

# **CSE 548: Analysis of Algorithms**

## **Lecture 9 ( Binomial Heaps )**

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# Mergeable Heap Operations

**MAKE-HEAP(  $x$  )**: return a new heap containing only element  $x$

**INSERT(  $H, x$  )**: insert element  $x$  into heap  $H$

**MINIMUM(  $H$  )**: return a pointer to an element in  $H$  containing the smallest key

**EXTRACT-MIN(  $H$  )**: delete an element with the smallest key from  $H$  and return a pointer to that element

**UNION(  $H_1, H_2$  )**: return a new heap containing all elements of heaps  $H_1$  and  $H_2$ , and destroy the input heaps

More mergeable heap operations:

**DECREASE-KEY(  $H, x, k$  )**: change the key of element  $x$  of heap  $H$  to  $k$  assuming  $k \leq$  the current key of  $x$

**DELETE(  $H, x$  )**: delete element  $x$  from heap  $H$

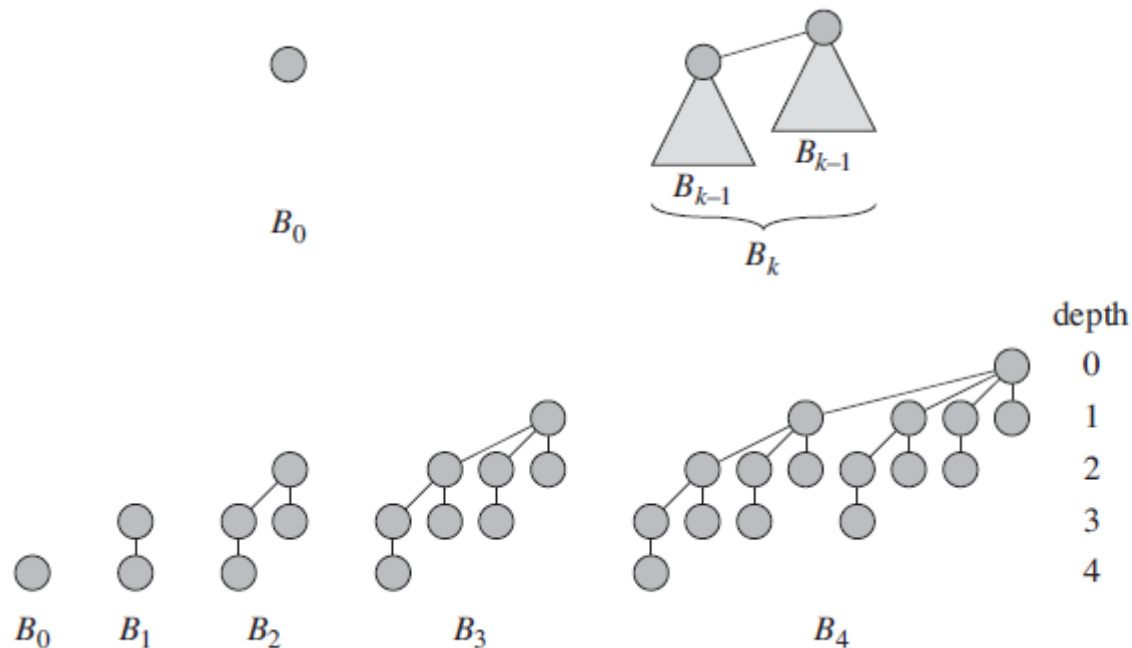
# Mergeable Heap Operations

Heap Operation	Binary Heap ( worst-case )	Binomial Heap ( amortized )
MAKE-HEAP	$\Theta(1)$	$\Theta(1)$
INSERT	$O(\log n)$	$\Theta(1)$
MINIMUM	$\Theta(1)$	$\Theta(1)$
EXTRACT-MIN	$O(\log n)$	$O(\log n)$
UNION	$\Theta(n)$	$\Theta(1)$
DECREASE-KEY	$O(\log n)$	—
DELETE	$O(\log n)$	—

# Binomial Trees

A binomial tree  $B_k$  is an ordered tree defined recursively as follows.

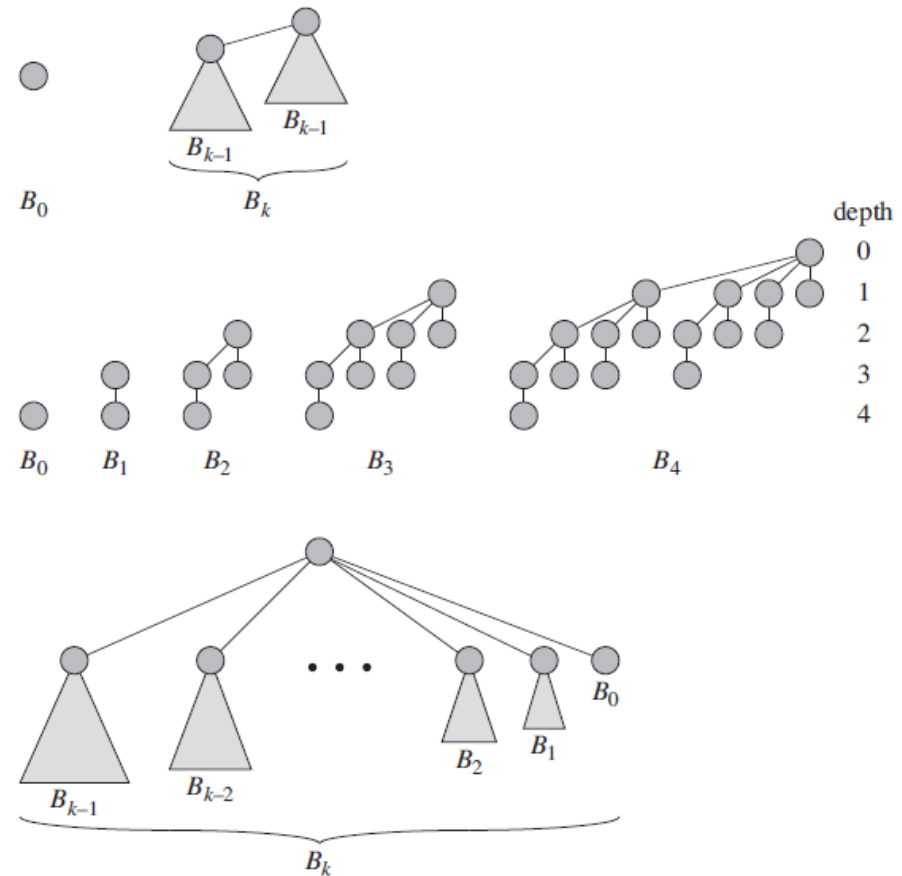
- $B_0$  consists of a single node
- For  $k > 0$ ,  $B_k$  consists of two  $B_{k-1}$ 's that are linked together so that the root of one is the left child of the root of the other



# Binomial Trees

Some useful properties of  $B_k$  are as follows.

1. it has exactly  $2^k$  nodes
2. its height is  $k$
3. there are exactly  $\binom{k}{i}$  nodes at depth  $i = 0, 1, 2, \dots, k$
4. the root has degree  $k$
5. if the children of the root are numbered from left to right by  $k - 1, k - 2, \dots, 0$ , then child  $i$  is the root of a  $B_i$

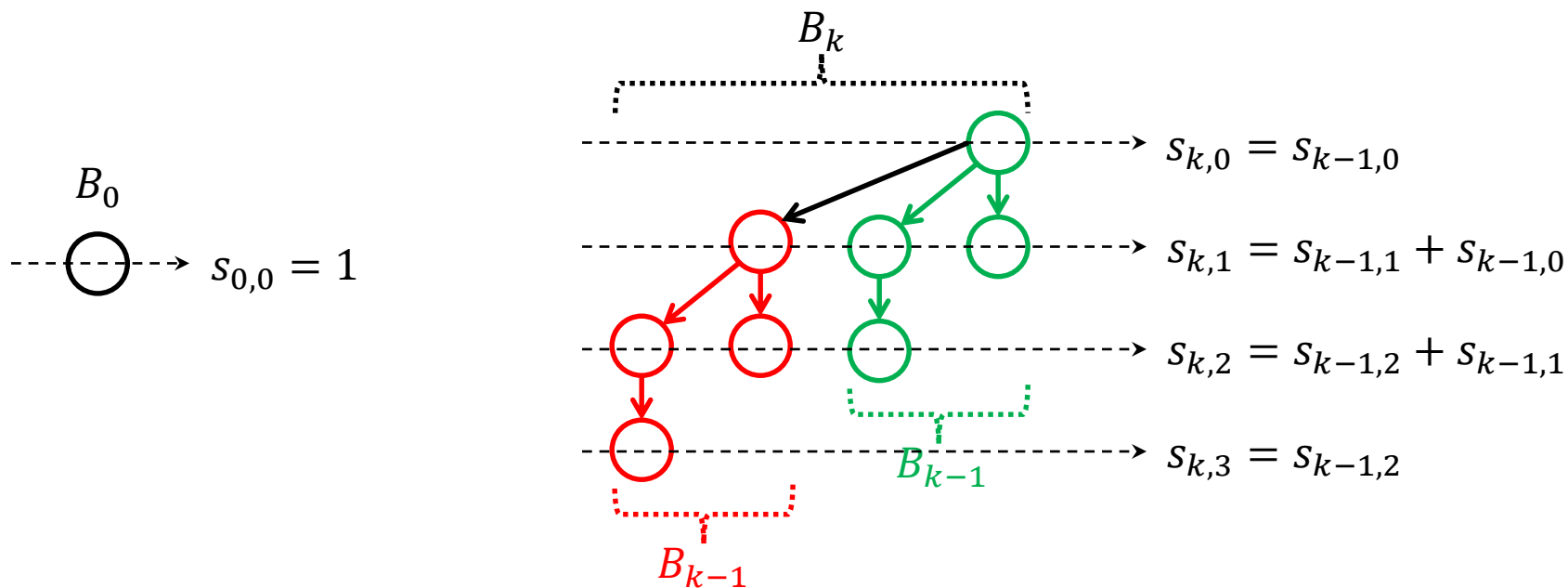


# Binomial Trees

**Prove:**  $B_k$  has exactly  $\binom{k}{i}$  nodes at depth  $i = 0, 1, 2, \dots, k$ .

**Proof:** Suppose  $B_k$  has  $s_{k,i}$  nodes at depth  $i$ .

$$s_{k,i} = \begin{cases} 0 & \text{if } i < 0 \text{ or } i > k, \\ 1 & \text{if } i = k = 0, \\ s_{k-1,i} + s_{k-1,i-1} & \text{otherwise.} \end{cases}$$



