# Lecture 3: Mergeable Heaps 

Binomial heaps

October 18, 2019

## Mergeable heaps

Data structures designed to support well the following operations:

- makeBinomialHeap(): creates and returns a new empty heap.
- insert $(H, x)$ : inserts node referred by $x$, whose key field has already been filled in, into heap $H$.
- minimum $(H)$ : returns a pointer to the node with minimum key in heap $H$.
- extractMin $(H)$ : deletes the node from heap $H$ whose key is minimum, returning a pointer to the node.
- union $\left(H_{1}, H_{2}\right)$ : creates and returns a new heap that contains all the nodes of heaps $H_{1}$ and $H_{2}$. Heaps $H_{1}$ and $H_{2}$ are "destroyed" by this operation.
and also
- DecreaseKey $(H, x, k)$ assigns to node referred by $x$ within heap $H$ the new key value $k$. It is assumed that key $\leq x$->key.
- Delete $(H, x)$ deletes node referred by $x$ from heap $H$.


## Content of the lecture

In this lecture we will study binomial heaps. Later on, after we will explain the notion of amortized time bounds of algorithms, we will study Fibonacci heaps. If we also add binary heaps to the picture, we will be able to draw a comparison table of their running times:

|  | Binary heap <br> (worst-case) | Binomial heap <br> (worst case) | Fibonacci heap <br> (amortized) |
| :--- | :--- | :--- | :--- |
| makeBinomialHeap | $\Theta(1)$ | $\Theta(1)$ | $\Theta(1)$ |
| insert | $\Theta(\log n)$ | $O(\log n)$ | $\Theta(1)$ |
| minimum | $\Theta(1)$ | $O(\log n)$ | $\Theta(1)$ |
| extractMin | $\Theta(\log n)$ | $\Theta(\log n)$ | $O(\log n)$ |
| union | $\Theta(\log n)$ | $O(\log n)$ | $\Theta(1)$ |
| DecreaseKey | $\Theta(\log n)$ | $\Theta(\log n)$ | $\Theta(1)$ |
| Delete | $\Theta(\log n)$ | $\Theta(\log n)$ | $\Theta(\log n)$ |

Note. The notion of amortized time bounds will be explained in another lecture.

## Binomial trees

A binomial tree is an element of the set of ordered trees $\left\{B_{k} \mid k \in \mathbb{N}\right\}$ defined recursively as follows:

- $B_{0}$ consists of a single node.
- $B_{k}$ consists of two binomial trees $B_{k-1}$ that are linked such that one of them has the other one as left child of its root.

Example $\left(B_{0}, B_{1}, B_{2}, B_{3}\right.$, and $\left.B_{4}\right)$

$B_{0} \quad B_{1}$

$B_{2}$

$B_{3}$

$B_{4}$

## The structure of binomial trees

- The recursive view (Cf. the definition)

- Another view



## Properties of binomial trees

## Lemma 1

The binomial tree $B_{k}$ has the following properties:
(1) It has $2^{k}$ nodes.
(2) It has height $k$.
(3) There are exactly $\binom{k}{i}$ nodes at depth $i$ for $i=0,1, \ldots, k$.
(4) The root has degree $k$, which is greater than that of any other node. Moreover, if the children of the root are numbered from left to right by $k-1, k-2, \ldots, 0$, then child $i$ is the root of a subtree $B_{i}$.

## Binomial heaps

A binomial heap $H$ is a set of binomial trees that satisfies the following binomial heap properties:
(1) Each binomial tree in $H$ is heap-ordered: the key of a node is greater than or equal to the key of its parent.
(2) There is at most one binomial tree in $H$ whose root has a given degree.

## Binomial heaps

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(2) There is at most one binomial tree in H whose root has a given degree.

## Consequences of the definition

- The root of a heap-ordered tree contains the smallest key in the tree.
- The second property implied that an n-node binomial heap $H$ has at most $\left\lfloor\log _{2} n\right\rfloor+1$ binomial trees. This is so because:
- The binary representation of $n$ has $\leq\left\lfloor\log _{2} n\right\rfloor+1$ bits, say $b_{\left\lfloor\log _{2} n\right\rfloor}, b_{\left\lfloor\log _{2} n\right\rfloor-1}, \ldots, b_{0}$, so that $n=\sum_{i=0}^{\left\lfloor\log _{2} n\right\rfloor} b_{i} \cdot 2^{i}$. From Lemma 1 we learn that $B_{i}$ appears in $H$ if and only if $b_{i}=1$. Thus, $H$ contains at most $\left\lfloor\log _{2} n\right\rfloor+1$ binomial trees.


