

COMMONWEALTH OF AUSTRALIA

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FIT3155: Advanced Algorithms and Data Structures

Week 5: **Binomial heap** and its amortized analysis

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What is covered in this lecture?

Binomial heap and its amortized analysis

Source material and recommended reading

- Weiss, Data Structures and Algorithm Analysis (Chapters 6.8, 11.1, 11.2)
- Cormen et al., Introduction to Algorithms (Chapter 19) [\[link\]](#)

Priority queues (implemented using heaps)

Recall from FIT2004 that the heap data structure was used in several applications:

- Heap sort
- Dijkstra's shortest path algorithm
- Prim's algorithm

Recall also that this data structure supports the following operations*:

- **insert** a new element (key/priority+payload) into a heap
- identify the **min** element in an existing heap
- **extract-min** (identify and delete min) element in an existing heap
- **decrease-priority** of an element in an existing heap

*As with these slides, default heap operations are defined over a **min**-heap. One could alternatively define **max**, **extract-max**, **increase-priority** operations on a **max**-heap.

Mergeable heaps

Today (binomial heap) and next start of next lecture (Fibonacci heap), we will learn about mergeable heaps that support (at least) the following operations:

insert: inserts a new element into the existing heap

min: finds the min element in the heap

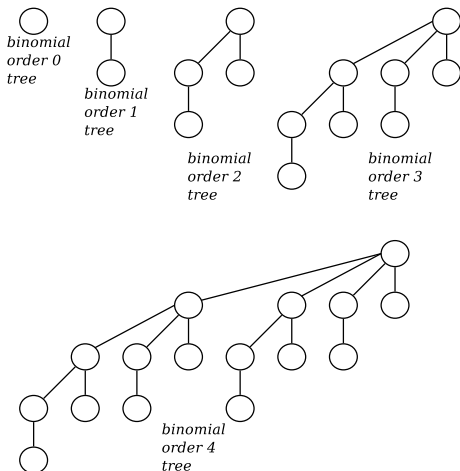
extract-min: finds and deletes the min element in the heap

merge: merges two heaps into one

decrease-priority: decreases the elements key/priority

delete: removes an element from the heap

Before Binomial **heap**, let us define a binomial **tree**



Binomial trees are defined recursively:

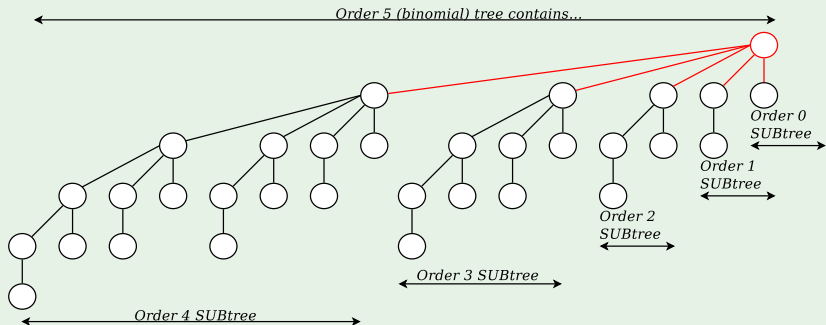
- The binomial tree of order **0** (or B_0 in short) is a single node tree
- The binomial tree of order **1** (B_1) is created from two B_0 trees, by making one B_0 tree the child of the other.
- The binomial tree of order **2** (B_2) is created from two B_1 trees, by making one B_1 tree the child of the other.
- and so on...

Properties of a Binomial tree

Any binomial tree of order k has the following properties:

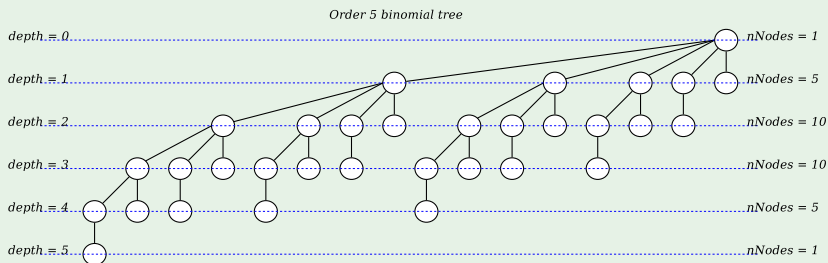
- The number of nodes in any B_k is 2^k .
- The height of any B_k is k .
- The root node of any B_k tree has k subtrees as children.
- Deleting the root node of B_k (with its edges/links) yields k independent lower order binomial trees $B_{k-1}, B_{k-2}, \dots, B_0$.