# Binomial Trees

$$s_{k,i} = \begin{cases} 0 & \text{if } i < 0 \text{ or } i > k, \\ 1 & \text{if } i = k = 0, \\ s_{k-1,i} + s_{k-1,i-1} & \text{otherwise.} \end{cases}$$

$$\Rightarrow s_{k,i} = [k \geq i \geq 0](s_{k-1,i} + s_{k-1,i-1} + [i = k = 0])$$

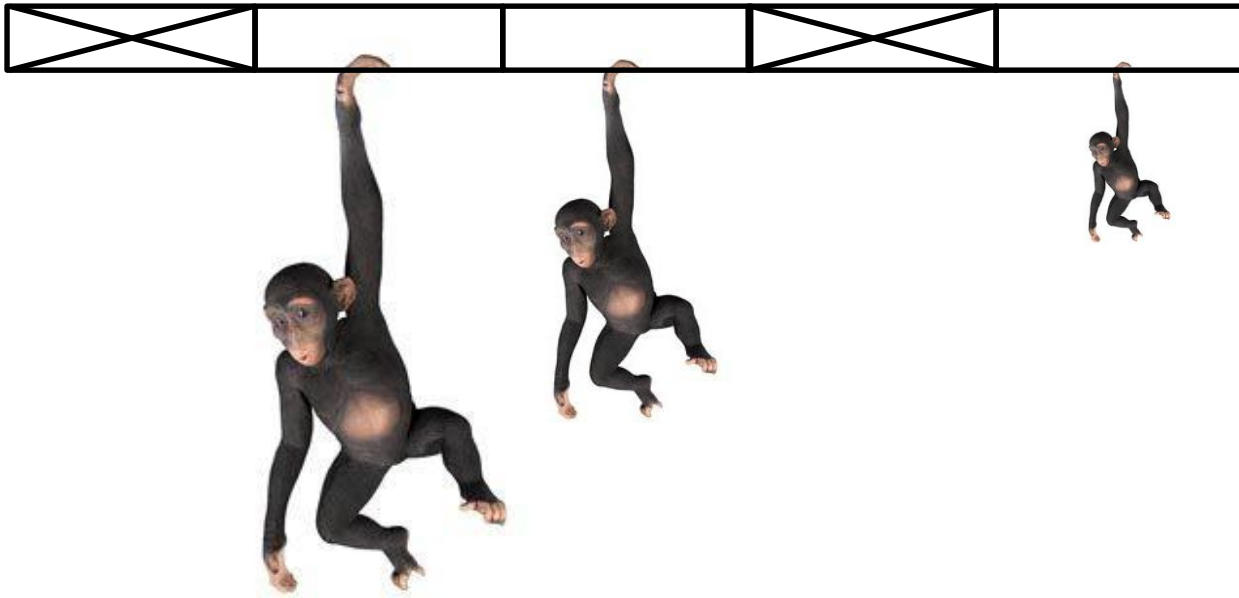
Generating function:  $S_k(z) = s_{k,0} + s_{k,1}z + s_{k,2}z^2 + \dots + s_{k,k}z^k$

$$\begin{aligned} S_{k \geq 0}(z) &= \sum_{i=0}^k s_{k,i}z^i = \sum_{i=0}^k s_{k-1,i}z^i + \sum_{i=0}^k s_{k-1,i-1}z^i + [k=0] \sum_{i=0}^k [i=0]z^i \\ &= \sum_{i=0}^{k-1} s_{k-1,i}z^i + z \sum_{i=0}^{k-1} s_{k-1,i}z^i + [k=0] \\ &= S_{k-1}(z) + zS_{k-1}(z) + [k=0] = (1+z)S_{k-1}(z) + [k=0] \\ \Rightarrow S_k(z) &= \begin{cases} 1 & \text{if } k = 0, \\ (1+z)S_{k-1}(z) & \text{otherwise.} \end{cases} \\ &= (1+z)^k \end{aligned}$$

Equating the coefficient of  $z^i$  from both sides:  $s_{k,i} = \binom{k}{i}$

# Binomial Heaps

A *binomial heap*  $H$  is a set of binomial trees that satisfies the following properties:

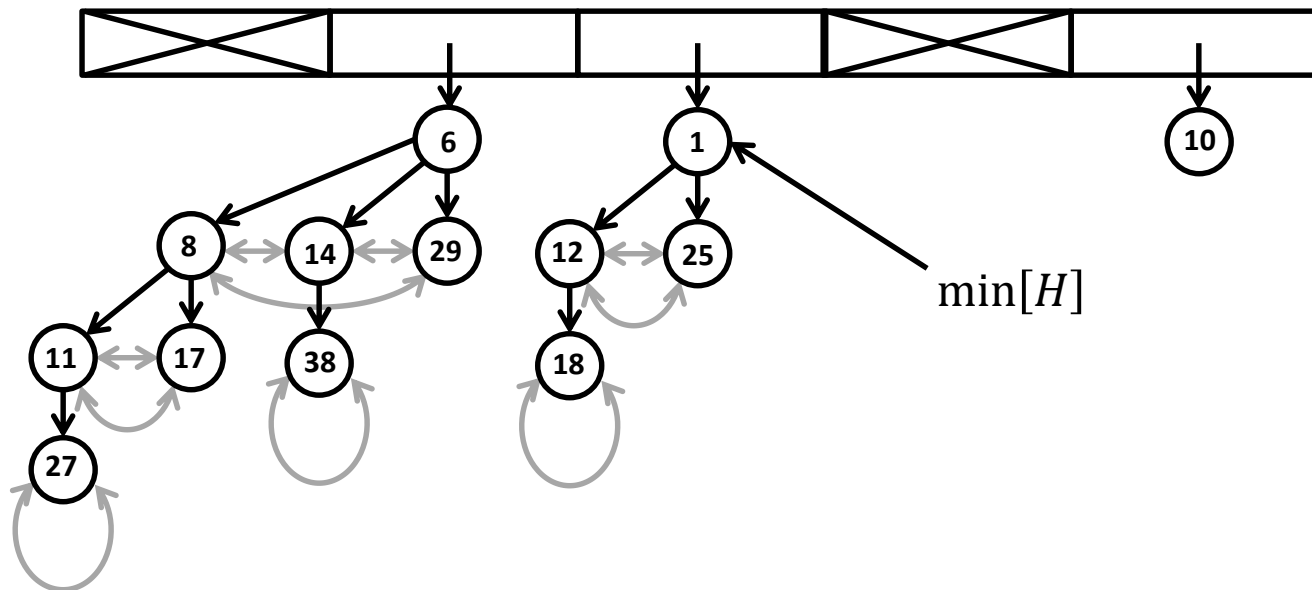




# Binomial Heaps

A *binomial heap*  $H$  is a set of binomial trees that satisfies the following properties:

1. each node has a key
2. each binomial tree in  $H$  obeys the min-heap property
3. for any integer  $k \geq 0$ , there is at most one binomial tree in  $H$  whose root node has degree  $k$



# Rank of Binomial Trees

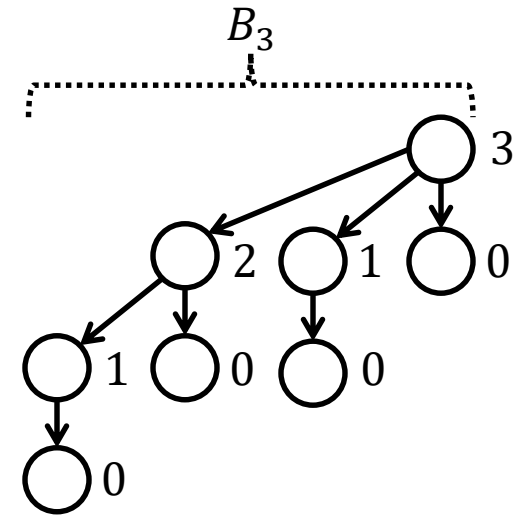
The *rank* of a binomial tree node  $x$ , denoted  $rank(x)$ , is the number of children of  $x$ .

The figure on the right shows the rank of each node in  $B_3$ .

Observe that  $rank(\text{root}(B_k)) = k$ .