## Representing binomial heaps

- Binomial trees are stored in the left-child, right-sibling representation.
- Each node $x$ contains the following pointers:
- x.p: pointer to its parent. It is nil if $x$ is a root.
- $x$.child: pointer to its left child. It is nil if $x$ has no children.
- $x$.sibling: pointer to its right sibling. It is nil if $x$ is the rightmost child of its parent.
and the fields
- x.key: the key of the node.
- possibly other fields for the satellite information required by the application.
- $x$.degree: the number of children of $x$.
- The roots of the binomial trees of a binomial heap $H$ for a linked list of nodes, connected through their sibling pointers. This list is called called the root list of the heap. The degrees of the roots strictly increase as we traverse the root list.
- A binomial heap $H$ is accessed by the field $H$.head, which is a pointer to the first node in the root list of $H$.


## Binomial heaps



A binomial heap $H$ with $n=13$ nodes, made of trees $B_{0}, B_{2}$, and $B_{3}$. The key of any node is $\geq$ the key of its parent.
The binomial heap is represented as a linked list of the roots of the trees it contains, in order of increasing degree.


A more detailed view of the illustrated heap.

## Binomial heaps

In a binomial heap with $n$ nodes, the degrees of the roots of its binomial trees are a subset of $\left\{0,1, \ldots,\left\lfloor\log _{2} n\right\rfloor\right\}$.

## Operations on binomial heaps (1)

- makeBinomialHeap() creates an empty binomial heap $h$ by allocating space for $H$ and setting $H$.head $=$ NIL. The running time is, obviously, $\Theta(1)$.
- minimum $(H)$ returns a pointer to the node with the minimum key in an $n$-node heap. It is assumed that there are no keys with value $\infty$ in $H$.
- Binomial heaps are heap-ordered $\Rightarrow$ the minimum key is in a node in the root list of the heap.
- The root list had length $\leq\left\lfloor\log _{2} n\right\rfloor+1 \Rightarrow$ the running time of this operation is $O\left(\log _{2} n\right)$.

```
minimum(H)
1y:= NIL
2x:= H.head
3 min := \infty
4 while }x\not=\mathrm{ NIL
if ix>key < min
6 min =x->key
7 y:=x
8 x := x-> sibling
9 return y
```


## Operations on binomial heaps (2)

union $\left(H_{1}, H_{2}\right)$ performs the union of binomial heaps $H_{1}, H_{2}$ by linking repeatedly binomial trees whose roots have the same degree.

- If $y$ and $z$ are roots of two $B_{k-1}$-trees, then BinomialLink $(y, z)$ makes a $B_{k}$-tree with root $z$ by setting the parent of $y$ to be $z$, the leftmost child of $z$ to be $y$, and increasing the degree of $z$ by 1 :

```
BinomialLink \((y, z)\)
\(1 y->p:=z\)
\(2 y\)->sibling := \(z\)->child
3 z->child :=y
\(4 z\)->degree := z->degree + 1
```

BinomialHeapMerge $\left(H_{1}, H_{2}\right)$ merges the root lists of binomial heaps $H_{1}$ and $H_{2}$ into a single linked list that is sorted by degree into monotonically increasing order.

