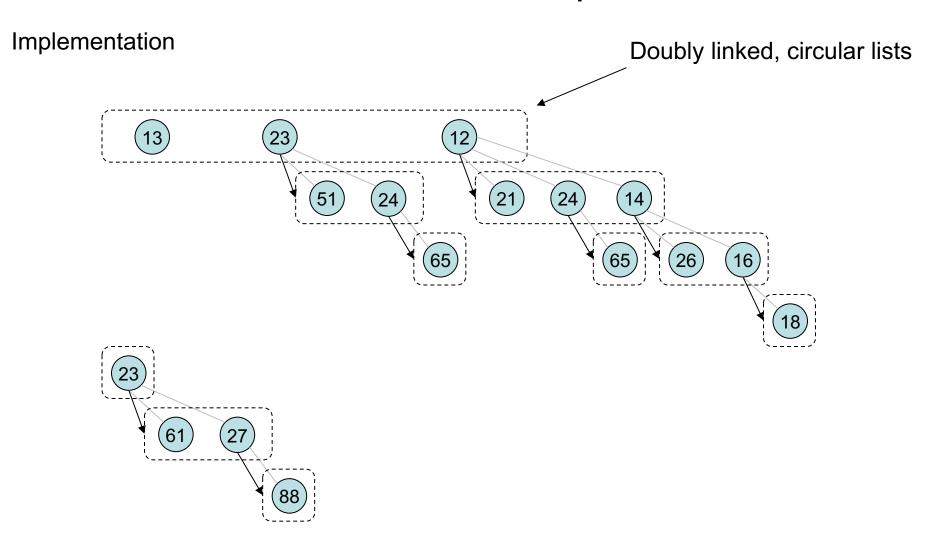
 $deleteMin() O(\log N) O(\log N)$

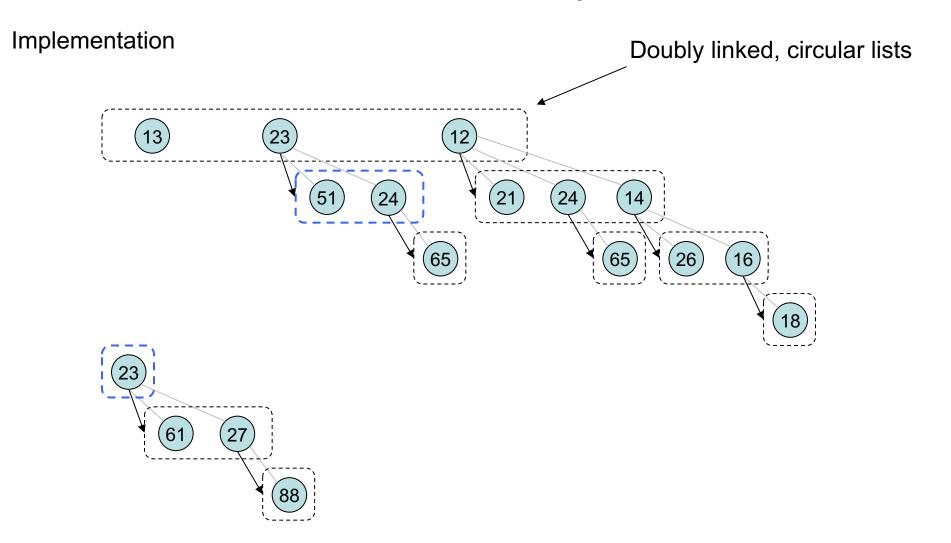
buildHeap() O(N)

(Run *N* insert() on an initially empty heap.)

(N = number of elements)







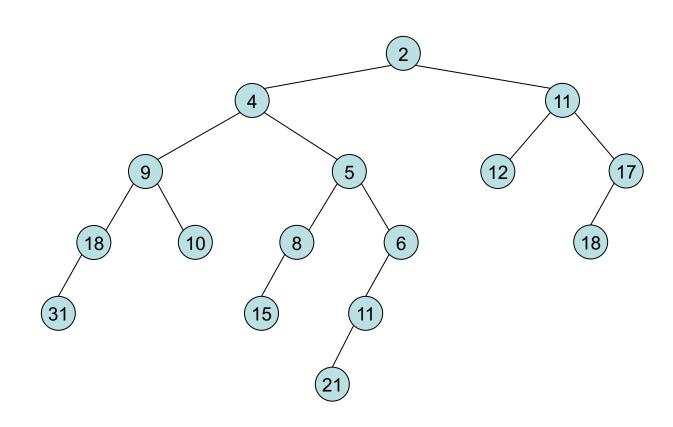
Very elegant, and in theory efficient, way to implement heaps: Most operations have O(1) amortized running time. (Fredman & Tarjan '87)