Rank of a binomial tree is the rank of its root. Hence,

$$rank(B_k) = rank(\text{root}(B_k)) = k$$

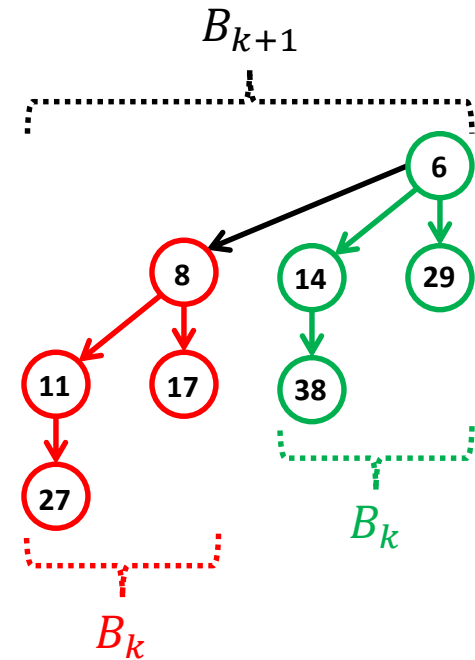


# A Basic Operation: Linking Two Binomial Trees

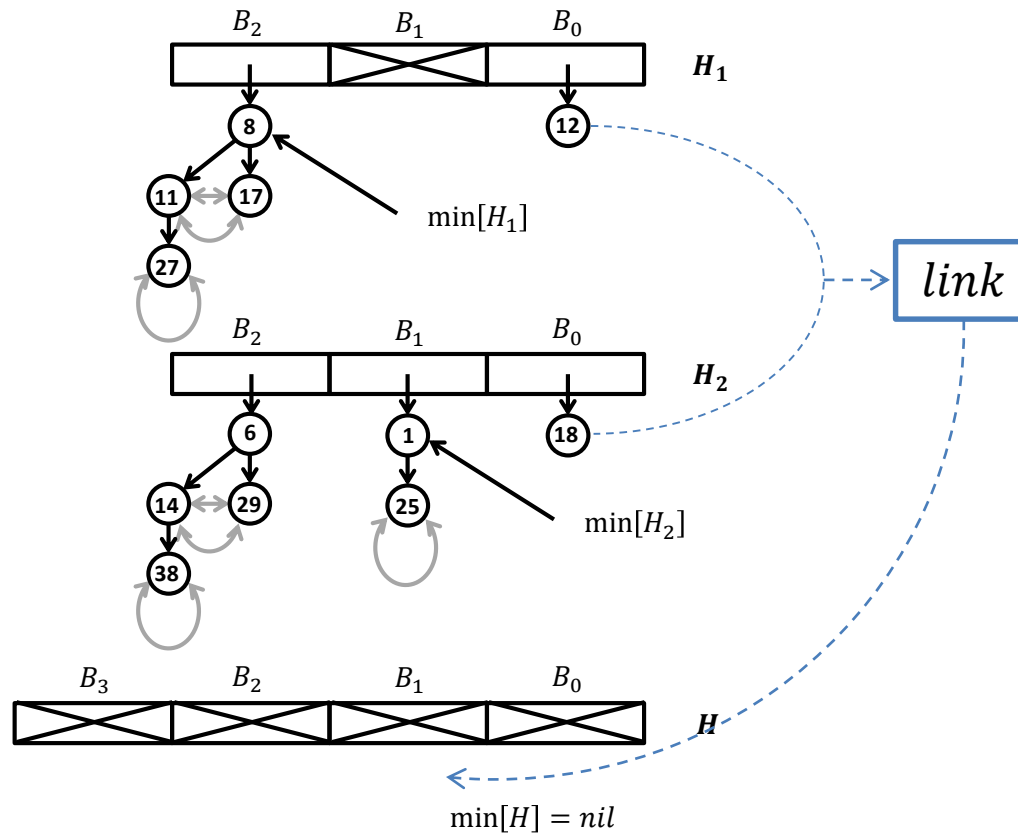
Given *two binomial trees of the same rank*, say, two  $B_k$ 's, we link them in constant time by making the root of one tree the left child of the root of the other, and thus producing a  $B_{k+1}$ .

If the trees are part of a binomial min-heap, we always make the root with the smaller key the parent, and the one with the larger key the child.

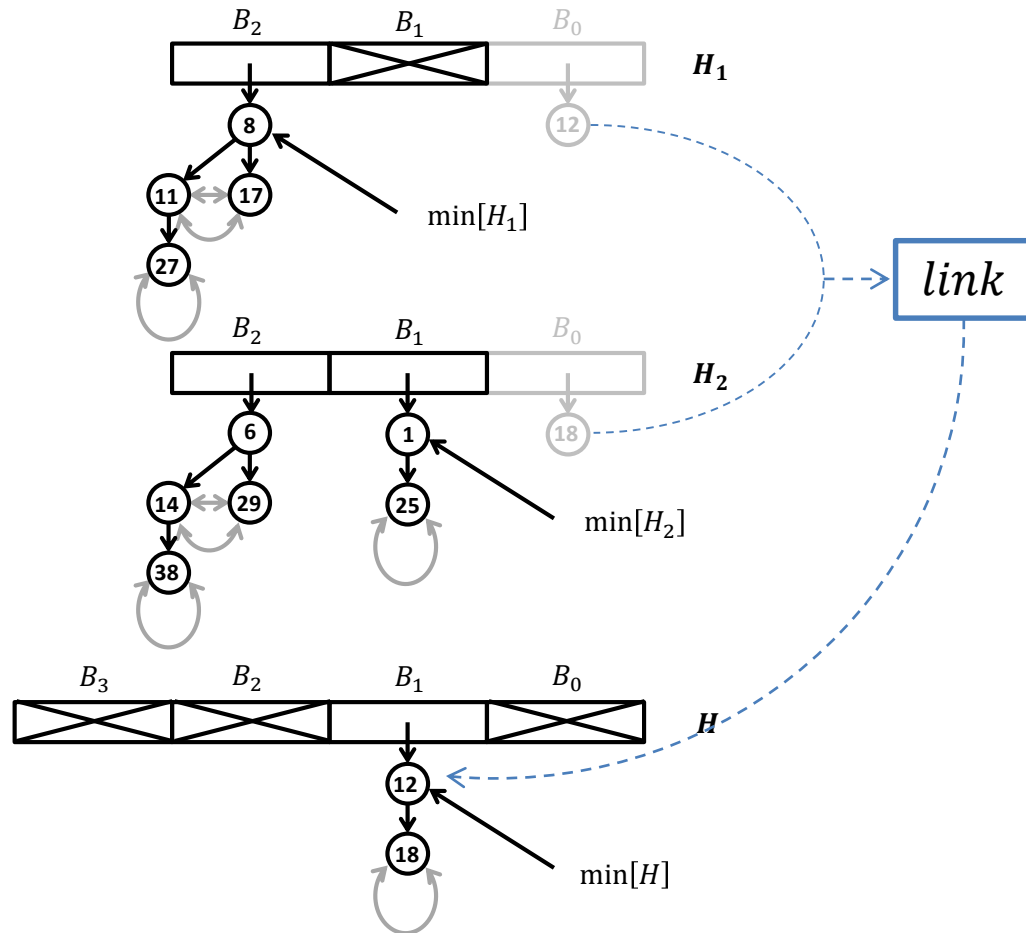
Ties are broken arbitrarily.



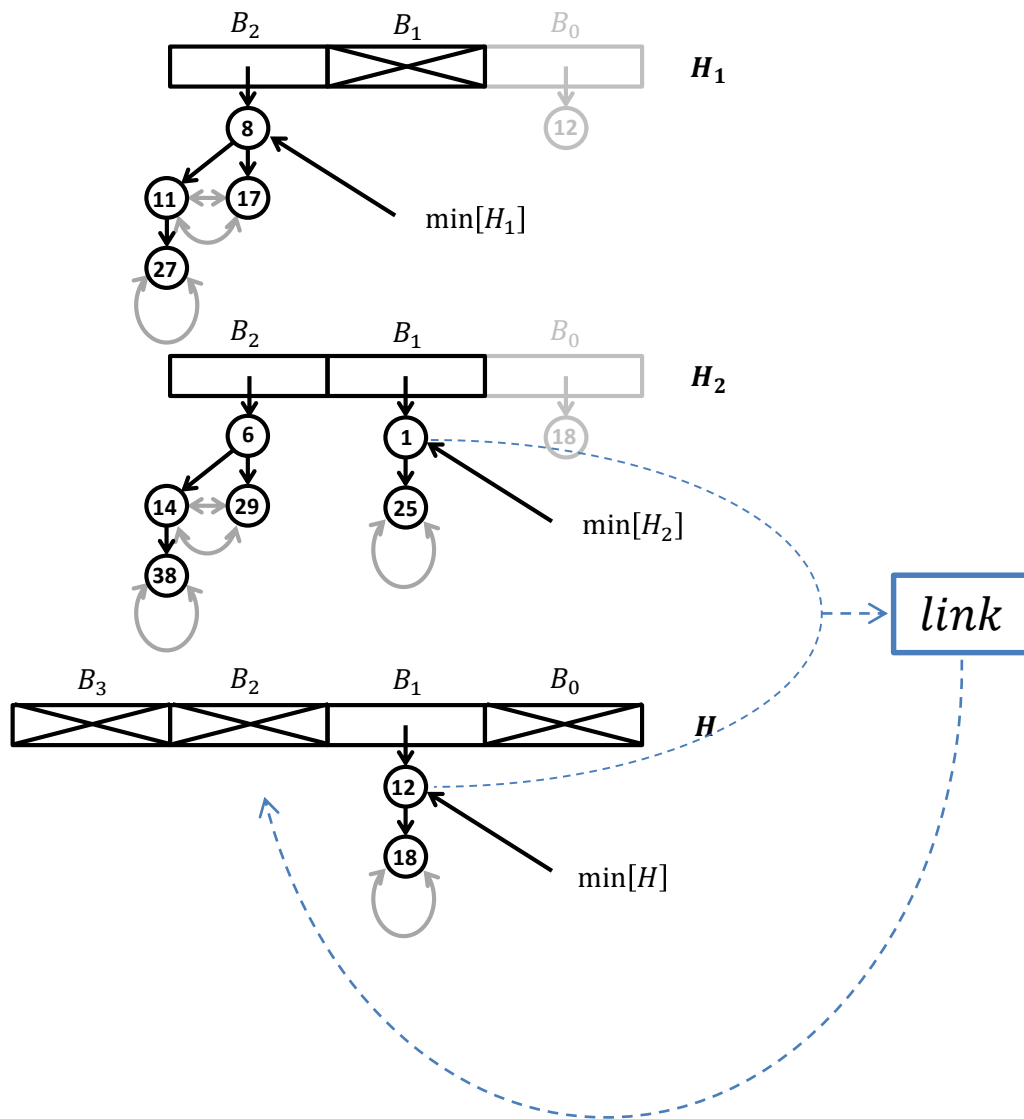
# Binomial Heap Operations: UNION( $H_1, H_2$ )