## Operations on binomial heaps (2)

```
union \(\left(H_{1}, H_{2}\right)\)
    \(1 H:=\) makeBinomialHeap()
    2 H.head = BinomialHeapMerge \(\left(H_{1}, H_{2}\right)\)
    3 free the objects \(H_{1}\) and \(H_{2}\), but not the lists they point to
    4 if \(H\).head \(=\) NIL
    5 return \(H\)
    6 prev_x \(:=\) Nil
    \(7 x:=\) H.head
    8 next_x :=x->sibling
    9 while next_ \(x \neq\) NIL
10 if \(x\)->degree \(\neq\) next_ \(x\)->degree or
    (next_x->sibling \(\neq\) NIL and next_x->sibling->degree \(=x->\) degree \()\)
        prev_x \(:=x\)
                            // Cases 1 and 2
                \(x:=\) next_x
                            // Cases 1 and 2
    else if \(x->\) key \(\leq n e x t \_x->k e y\)
        \(x->\) sibling \(:=\) next_x->sibling // Case 3
        BinomialLink(next_x,x) // Case 3
            else if prev_ \(x=\) NIL \(\quad / /\) Case 4
                H.head := next_x // Case 4
            else prev_x->sibling := next_x // Case 4
            BinomialLink \((x\), next_x) // Case 4
            \(x:=\) next_ \(x \quad\) // Case 4
            next_x \(:=x->\) sibling
    return \(H\)
```

It works in 2 phases:
(1) First, BinomialHeapMerge $\left(H_{1}, H_{2}\right)$ merges the root lists of $H_{1}$ and $H_{2}$ into a single list $H$ that is sorted by degree into monotonically increasing order.
(2) There might be at most 2 roots of each degree $\Rightarrow$ the second phase links roots of equal degree until at most one root remains of each degree.
The meaning of the local pointers used in the algorithm is:

- $x$ points to the root currently being examined.
- prev_x points to the root preceding $x$ on the root list, thus prev_ $x->$ sibling $=x$.
- next_ $x$ points to the root following $x$ on the root list, thus $x->$ sibling $=$ next_x.

There are four cases, illustrated below, where $I>k$.
Labels $a, b, c, d$ serve only to identify the roots involved; they do not indicate the degrees or keys of these roots.

Case 1: $x$->degree $\neq$ next_ $x$->degree.


Case 2: $x$->degree $=$ next_x->degree $=$ next_x->sibling->degree.


## union <br> A study by cases (continued)

Case 3: $x$->degree $=$ next_ $x$->degree $\neq n_{\text {next_ }}$ x->sibling $->$ degree and $x->k e y \leq n e x t \_x->k e y$.


Case 4: $x$->degree $=$ next_ $x$->degree $\neq$ next_ $x->$ sibling $->$ degree and $x->k e y>n e x t \_x->k e y$.


## Execution of union

Running example

(b) head $[H] \rightarrow$ (12)
(c) head $[H] \rightarrow$ (12)

Case 2


## Execution of union

Running example



$$
\text { Case } 3
$$

(45) 32 24

50


## Run time complexity of union

Assumptions. $H_{1}$ contains $n_{1}$ nodes, $H_{2}$ contains $n_{2}$ nodes, and let $n=n_{1}+n_{2}$.
Then

- $H_{1}$ contains $\leq\left\lfloor\log _{2} n_{1}\right\rfloor+1$ roots, and $H_{2}$ contains $\leq\left\lfloor\log _{2} n_{2}\right\rfloor+1$ roots
- $\Rightarrow H$ contains at most
$\left\lfloor\log _{2} n_{1}\right\rfloor+\left\lfloor\log _{2} n_{2}\right\rfloor+2 \leq 2\left\lfloor\log _{2} n\right\rfloor+2=O\left(\log _{2} n\right)$ roots immediately after the call of BinomialHeapMerge.
- BinomialHeapMerge $\left(H_{1}, H_{2}\right)$ takes $O\left(\log _{2} n\right)$ time. Each iteration of the while loop takes $O(1)$ time, and there are at most $\left\lfloor\log _{2} n_{1}\right\rfloor+\left\lfloor\log _{2} n_{2}\right\rfloor+2$ iterations because each iteration either advances the pointers one position down the root list or removes a root from the root list.
- $\Rightarrow$ total run time is $O\left(\log _{2} n\right)$.