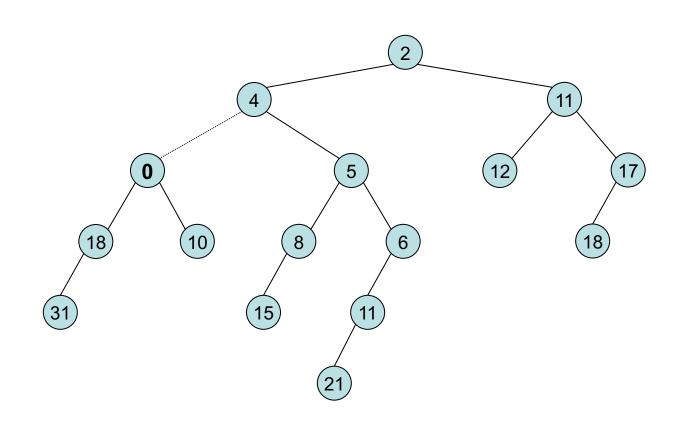
insert(), decreaseKey() and merge() O(1) amortized time
deleteMin()
O(log N) amortized time

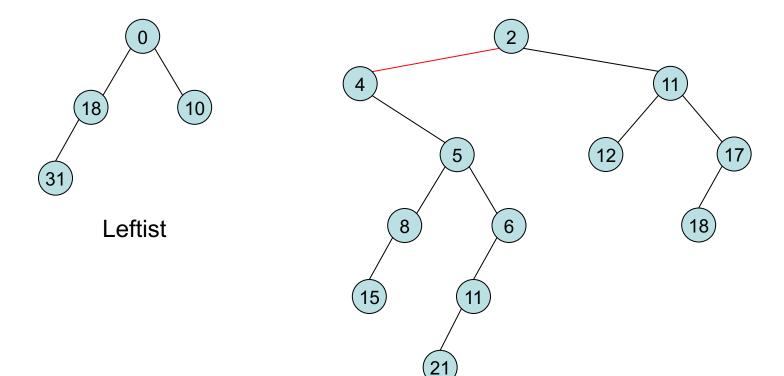
Combines elements from leftist heaps and binomial heaps.

A bit complicated to implement, and certain hidden constants are a bit high.

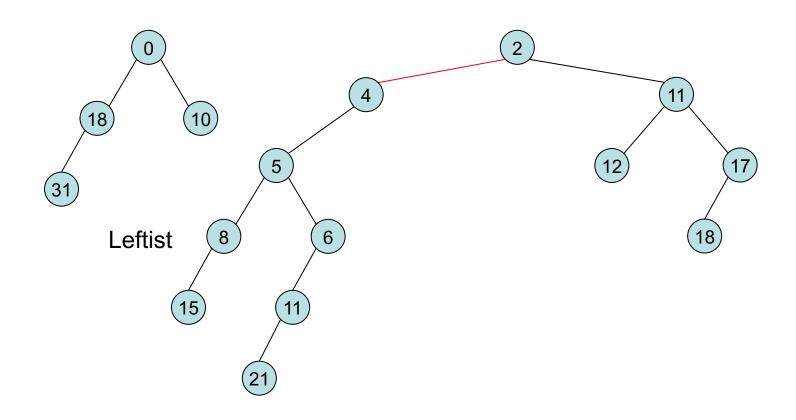
Best suited when there are few deleteMin() compared to the other operations. The data structure was developed for a shortest path algorithm (with many decreaseKey() operations), also used in spanning tree algorithms.