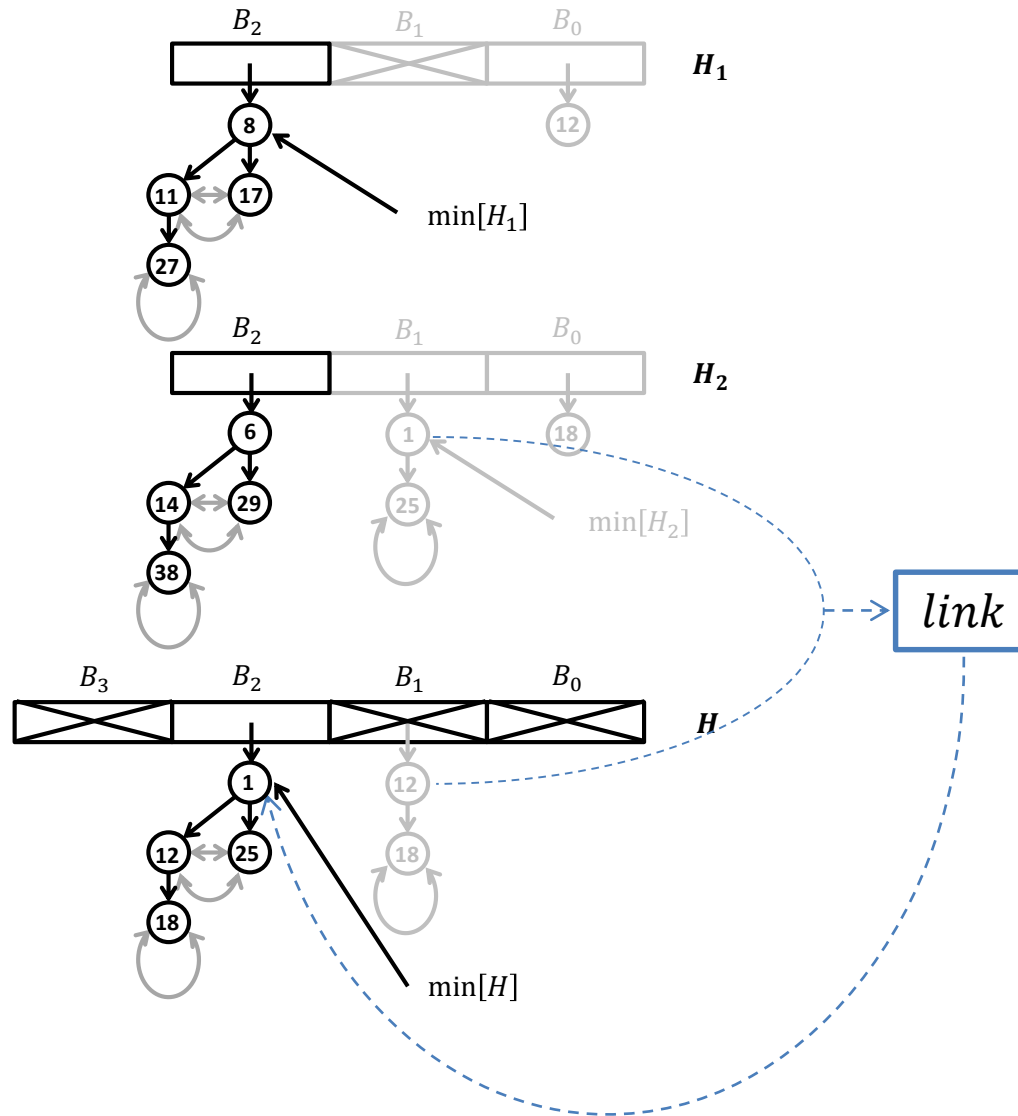
# Binomial Heap Operations: UNION( $H_1, H_2$ )



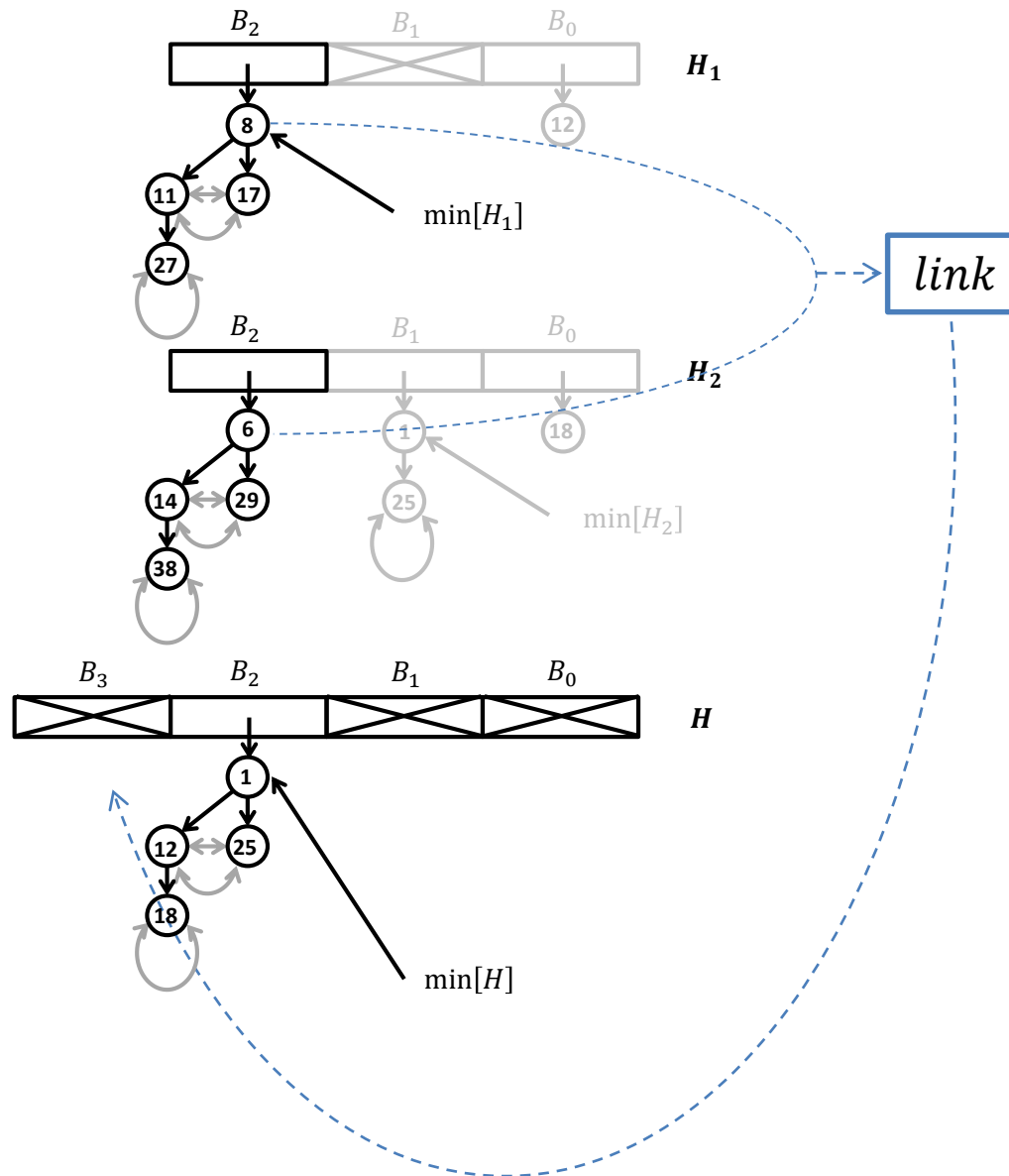
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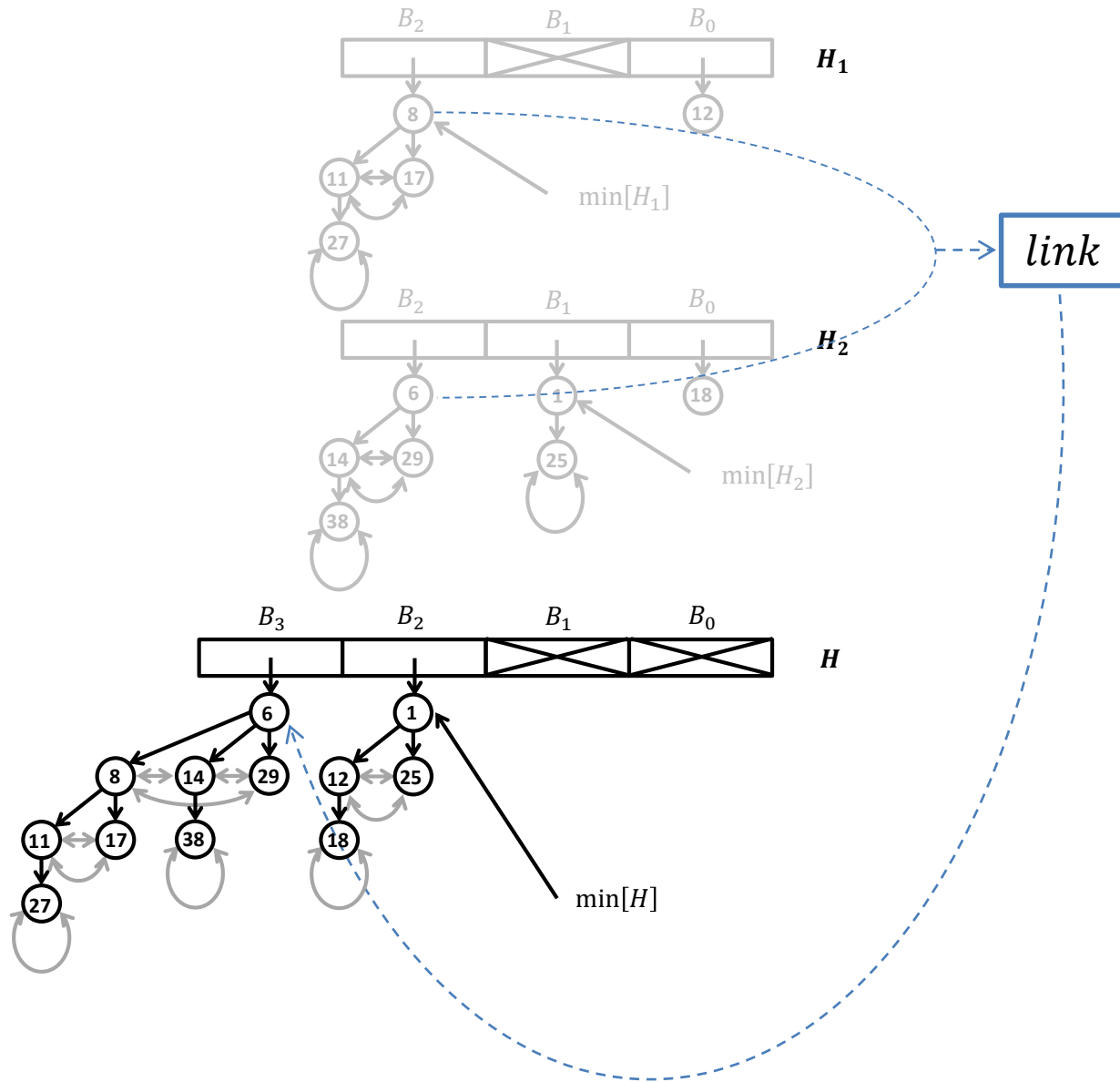