## Inserting a node

Assumptions: Node $x$ has already been allocated, and key $x->k e y$ has already been filled in.
insert $(H, x)$
$1 H^{\prime}:=$ makeBinomialHeap()
$2 x->p:=\mathrm{NIL}$
$3 x->$ child $:=$ NIL
$4 x->$ sibling $:=$ NIL
$5 x->$ degee $:=0$
$6 H^{\prime}$.head $=x$
$7 H:=$ union $\left(H, H^{\prime}\right)$

## Extracting the node with minimum key

extractMin(H)
1 Find the root $x$ with the minimum key in the root list of $H$ and remove $x$ from the root list of $H$
$2 H^{\prime}:=$ makeBinomialHeap()
3 Reverse the order of the linked list of $x$ 's children, and set $H^{\prime}$.head to point to the head of the resulting list
$4 H:=$ union $\left(H, H^{\prime}\right)$
5 return $x$

## Extracting the node with minimum key

## Example


(b) head $[\mathrm{H}] \rightarrow$ (37)


## Extracting the node with minimum key

## Example continued

(c) head $[\mathrm{H}] \rightarrow$ (37)
(d) head $[H] \rightarrow 25$
(41)



## Decreasing a key

DecreaseKey $(H, x, k)$ decreases the key of a node $x$ in a binomial heap to a new value $k$. It signals an error if $k>x$->key.

```
DecreaseKey \((H, x, k)\)
    1 if \(k>x->\) key
    2 error "new key is greater than current key"
    \(3 x\)->key :=k
    \(4 y:=x\)
    \(5 z:=y->p\)
    6 while \(z \neq\) NIL and \(y\)->key < z->key
    7 exchange \(y\)->key \(\leftrightarrow z->k e y\)
    8 if \(y\) and \(z\) have satellite fields, exchange them, too
    \(9 \quad y:=z\)
\(10 \quad z:=y->p\)
```


## DecreaseKey $(H, x, k)$

## Example

Let's decrease the key of node $x$ in $H$ to $k=7$.

( a ) H.head $\longrightarrow$ (25)


## DecreaseKey $(H, x, k)$

## Example (continued)



## DecreaseKey $(H, x, k)$

## Remarks about the implementation

The decreased key "bubbles up" in the heap:

- After ensuring that $k \leq x->$ key and then assigning key $k$ to $x$, the procedure goes up the tree, with $y$ initially pointing to $x$.
- In each iteration of the while loop of lines 6-10, $y$->key is checked against the key of $y$ 's parent $z$ :
- If $y$ is the root or $y$->key $\geq z$->key, the tree is heap-ordered.
- Otherwise, node $y$ violates the heap ordering, therefore its key is exchanged with the key of its parent $z$, along with any other satellite information.

The procedure then sets $y$ to $z$, going up one level in the tree, and continues with the next iteration.

- The time complexity of DecreaseKey $(H, x, k)$ is $O\left(\log _{2} n\right)$ because the maximum depth of $x$ is $\left\lfloor\log _{2} n\right\rfloor$, so the while loop iterates at most $\left\lfloor\log _{2} n\right\rfloor$ times.


## Deleting a node

Deleting a node $x$ from a binomial heap $H$ is trivial:

- First, decrease the key of $x$ to a value smaller than any key in $H$, e.g., $-\infty$.
- Next, extract from $H$ the node with minimal key, which is $x$ with key $-\infty$.

```
Delete( \(H, x\) )
1 DecreaseKey \((H, x,-\infty)\)
2 extractMin(H)
```

- If $H$ has $n$ nodes, then Delete $(H, x)$ takes $O\left(\log _{2} n\right)$ time.


## Fibonacci heaps

## Structure of Fibonacci heaps (I)

A Fibonacci heap is a collection of heap-ordered trees. The trees are rooted, but unordered:

- Each node $x$ contains
- the fields
- x.key and possibly more fields for satellite data associated with the key.
- $x$.degree: number of children in the child list of node $x$.
- x.mark: a boolean value, which indicates whether node $x$ has lost a child since the last time $x$ was made the child of another node. Newly created nodes are unmarked.
- the pointers
- x.p: pointer to the parent node; NIL if $x$ is a root node.
- x.child: pointer to any one of its children.
- The children of a node $x$ are linked together in a circular, doubly-linked list, called the child list of $x$.
- Each node $n$ in a child list has pointers $n$.left and n.right which point to its left and right siblings, respectively. If $y$ is a reference to the only child, then $y->$ left $=y->$ right $=y$.
- The order in which children appear in the child list is arbitrary.