Example



Why are these trees called **binomial**?

Example: B_5



Main property

A main property of any B_k tree is that the **number of nodes** at any given depth d is given by the **binomial coefficient** $\binom{k}{d}$, that is “ k -choose- d ”

What is a binomial **heap**?

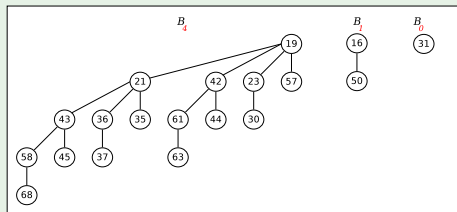
A binomial **heap** is a collection/set of binomial **trees** such that:

- **each** binomial tree in the set satisfies the heap property – i.e., **each** tree-node's key/priority is \leq its children's keys/priorities.
- There is **at most** one (i.e. either 0 or 1) binomial tree of any given order in that set.

Example

On the right is a binomial **heap** that contains a collection/set of binomial **trees**:

- one B_4 tree
- zero B_3 tree
- zero B_2 tree
- one B_1 tree
- one B_0 tree



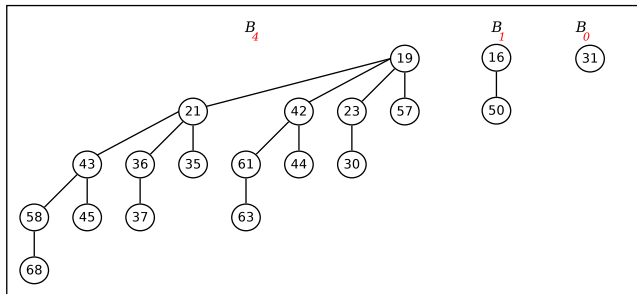
Binomial heap properties

Properties

For any binomial **heap** containing N elements, the following properties hold:

- There are at most $\lfloor \log_2 N \rfloor + 1$ binomial **trees**
- The height of each binomial **tree** is $\leq \lfloor \log_2 N \rfloor$
- The '1's in the **binary representation** of N tell us which order binomial **trees** are present in the collection forming this binomial **heap** of N elements.
- the element with **minimum** key is one of the root nodes of the **trees** in the collection.

Binomial heap properties – Example



Example

For the above binomial **heap**:

- $N = 19$.
- Number of trees is 3
- binary representation of 19 is: $\overset{4}{1} \overset{3}{0} \overset{2}{0} \overset{1}{1} \overset{0}{1}$ (therefore contains B_4, B_1, B_0)

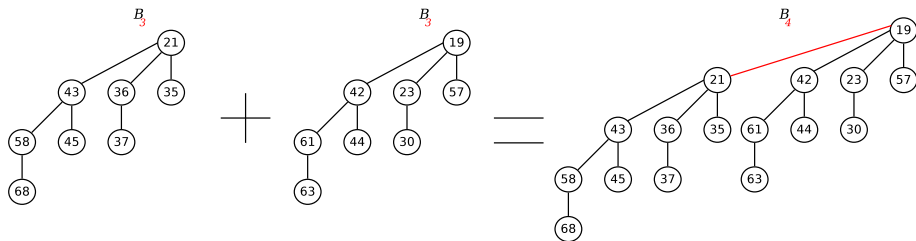
Representing a binomial heap

- Unlike **binary** heaps, **binomial** heaps are stored explicitly using a **tree** data structure.
- Each node x :
 - ▶ is denoted by a **key**,
 - ▶ has associated **payload** information
 - ▶ has a pointer **parent** $[x]$ to its parent node
 - ▶ has a pointer **child** $[x]$ to its **leftmost** child node
 - ★ If node x has zero children, then **child** $[x] = nil$
 - ▶ has a pointer **sibling** $[x]$ to the immediate **sibling** of x to its right.
 - ★ If node x is the rightmost child of its parent, then **sibling** $[x] = nil$
 - ▶ stores **degree** $[x]$ which is the number of children of x (i.e., same as the **order** of the binomial tree rooted at x)
- Finally, the roots of the binomial trees within a binomial heap are organized in a linked list, referred to as the **root list**.