# Binomial Heap Operations: UNION( $H_1, H_2$ )

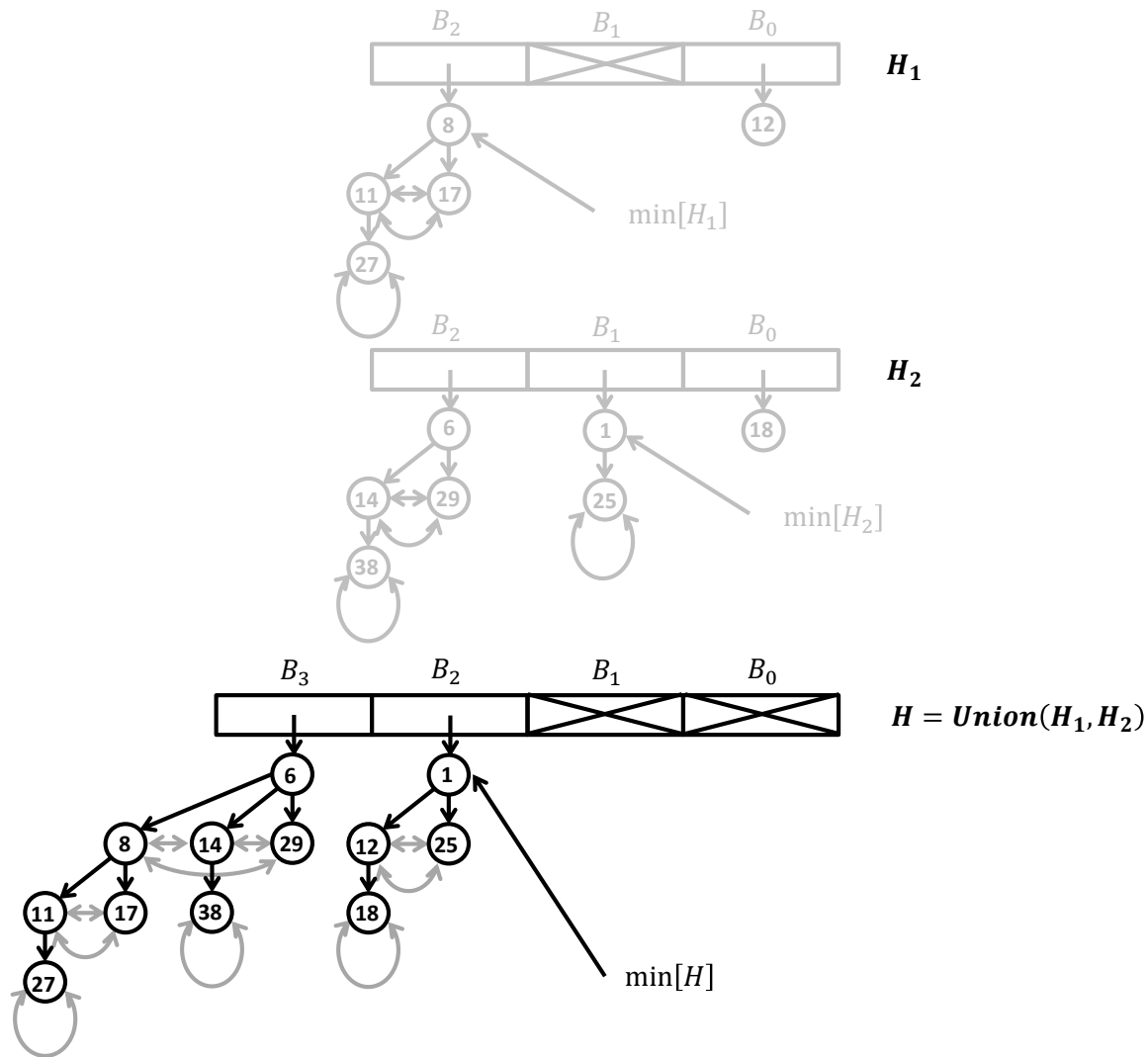




# Binomial Heap Operations: UNION( $H_1, H_2$ )



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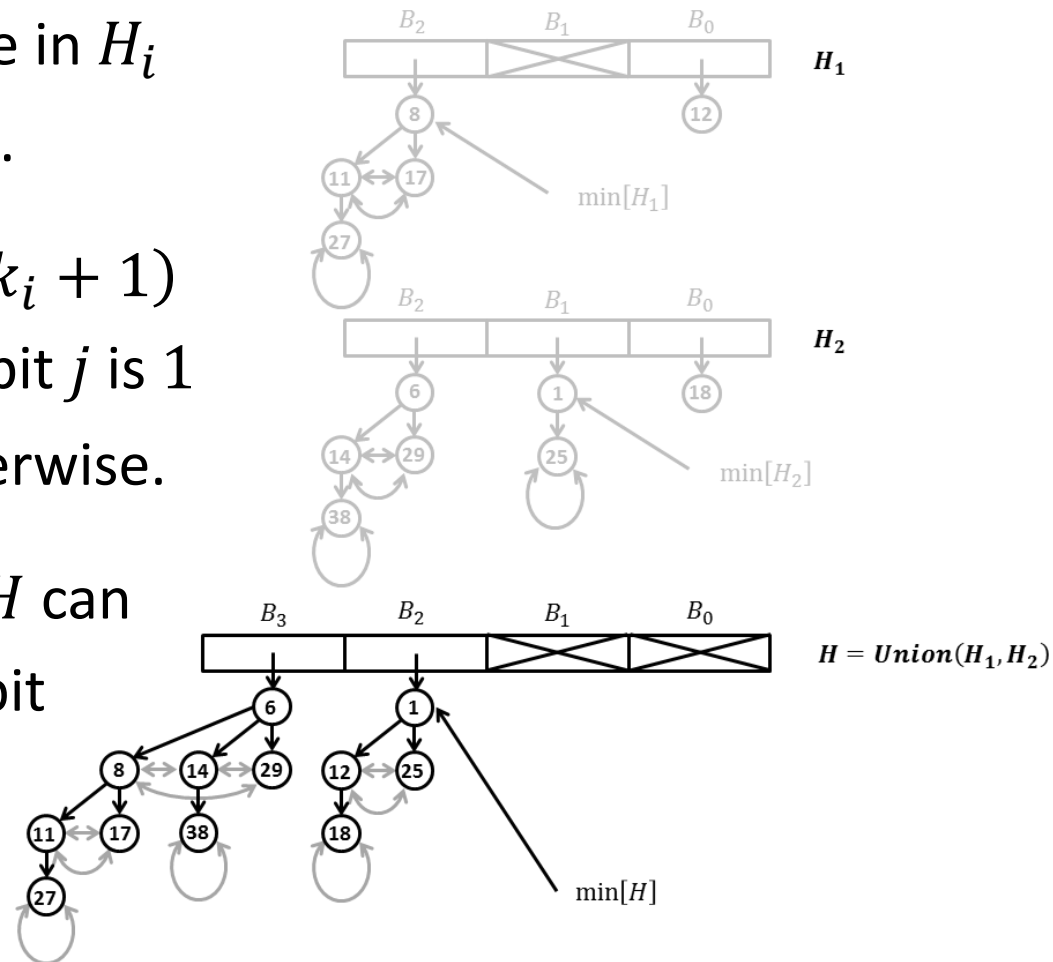
UNION( $H_1, H_2$ ) works in exactly the same way as binary addition.

Let  $n_i$  be the number of nodes in  $H_i$  ( $i = 1, 2$ ).

Then the largest binomial tree in  $H_i$  is a  $B_{k_i}$ , where  $k_i = \lfloor \log_2 n_i \rfloor$ .

Thus  $H_i$  can be treated as a  $(k_i + 1)$  bit binary number  $x_i$ , where bit  $j$  is 1 if  $H_i$  contains a  $B_j$ , and 0 otherwise.

If  $H = \text{Union}(H_1, H_2)$ , then  $H$  can be viewed as a  $k = \lfloor \log_2 n \rfloor$  bit binary number  $x = x_1 + x_2$ , where  $n = n_1 + n_2$ .



# Binomial Heap Operations: UNION( $H_1, H_2$ )

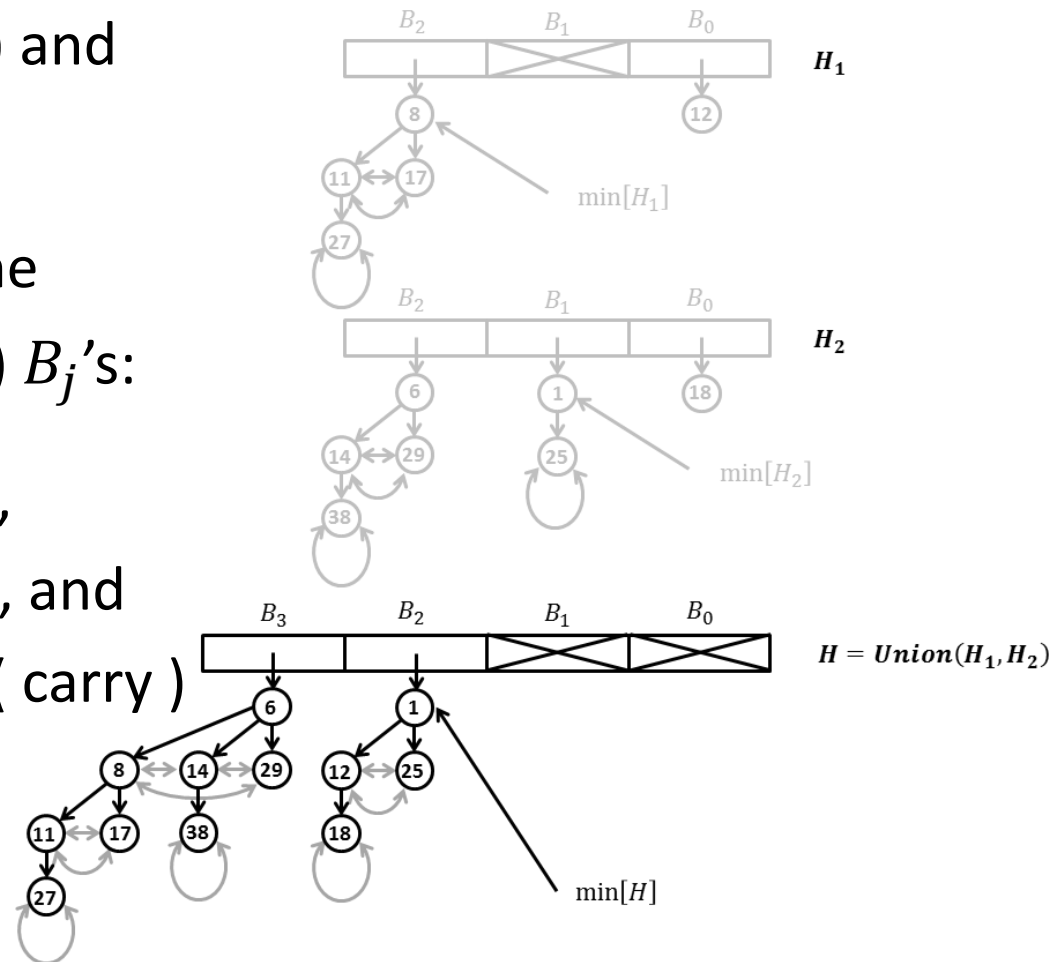
UNION( $H_1, H_2$ ) works in exactly the same way as binary addition.