## Fibonacci heaps

## Structure of Fibonacci heaps (II)

- The roots of all trees in a Fibonacci heap $H$ are linked together into a circular, doubly linked list, using the left and right fields of the root nodes. This circular list is called the root list of the Fibonacci heap.
- The order of the trees in the root list is arbitrary.
- A Fibonacci heap $H$ has 2 fields:
- H.min: a pointer to the root of the tree in $H$ with minimum key; this node is called the minimum node of the Fibonacci heap.
- H.n: the number of nodes currently in $H$.


## Fibonacci heaps

## Structure of Fibonacci heaps (III)



A Fibonacci heap made of 5 heap-ordered trees. The dashed line indicates the root list. The 3 marked nodes are blackened.

A more complete representation of the previous heap, showing the pointers $p$ (up arrows), child (down arrows), and left and right (sideways arrows), is illustrated below.


## Fibonacci heaps

## The potential function

For a Fibonacci heap $H$, we define:

- $t(H)$ : the number of trees in the root list of $H$
- $m(H)$ : the number of marked nodes in $H$.
- $\Phi(H):=t(H)+2 \cdot m(H)$ is called the potential of $H$.
- The potential of a set of Fibonacci heaps is the sum of the potentials of its constituent Fibonacci heaps.


## Example

The Fibonacci heap illustrated before has $t(H)=5$ and $m(H)=3$. Thus, the potential of $H$ is

$$
\Phi(H)=5+2 \cdot 3=5+6=11 .
$$

- $\Phi$ will be used for amortized analysis with the potential method (See Lecture 7.)
- 1 unit of potential can pay for a constant amount of work, where the constant is sufficiently large to cover the cost of any specific constant-time piece of work that we might encounter.
- The Fibonacci heap is initially empty $\Rightarrow$ initial potential is 0 .
- By definition, $\Phi(H) \geq 0$ always holds $\Rightarrow$ for a sequence of heap operations, the total amortized cost is an upper bound of the total actual cost.
- $D(n)$ is an upper bound on the maximum degree of any node in an $n$-node Fibonacci heap.


## Mergeable heap operations

Assumption. Only makeBinomialHeap, Insert, Minumum, EXTRACTMIN and BINOMIALHEAPUNION are supported. $\Rightarrow$
Fibonacci heap = collection of unordered binomial trees, that is, elements of the set $\left\{U_{n} \mid n \in \mathbb{N}\right\}$ defined recursively as follows:

- $U_{0}$ has a single node.
- $U_{k}$ is obtained from two trees $U_{k-1}$, for which the root of one is made into any child of the root of the other.


## Properties of unordered binomial trees

For the unordered binomial tree $U_{k}$,
(1) there are $2^{k}$ nodes,
(2) the height is $k$,
(3) there are exactly $\binom{k}{i}$ nodes at depth $i$ for $i=0,1, \ldots, k$,
(4) the root has degree $k$, which is greater than that of any other node. The children of the root are roots of subtrees $U_{0}, U_{1}, \ldots, U_{k-1}$ in some order.

## Fibonacci heaps

## Properties

If an $n$-node Fibonacci heap is made of unordered binomial trees, then $D(n)=\log _{2} n$.

Proof. Obvious.
Main idea. To make mergeable-heap operations performant, delay their work as long as possible.

## Creating a heap

## MakeFibHEAP

MakeFibHeap allocates and returns the Fibonacci heap object $H$ with $H . n=0$ and $H . \min =$ NIL.
$\triangleright \Phi(H)=0$
$\triangleright$ For MakeBinHeap: amortized cost $=$ actual cost $=O(1)$.

## Inserting a node

FibHeaplnsert $(H, x)$
$1 x$->degree :=0
$2 x->p:=\mathrm{NIL}$
$3 x->$ child $:=$ NIL
$4 x->$ left $:=x$
$5 x->$ right $:=x$
$6 x->$ mark := false
7 concatenate the root list containing $x$ with the root list $H$
8 if H.min $=$ NIL or $x->$ key $<$ (H.min) $->$ key
9 H.min $=x$
10 H.n:=H.n+1
Note. Unlike BinomialHeaplnsert, FibHeaplnsert does not attempt to merge trees within the Fibonacci heap.