operations on a binomial heap

Merging two binomial **trees** into one

- First, merging two binomial **trees**, each of the **same** order (say) k results in an order $k + 1$ binomial tree, where:
 - the two roots are linked, such that...
 - ...the root containing the **larger** key becomes the **child** of the smaller root.



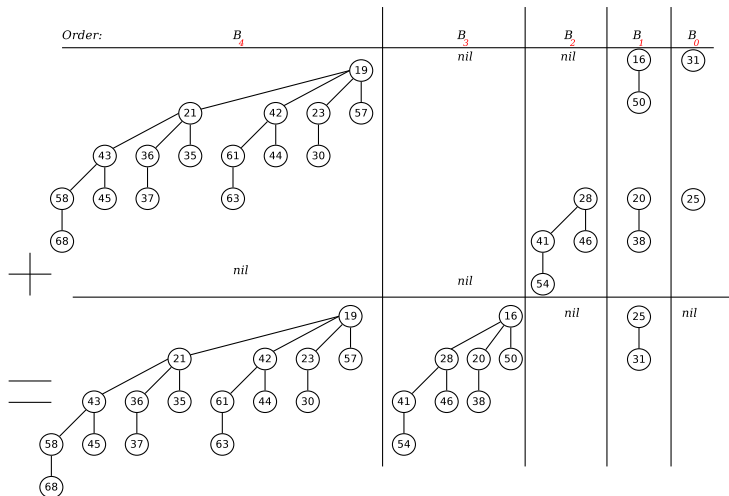
Binomial **heap** operation – **merge** two binomial **heaps** into one

- With merging of two binomial **trees** established (see previous slide), we can now define **merging** of two binomial **heaps**.
- Heaps are merged in a way that is reminiscent of how we add two numbers in binary:

Example: addition of $19 + 7 = 26$ in binary

Order:	4	3	2	1	0
carry:		1	1	1	
	1	0	0	1	1
+	0	0	1	1	1
Result:	1	1	0	1	0

Example of merging 2 binomial heaps containing 19 and 7 elements each



(To be discussed during the lecture)

Running time of **merge** operation between 2 binomial heaps

- Running time is $O(\log N)$ worst-case – **why?**
 - ▶ time is bounded by maximum number of possible merges between trees of the same order within the heaps.
 - ▶ the number of trees in each heap containing N elements is bounded by $\lfloor \log N \rfloor + 1$
 - ▶ merging two heaps in worst case requires $2 \times (\lfloor \log N \rfloor + 1)$ tree merges

Binomial **heap** operation – **extract-min**

We use this to identify and delete the minimum element among all **root nodes** of the trees in the heap.

- Identify the **min** root node among the trees in the heap.
- From slide #8, we know that deleting the root node of any B_k tree yields: $B_{k-1}, B_{k-2}, \dots, B_0$.
- If we promote these subtrees to the root level of the existing binomial heap...
- ...this might create multiple trees of the same order (violating the definition of a binomial heap – see slide #10).
- So, progressively **merge** the binomial trees of the same order (starting from 0) until the binomial heap definition is satisfied.

(Example will be handled during the lecture)

Running time of **extract-min** operation

- Running time is $O(\log N)$ worst-case – **why?**
- Effort required to find the **min** is $O(\log N)$. (see slide #11)
- Effort required to promote subtrees formed upon deletion to root level is $O(\log N)$ – the number of these subtrees is bounded by $\lfloor \log N \rfloor$.
- Effort required to merge multiple trees into a binomial heap is also $O(\log N)$. (see slide #18)
- Total effort: $O(\log N)$

Binomial **heap** operation – **decrease-priority**

We want to decrease priority of any node x in a binomial heap containing N elements. [†]

- decrease priority of node x .
- if min-heap property is violated (i.e. $x < \text{parent}[x]$), bubble up node x .
- Running time (worst-case): $O(\log N)$ – depth of the binomial tree in which x resides is bounded above by $\lfloor \log N \rfloor$

(Example will be handled during the lecture)

[†]Note: as with binary heaps, binomial (and Fibonacci) heaps are inefficient to **search** for any node x (except the root); For this reason, **decrease-priority**(x) assumes a pointer to x as part of its input.