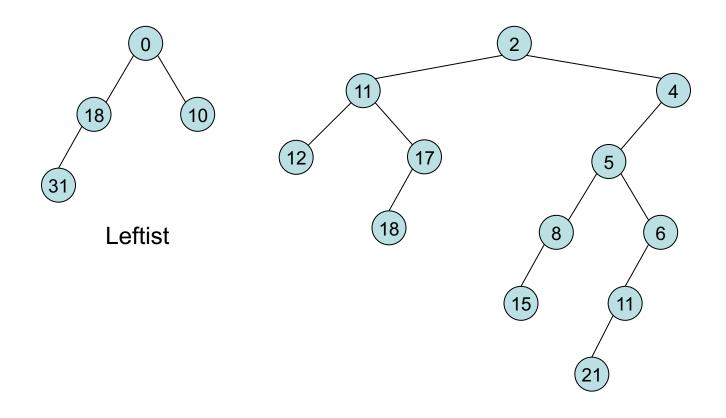




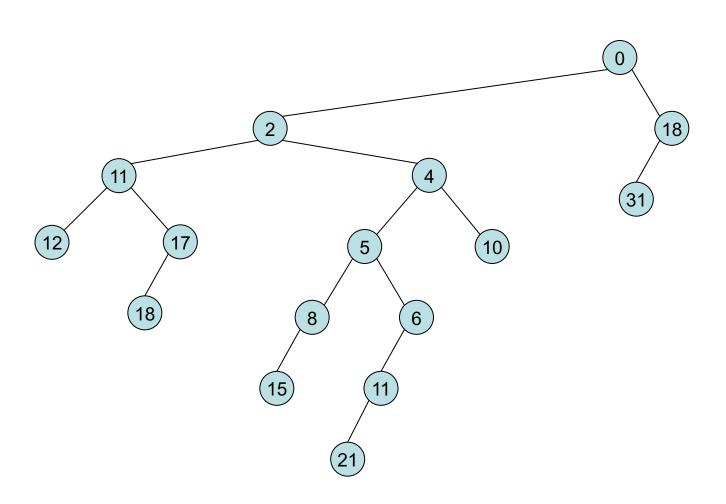
Ikke leftist



Ikke leftist

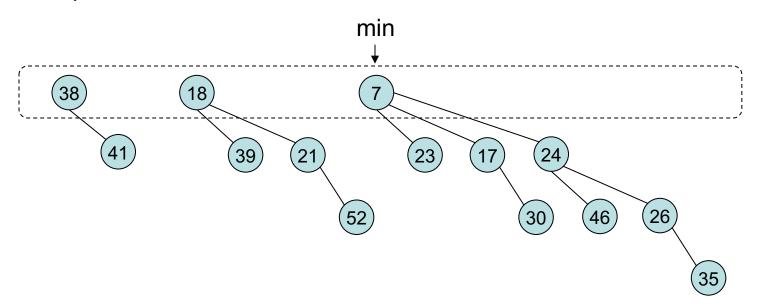


Leftist



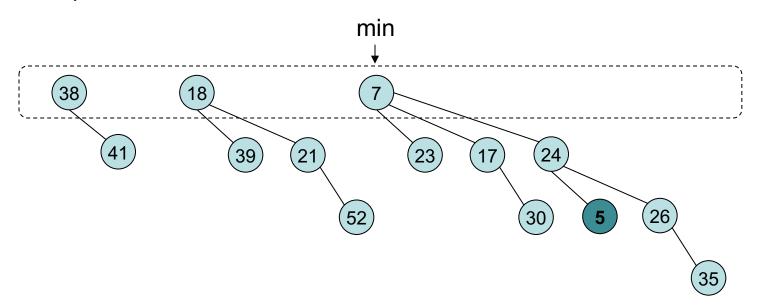
We include a smart decreaseKey() method from leftist heaps.

- Nodes are marked the first time child is removed.
- The second time a node gets a child removed, it is cut off, and becomes the root of a separate tree