## Inserting a node

Illustrated example. The amortized cost

(a)

(b)
(a) A Fibonacci heap $H$. (b) Fibonacci heap $H^{\prime}$ produced after inserting the node with key 21 in $H$. The node becomes its own heap-ordered tree and is then added to the root list, becoming the left sibling of the root.

$$
\left.\begin{array}{r}
t\left(H^{\prime}\right)=t(H)+1 \\
m\left(H^{\prime}\right)=m(H)
\end{array}\right\} \Rightarrow \begin{aligned}
& \text { the increase in potential is } \\
& ((t(H)+1+2 m(H))-(t(H)+2 m(H))=1
\end{aligned}
$$

Actual cost of FibHeaplnsert is $O(1) \Rightarrow$ Amortized cost of FibHeaplnsert is $O(1)+1=O(1)$.

## Finding the minimum node

FibHeapminimum

H.min points to the minimum node of $H \Rightarrow$ finding it takes $O(1)$ actual time.

- $\Phi(H)$ does not change $\Rightarrow$ the amortized cost of FibHeapminimum is $O(1)$.


## Merging (or uniting) Fibonacci heaps

FibHeapbinomialHeapUnion $\left(H_{1}, H_{2}\right)$
$1 \mathrm{H}:=$ MakeFibHeap()
2 H.min := $H_{1}$. min
3 concatenate the root list of $\mathrm{H}_{2}$ with the root list of H
4 if $\left(H_{1} \cdot \min =\mathrm{NIL}\right)$ or $\left(H_{2} \cdot \min \neq \mathrm{NIL}\right.$ and $\left.H_{2} \cdot \min <H_{1} \cdot \min \right)$
5 H.min := $\mathrm{H}_{2}$. min
6 H.n := $H_{1} \cdot n+H_{2} . n$
7 free the objects $\mathrm{H}_{1}$ and $\mathrm{H}_{2}$
8 return $H$
Note. No consolidation of trees occurs in the Fibonacci heap.

## Merging (or uniting) Fibonacci heaps

The amortized cost of FIbHEAPBINOMIALHEAPUNION

- The change in potential is

$$
\begin{aligned}
\Phi(H)- & \left(\Phi\left(H_{1}\right)+\Phi\left(H_{2}\right)\right) \\
= & (t(H)+2 m(H))-\left(\left(t\left(H_{1}\right)+2 m\left(H_{1}\right)\right)\right. \\
& +\left(\left(t\left(H_{2}\right)+2 m\left(H_{2}\right)\right)\right. \\
= & 0
\end{aligned}
$$

because $t(H)=t\left(H_{1}\right)+t\left(H_{2}\right)$ and $m(H)=m\left(H_{1}\right)+m\left(H_{2}\right)$
$\Rightarrow$ amortized cost of FibHEAPBINOMIALHEAPUNION = actual cost $=O(1)$.

## Extracting the minimum node

## FibHeapextractMin

Preliminary remark. This is the operation where all work delayed by other operations is done.

- delayed work = consolidation (or merging) of the trees in the root list.

FibHeapextractMin(H)
$1 z:=$ H.min
2 if $z \neq$ NIL
3 for each child $x$ of $z$ add $x$ to the root list of $H$ $x->p:=$ NIL
remove $z$ from the root list of $H$
if $z==z$-> right
H.min $=$ NIL
else H.min = z-> right
Consolidate(H)
H.n:=H.n-1

12 return $z$

## Extracting the minimum node

- FibHeapextractMin makes a root out of each of the minimum node-s children and removes the minimum node from the root list.
- Next, it consolidates the root list by linking roots of equal degree until at most one root remains of each degree.
- Consolidate $(H)$ consolidates the root list of $H$ by executing repeatedly the following steps until every root in the root list has a distinct degree value:
(1) Find two roots $x$ and $y$ from the root list with the same degree, and with $x->k e y \leq y->k e y$.
(2) Link $y$ to $x$ : remove $y$ from the root list, and make $y$ a child of $x$. This operation is performed by FibHEAPLInk.
Consolidate $(H)$ uses an auxiliary array $A[0 . . D(H . n)]$; if $A[i]=y$ then $y$ is currently a root with $y->$ degree $=i$.