Initially,  $H$  does not contain any binomial trees.

Melding starts from  $B_0$  ( LSB ) and continues up to  $B_k$  ( MSB ).

At each location  $j \in [0, k]$ , one encounters at most three ( 3 )  $B_j$ 's:

- at most 1 from  $H_1$  ( input ),
- at most 1 from  $H_2$  ( input ), and
- if  $j > 0$ , at most 1 from  $H$  ( carry )

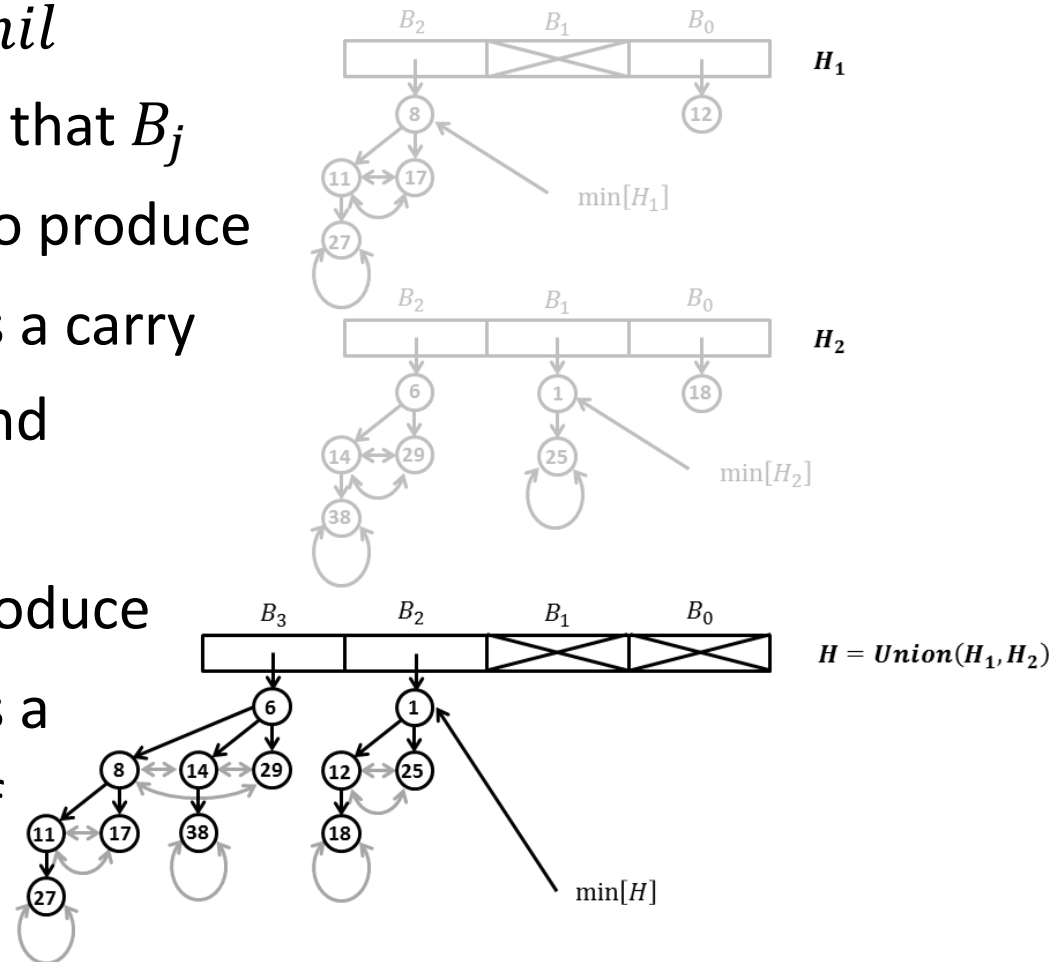


# Binomial Heap Operations: UNION( $H_1, H_2$ )

UNION( $H_1, H_2$ ) works in exactly the same way as binary addition.

When the number of  $B_j$ 's at location  $j \in [0, k]$  is:

- 0: location  $j$  of  $H$  is set to *nil*
- 1: location  $j$  of  $H$  points to that  $B_j$
- 2: the two  $B_j$ 's are linked to produce a  $B_{j+1}$  which is stored as a carry at location  $j + 1$  of  $H$ , and location  $j$  is set to *nil*
- 3: two  $B_j$ 's are linked to produce a  $B_{j+1}$  which is stored as a carry at location  $j + 1$  of  $H$ , and the 3<sup>rd</sup>  $B_j$  is stored at location  $j$



# Binomial Heap Operations: UNION( $H_1, H_2$ )

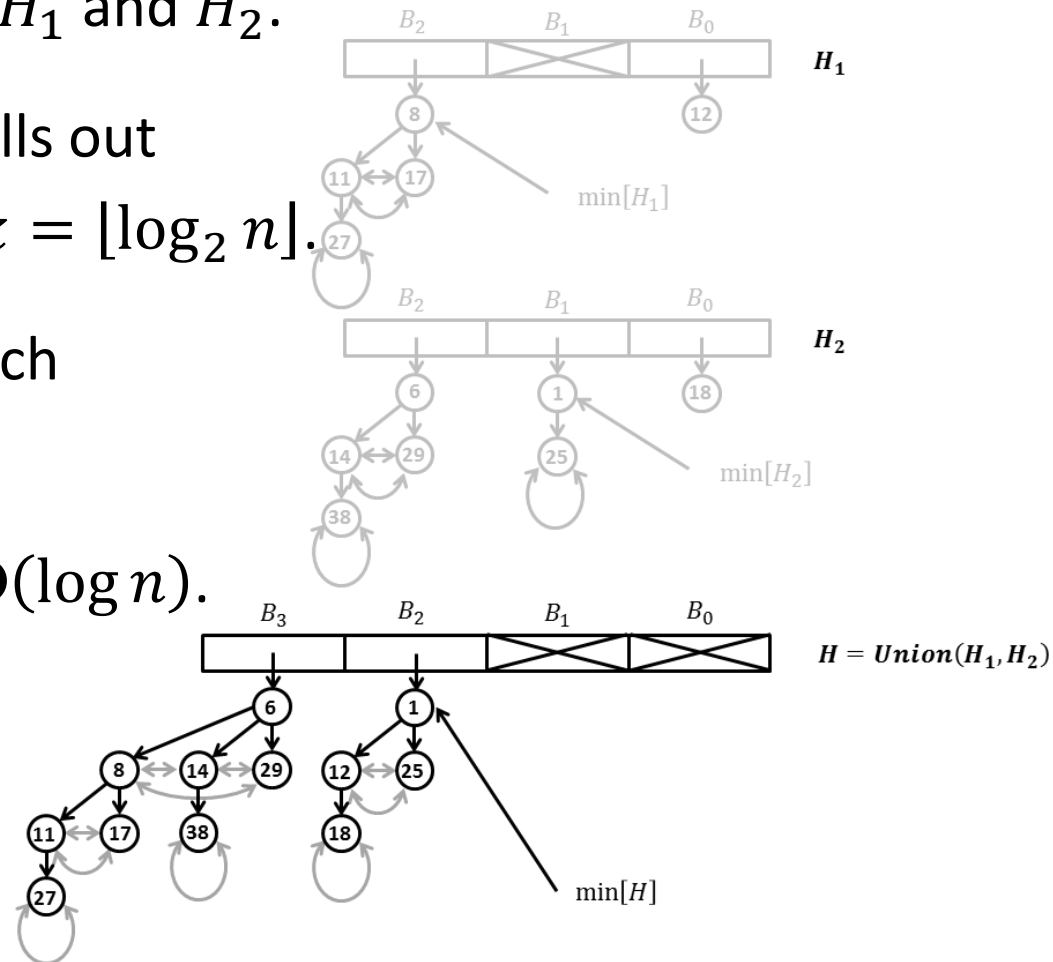
UNION( $H_1, H_2$ ) works in exactly the same way as binary addition.

Worst case cost of UNION( $H_1, H_2$ ) is clearly  $\Theta(\log n)$ , where  $n$  is the total number of nodes in  $H_1$  and  $H_2$ .

Observe that this operation fills out  $k + 1$  locations of  $H$ , where  $k = \lfloor \log_2 n \rfloor$ .

It does only  $\Theta(1)$  work for each location.

Hence, total cost is  $\Theta(k) = \Theta(\log n)$ .



# Binomial Heap Operations: UNION( $H_1, H_2$ )

One can improve the performance of UNION( $H_1, H_2$ ) as follows.

W.l.o.g., suppose  $H_2$  is at least as large as  $H_1$ , i.e.,  $n_2 \geq n_1$ .

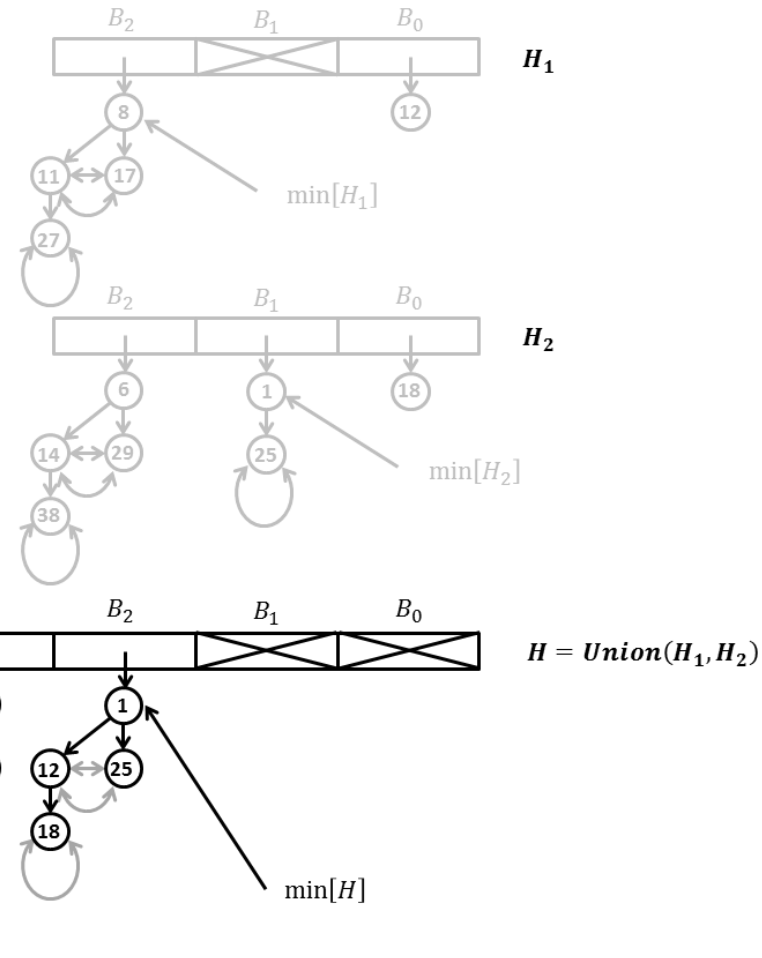
We also assume that  $H_2$  has enough space to store at least up to  $B_k$ , where,  $k = \lfloor \log_2(n_1 + n_2) \rfloor$ .