Binomial **heap** operation – **delete**

We want to delete any node x in a binomial heap containing N elements.
‡

- run **decrease-priority** by setting x to $-\infty$.
- run **extract-min**.
- Running time (worst-case): $O(\log n)$.

‡Note: as with binary heaps, binomial (and Fibonacci) heaps are inefficient to **search** for any node x (except the root); For this reason, **delete**(x) assumes a pointer to x as part of its input.

Binomial **heap** operation – **insert**

We want to insert a new element x into an existing binomial heap H_1

- Make a new binomial heap H_2 with x as its **only** element.
- run **merge**(H_1, H_2).
- At face value, the runtime per single **insert** takes $O(\log N)$ effort.

Amortized analysis of **insert** operation

Consider the problem of building a **binomial** heap of N elements:

- From FIT2004, we know that at least a **binary** heap of N elements can be built in $O(N)$ time.
- What about a **binomial** heap then?

claim

A **binomial** heap of N elements can be built by N successive inserts in $O(N)$ -time.

Amortized analysis of **insert** operation ...continued(2)

- Time required for inserting **each** element x into a heap H_1 (starting from an empty heap) involves:
 - ▶ time to create a new binomial heap H_2 containing only 1 element x – which requires constant effort, **plus**
 - ▶ time to merge H_2 into H_1 . It isn't fully clear yet how many merges (between same-order binomial trees) will be required in each insert operation.
- Total over N insertions requires:
 - ▶ $O(N)$ **plus**
 - ▶ total merging time.

Amortized analysis of **insert** operation ...continued(3)

It is easy to see (by beholding how the numbers starting from 0 change when 1 is added each time):

- the **first insertion** into an empty H_1 heap requires **zero** merges. **Why?**
- the second insertion involves exactly **one** merge between two B_0 binomial trees, yielding a heap containing one B_1 tree.
- the **third insertion** involves **zero** merges
 - ▶ H_1 before insertion contains 2 elements (contained in 1 B_1 tree).
 - ▶ merging the new inserted element into H_1 adds only a new B_0 tree to the existing B_1 tree. Therefore no merges.
- the fourth insertion involves exactly **two** merges – **why?**
- the **fifth insertion** involves **zero** merges – **why?**
- the sixth insertion involves **zero** merges – **why?**
- \vdots

Amortized analysis of **insert** operation ...continued(3)

When inserting N elements, if the binary representation of number elements in H_1 before each insertion ends in

-**0**, the effort takes only 1 unit of time.
-**01**, the effort takes only 2 units of time.
-**011**, the effort takes only 3 units of time.
-**0111**, the effort takes only 4 units of time.
- ...**01111**, the effort takes only 5 units of time.
- \vdots

Total time over N insertions

- $T = \frac{N}{2} \times 1 + \frac{N}{4} \times 2 + \frac{N}{8} \times 3 \dots \leq 2N$
- Such series is called an **Arithmetico-Geometric series**.

Thus total time is bounded by $O(N)$, implying that each **insert** into a binomial heap is $O(1)$ amortized!

Summary

Operation	Binary heap	Binomial heap
make-new-heap	$O(1)$	$O(1)$
min	$O(1)$	$O(\log N)$
extract-min	$O(\log N)$	$O(\log N)$
merge	$O(N)$	$O(\log N)$
decrease-priority	$O(\log N)$	$O(\log N)$
delete	$O(\log N)$	$O(\log N)$
insert	$O(\log N)$ worst-case $O(1)$ amortized	$O(\log N)$ worst-case $O(1)$ amortized

In the next lecture...

(1) Fibonacci heaps and amortized analysis (2) B-Trees

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END

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