## Extracting the minimum node

## The Consolidate procedure

Consolidate $(H)$

1. for $i:=0$ to $D$ (H.n)
2. $A[i]=\mathrm{NIL}$
3. for each node $w$ in the root list of $H$
4. $x:=w$
5. $d:=x->$ degree
6. while $A[d] \neq$ NIL
7. $y:=A[d]$
8. if $x->$ key $>y->$ key
$9 \quad$ exchange $x \leftrightarrow y$
9. $\quad \operatorname{FibHeaplink}(H, y, x)$
10. $\quad A[d]:=\mathrm{NIL}$
11. $d:=d+1$
12. $A[d]:=x$
13. H.min $:=\mathrm{NIL}$
14. for $i:=0$ to $D$ (H.n)
15. if $A[i] \neq$ NIL
16. add $A[i]$ to the root list of $H$
17. if H.min $=$ NiL or $A[i]->k e y<H . m i n->k e y$
18. H.min := A[i]

## Extracting the minimum node

FibHEApLINK

FibHeapLink( $H, y, x)$

1. remove $y$ form the root list of $H$
2. make $y$ a child of $x$, incrementing $x->$ degree
3. $y$->mark := false

## Extracting the minimum node <br> \section*{Illustrated example}

(a)

(b)

(c)

(d) (23) - (21)

(b) The situation after the minimum node $z$ is removed from the root list and its children are added to the root list.
(c)-(d) The array $A$ and the trees after each of the first 2 iterations of the for loop of lines 3-13 of Consolidate.

## Extracting the minimum node

## Illustrated example continued


(e) The array $A$ and the trees after each of the 3rd iteration of the for loop of lines 3-13 of CONSOLIDATE. (f)-(h) The next iteration of the for loop, with the values of $w$ and $x$ shown at the end of each iteration of the while loop of lines 6-12. Part (f) shows the situation after the first time through the while loop. The node with key 23 has been linked to the node with key 7 , which is now pointed to by $x$. In part (g), the node with key 17 has been linked to the node with key 7 , which is still pointed to by $x$.

## Extracting the minimum node

## Illustrated example continued


(i)-(I) The situation after each of the next four iterations of the while loop.

## Extracting the minimum node

## Illustrated example continued


(m) Fibonacci heap after reconstruction of the root list from the array $A$ and determination of the new H.min pointer.

## Extracting the minimum node

## The amortized cost (1)

Let $H$ be the Fibonacci heap before the call of
FibHeapextractMin(H).

- FibHeapextractMin $(H)$ contributes $O(D(n))$ from the extraction of at most $D(n)$ children of the minimum node that are processed in FibHeapextractMin and from the work done in lines 1-2 and 14-19 of Consolidate.
- It remains to analyze the contribution of the for loop of lines 3-13.
- When Consolidate is called, the root list has size $\leq D(n)+t(H)-1$.
- Every while loop of lines 6-12 links one root to another $\Rightarrow$ the amount of work performed in the while loop is $\leq D(n)+t(H)$.
$\Rightarrow$ total actual work is $O(D(n)+t(H))$.


## Extracting the minimum node

## The amortized cost (2)

- The potential before extracting the minimum node is $t(H)+2 m(H)$.
- At most $D(n)+1$ roots remain and no nodes become marked during the operation $\Rightarrow$ the potential after extracting the minimum node is $\leq(D(n)+1)+2 m(H)$.
$\Rightarrow$ the amortized cost is at most

$$
\begin{aligned}
O(D(n)+t(H)) & +((D(n)+1)+2 m(H))-(t(H)+2 m(H)) \\
& =O(D(n))+O(t(H))-t(H) \\
& =O(D(n))
\end{aligned}
$$

because the units of potential can be scaled to dominate the constant hidden in $O(t(H))$.

## Decreasing a key and deleting a node

- The operations presented so far for Fibonacci heaps did preserve the property that all trees in the Fibonacci heap are unordered binomial trees $U_{n}$.
- The operations that will be presented do not preserve this property.


## Decreasing a key

FibHeapDecreaseKey $(H, x, k)$ decreases the key of a node $x$ in a binomial heap to a new value $k$. It signals an error if $k>x->k e y$.