Then instead of melding  $H_1$  and  $H_2$  to a new heap  $H$ , we can meld them in-place at  $H_2$ .

After melding till  $B_{k_1}$ , we stop once the carry stops propagating.

The cost is  $\Omega(k_1)$ , but  $O(k_2)$ .

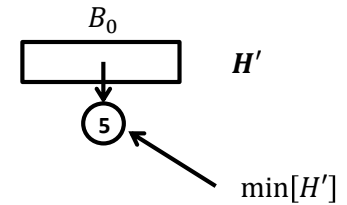
Worst-case cost is still  $O(k) = O(\log n)$ .



# Binomial Heap Operations: INSERT( $H, x$ )

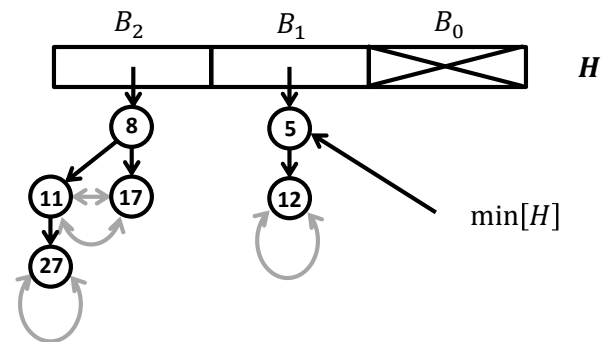
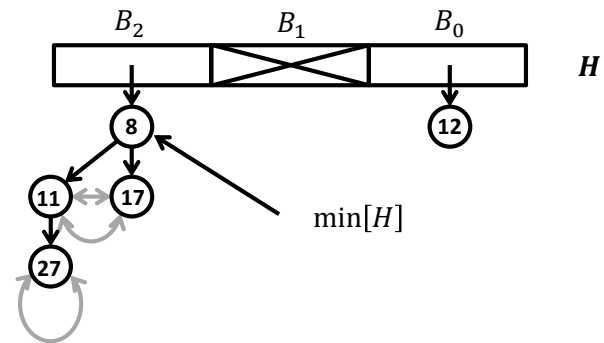
**Step 1:**  $H' \leftarrow \text{MAKE-HEAP}( x )$

Takes  $\Theta(1)$  time.



**Step 2:**  $H \leftarrow \text{UNION}( H, H' )$   
( in-place at  $H$  )

Takes  $O(\log n)$  time, where  
 $n$  is the number of nodes in  $H$ .

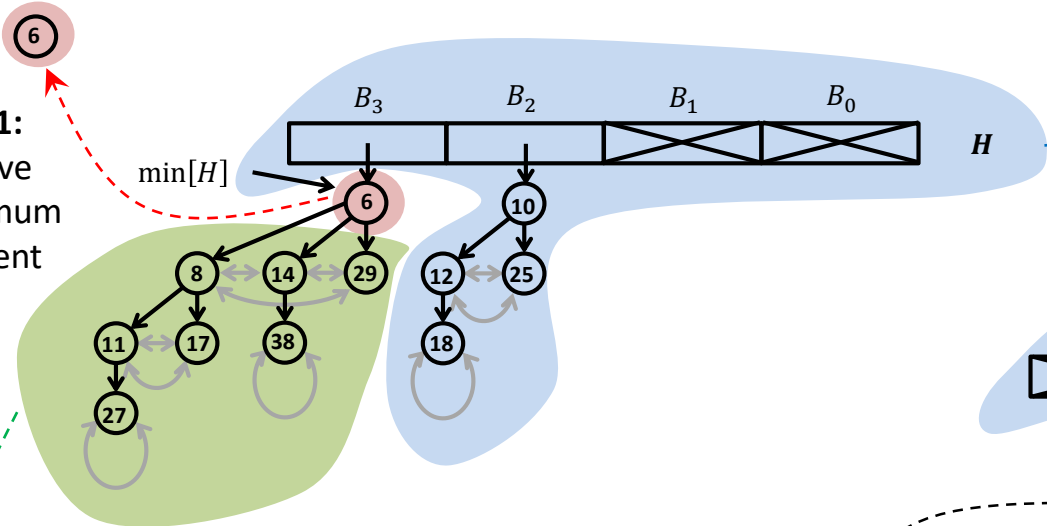


Thus the worst-case cost of  
 $\text{INSERT}( H, x )$  is  $O(\log n)$ , where  
 $n$  is the number of items already  
in the heap.

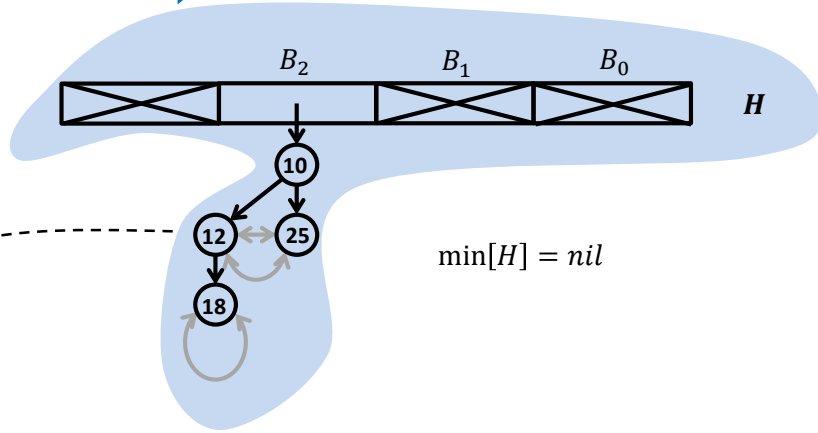


# Binomial Heap Operations: EXTRACT-MIN( $H$ )

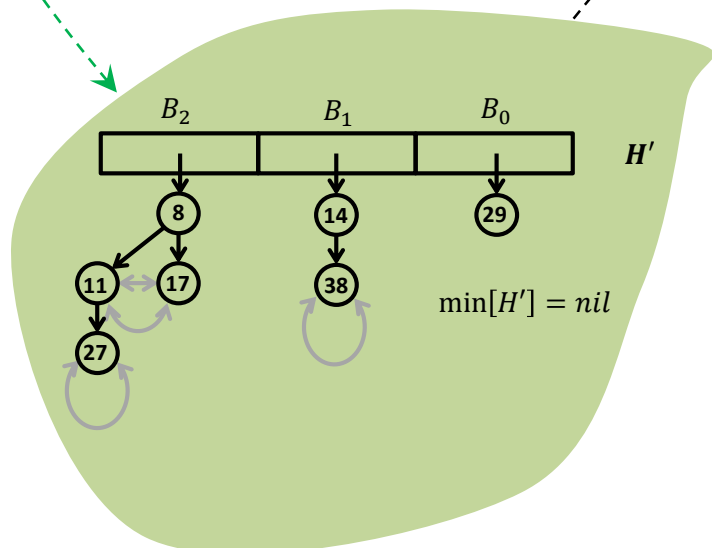
**Step 1:**  
remove  
minimum  
element



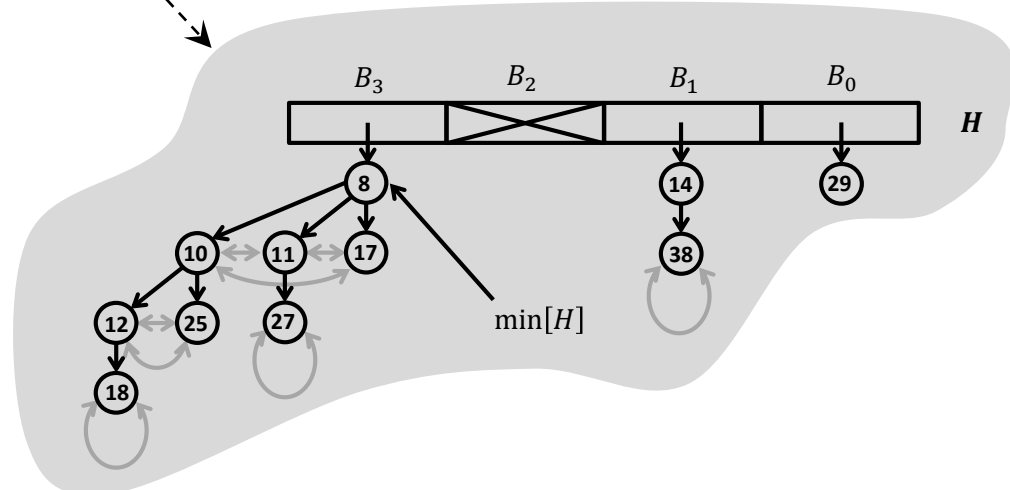
**Step 2:** remove the binomial tree with the smallest root from the input heap



**Step 3:** remove the root of the binomial tree with the minimum element, and form a new binomial heap from the children of the removed root