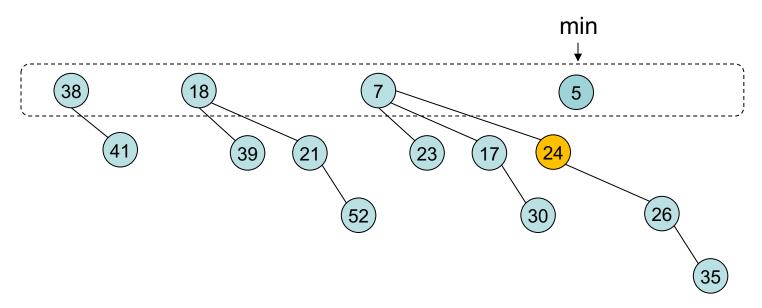
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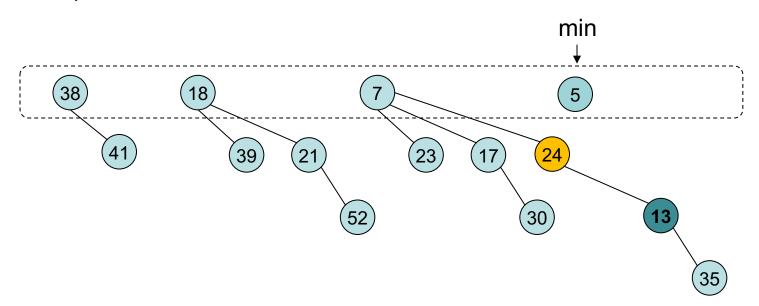
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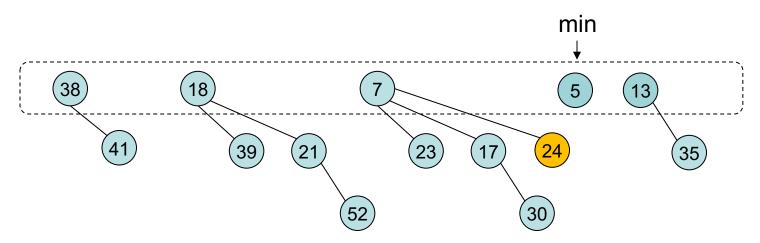
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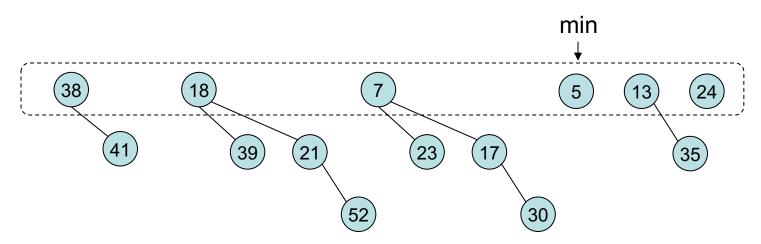
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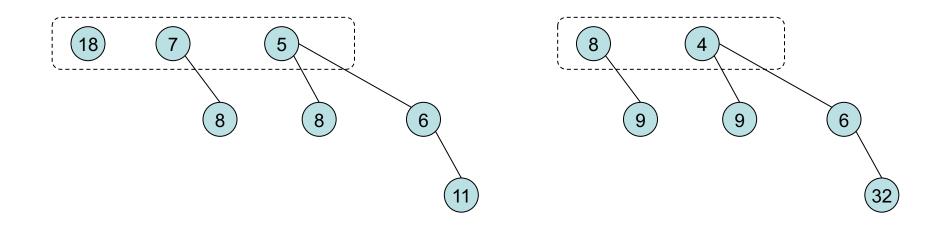


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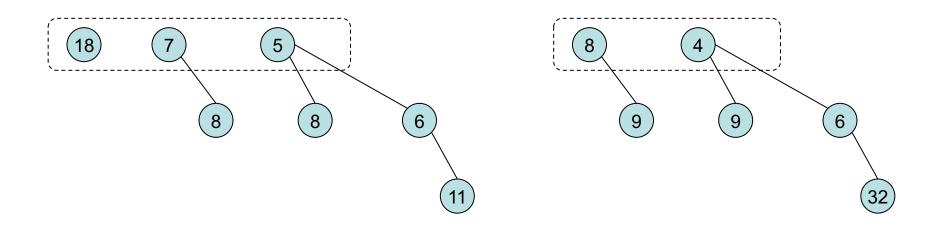
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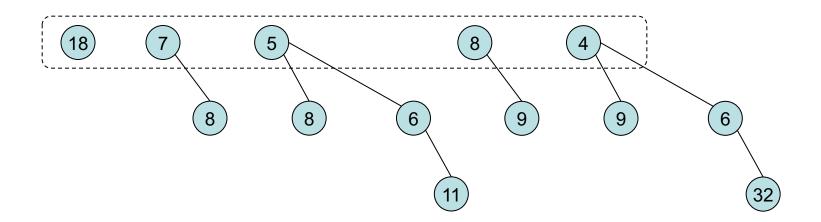


We use *lazy merging | lazy binomial queue*.



We use lazy merging / lazy binomial queue.





The problem with our decreaseKey()-method and *lazy merging* is that we have to clean up afterwards. This is done in by the deleteMin()-method which becomes expensive (O(log N) amortized time):

All trees are examined, we start with the smallest, and merge two and two, so that we get at most one tree of each size.

Each root has a number of children – this is used as the size of the tree. (Recall how we construct binomial trees, and that they may be partial as a result of decreaseKey()'s)

The trees are put in lists, one per size, and we begin merging, starting with the smallest.

Amortized time

insert() O(1)

decreaseKey() O(1)

merge() O(1)

deleteMin() O(log N)

buildHeap() O(N)

(Run *N* insert() on an initially empty heap.)

(N = number of elements)