FibHeapDecreaseKey $(H, x, k)$
1 if $k>x->k e y$
2 error "new key is greater than current key"
$3 x->k e y:=k$
$4 y:=x->p$
5 if $y \neq$ NIL and $x->$ key $<y->$ key
$6 \operatorname{CuT}(H, x, y)$
7 CASCADINGCUT( $H, y$ )
8 if $x->$ key $<$ H.min $->$ key
9 H.min :=x
$\operatorname{CuT}(H, x, y)$
1 remove $x$ from the child list of $y$, decrementing $y$->degree
2 add $x$ to the root list of $H$
$3 x->p:=$ NIL
$4 x->$ mark := false

## Decreasing a key (2)

```
CascadingCut( \(H, y\) )
    \(1 z:=y->p\)
    2 if \(z \neq\) NIL
    3 if \(y\)->mark \(=\) false
    \(4 \quad y\)->mark := true
    5 else Cut \((H, y, z)\)
    6
    CascadingCut( \(H, z\) )
```


## Decreasing a key

## Comments on the implementation

Lines 1-3 of FIbHEAPDECREASEKEY ensure that new key should be < current key. If $x$ is a root (that is, $x->p=$ NIL) or else $x->k e y \geq x->p->k e y$, then no structural changes need occur because the heap order is preserved by the key replacement. If the heap order is violated $\Rightarrow x$ is cut out from its siblings in line 6 , and moved into the root list of the heap.
The purpose of the mark fields is to ensure short time bounds for the heap operations. To understand how it is used, let's consider that $x$ is a pointer to a node that went through the following situations:
(1) At some point, $x$ was a root.
(2) Later on, $x$ was linked to another node.
(3) Later on, two children of $x$ were removed by cuts.

When the 2nd child is cut, $x$ is cut from its parent and becomes a new root. We have $x->$ mark $=$ true if steps 1 and 2 happened and one child of $x$ has been cut. Thus, Cut sets $x$->mark $=$ false in line 4 because it performs step 1.
The CascadingCut calls in line 7 of FibHEApDecreasekey take care of the situation when $x$ might be the second child cut from its parent $y$ since the time that $y$ was linked to another node. It recurses up the tree until either a root or an unmarked node is found. After all cascading cuts were done, lines 8-9 of FIbHEAPDECREASEKEY update the value of $H$.min accordingly.

## Decreasing a key: Example


(b)

(d)

(e)

(a) The initial Fibonacci heap. (b) key 46 was decreased to 15. (3) The node becomes a root, and its parent (with key 24) gets marked. (c)-(e) the node with key 35 has its key decreased to 5 . Its parent (key 26) is marked, and a cascading cut occurs. The node with key 26 is cut from its parent and becomes an unmarked root in (d). Another cascading occurs since node with key 24 is marked too. This node gets cut from its parent and made an unmarked root in (e). At this stage, cascading stops.

- Actual cost $=O(1)+$ time for cascading cuts $=O(c)$ where $c=$ max. number of recursive calls of CascadingCut from a call of FibHeapDecreaseKey
- Change in potential during a FibHeapDecreaseKey operation is at most (see References)

$$
((t(H)+c)+2(m(H)-c+2))-(t(H)+2 m(H))=4-c
$$

$\Rightarrow$ the amortized cost of FibHeapDecreaseKey is at most $O(c)+4-c=O(1)$

## Deleting a node

FibHeapDelete $(H, x)$
1 FibHeapDecreaseKey $(H, x,-\infty)$
2 FibHeapextractMin( $H$ )

- The amortized execution time of $\operatorname{FibHeapDelete}(H, x)$ is the sum of the amortized time $O(1)$ to perform
FibHeapDecreaseKey $(H, x,-\infty)$, with the amortized time $O(D(n))$ to perform FibHeapextractMin $(H)$.


## Bounding the maximum degree

- The last thing to do is to compute an upper bound $D(n)$ for the maximal degree of the unordered trees in the Fibonacci heap.
- We noticed that if all trees in the Fibonacci heap are unordered binomial trees, then $D(n)=\left\lfloor\log _{2} n\right\rfloor$. But the (cascading) cuts may cause the occurrence of trees that are not unordered binomial.
- Therefore, a slightly weaker result still holds: $D(n) \leq\left\lfloor\log _{\phi}(n)\right\rfloor$ where $\phi=(1+\sqrt{5}) / 2$.
- For a proof of this result, see Chapter 21 of the book Introduction to Algorithms by Cormen et al.