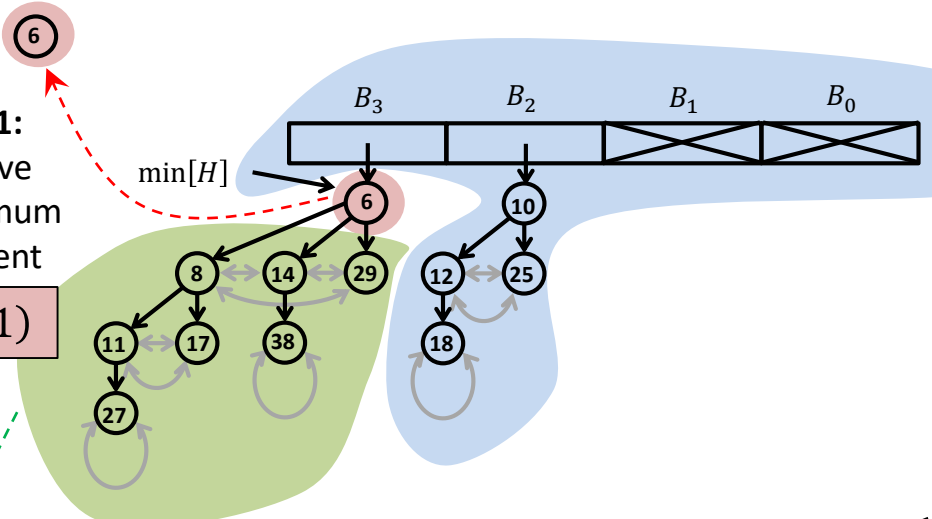


**Step 4:** UNION( $H, H'$ ) and update the min pointer

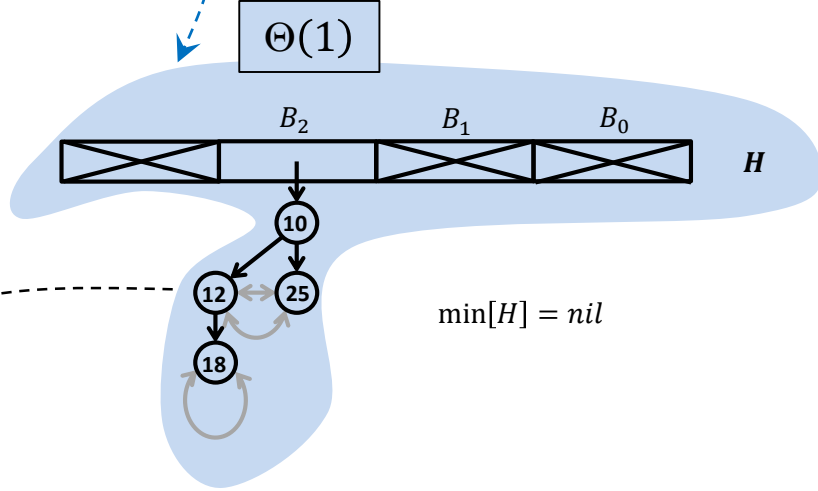


# Binomial Heap Operations: EXTRACT-MIN( H )

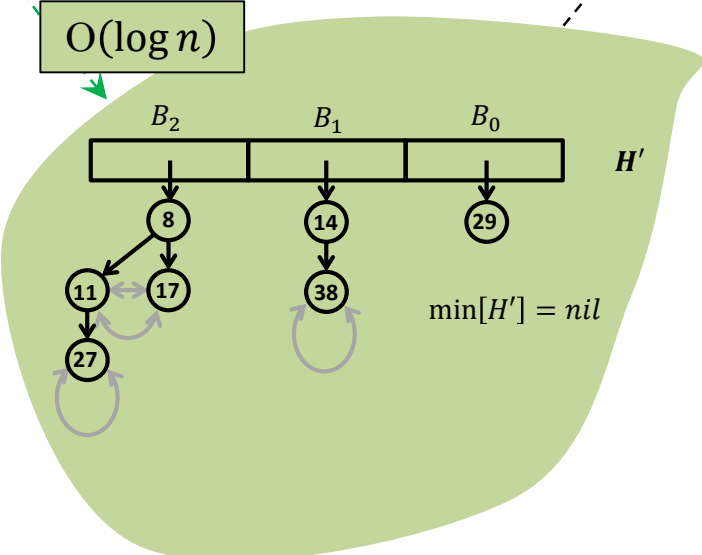
**Step 1:** remove minimum element  
 $\Theta(1)$



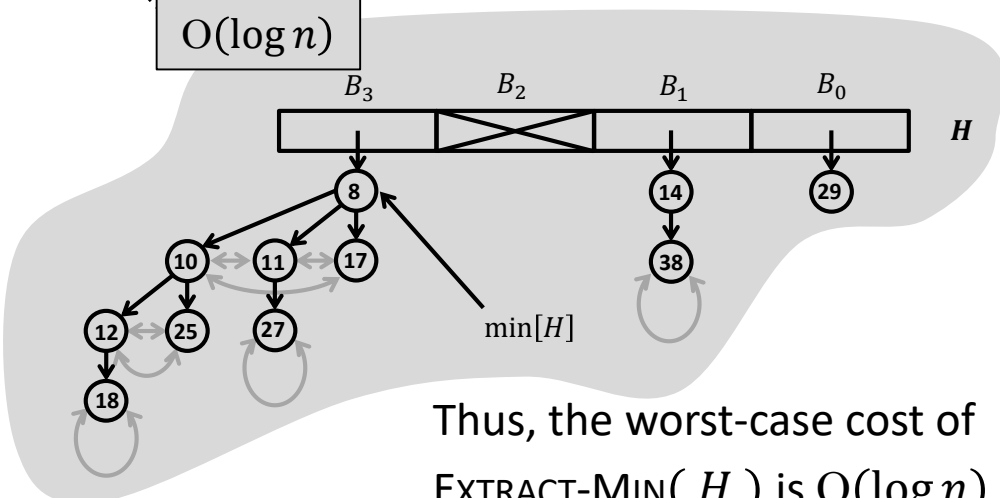
**Step 2:** remove the binomial tree with the smallest root from the input heap  
 $\Theta(1)$



**Step 3:** remove the root of the binomial Tree with the minimum element, and form a new binomial heap from the children of the removed root  
 $O(\log n)$



**Step 4:** UNION( H, H' ) and update the min pointer  
 $O(\log n)$



Thus, the worst-case cost of EXTRACT-MIN( H ) is  $O(\log n)$

# Binomial Heap Operations

Heap Operation	Worst-case
MAKE-HEAP	$\Theta(1)$
INSERT	$O(\log n)$
MINIMUM	$\Theta(1)$
EXTRACT-MIN	$O(\log n)$
UNION	$O(\log n)$

# Amortized Analysis ( Accounting Method )

We maintain a credit account for every tree in the heap, and always maintain the following invariant:

$$\bigwedge_{B_j \in H} \text{credit}(B_j) = 1$$

**MAKE-HEAP(  $x$  ):**

actual cost,  $c_i = 1$  ( for creating the singleton heap )

extra charge,  $\delta_i = 1$  ( for storing in the credit account  
of the new tree )

amortized cost,  $\hat{c}_i = c_i + \delta_i = 2 = \Theta(1)$

# Amortized Analysis ( Accounting Method )

We maintain a credit account for every tree in the heap, and always maintain the following invariant:

$$\bigwedge_{B_j \in H} \text{credit}(B_j) = 1$$

**LINK(  $B_k^{(1)}$ ,  $B_k^{(2)}$  ):**

actual cost,  $c_i = 1$  ( for linking the two trees )

We use  $\text{credit}(B_k^{(1)})$  pay for this actual work.

Let  $B_{k+1}$  be the newly created tree. We restore the credit invariant by transferring  $\text{credit}(B_k^{(2)})$  to  $\text{credit}(B_{k+1})$ .

Hence, amortized cost,  $\hat{c}_i = c_i + \delta_i = 1 - 1 = 0$

# Amortized Analysis ( Accounting Method )

We maintain a credit account for every tree in the heap, and always maintain the following invariant:

$$\bigwedge_{B_j \in H} \text{credit}(B_j) = 1$$

**INSERT(  $H, x$  ):**

Amortized cost of MAKE-HEAP(  $x$  ) is = 2

Then UNION(  $H, H'$  ) is simply a sequence of free LINK operations with only a constant amount of additional work that do not create any new trees. Thus the credit invariant is maintained, and the amortized cost of this step is = 1.

Hence, amortized cost of INSERT,  $\hat{c}_i = 2 + 1 = 3 = \Theta(1)$

# Amortized Analysis ( Accounting Method )

We maintain a credit account for every tree in the heap, and always maintain the following invariant:

$$\bigwedge_{B_j \in H} \text{credit}(B_j) = 1$$

**UNION(  $H_1, H_2$  ):**

UNION(  $H_1, H_2$  ) includes a sequence of free LINK operations that maintain the credit invariant.

But it also includes  $O(\log n)$  other operations that are not free ( e.g., consider melding a heap with  $n = 2^k$  elements with one containing  $n - 1$  elements ). These operations do not create new trees (and so do not violate the credit invariant), and each cost  $\Theta(1)$ .

Hence, amortized cost of UNION,  $\hat{c}_i = O(\log n)$

# Amortized Analysis ( Accounting Method )

We maintain a credit account for every tree in the heap, and always maintain the following invariant:

$$\bigwedge_{B_j \in H} \text{credit}(B_j) = 1$$

**EXTRACT-MIN(  $H$  ):**

Steps 1 & 2: The  $\Theta(1)$  actual cost is paid for by the credit released by the deleted tree.

Step 3: Exposes  $O(\log n)$  new trees, and we charge 1 unit of extra credit for storing in the credit account of each such tree.

Step 4: Performs a UNION that has  $O(\log n)$  amortized cost.

Hence, amortized cost of EXTRACT-MIN,  $\hat{c}_i = O(\log n)$



# Amortized Analysis ( Potential Method )

Potential Function,

$$\Phi(D_i) = c \times ( \text{\#trees in the data structure after the } i\text{-th operation} ),$$

where  $c$  is a constant.

Clearly,  $\Phi(D_0) = 0$  ( no trees in the data structure initially )

and for all  $i > 0$ ,  $\Phi(D_i) \geq 0$  ( #trees cannot be negative )

**MAKE-HEAP(  $x$  ):**

actual cost,  $c_i = 1$  ( for creating the singleton heap )

potential change,  $\Delta_i = \Phi(D_i) - \Phi(D_{i-1}) = c$

( as #trees increases by 1 )

amortized cost,  $\hat{c}_i = c_i + \Delta_i = 1 + c = \Theta(1)$

# Amortized Analysis ( Potential Method )

Potential Function,

$$\Phi(D_i) = c \times ( \text{\#trees in the data structure after the } i\text{-th operation} ),$$

where  $c$  is a constant.

**INSERT(  $H, x$  ):**

The number of trees increases by 1 initially.

Then the operation scans  $k > 0$  ( say ) locations of the array of tree pointers. Observe that we use tree linking  $(k - 1)$  times each of which reduces the number of trees by 1.

$$\text{actual cost, } c_i = 1 + k$$

$$\begin{aligned} \text{potential change, } \Delta_i &= \Phi(D_i) - \Phi(D_{i-1}) = c(1 - (k - 1)) \\ &= c - c(k - 1) \end{aligned}$$

$$\text{amortized cost, } \hat{c}_i = c_i + \Delta_i = 2 + c - (c - 1)(k - 1)$$

$$\text{For } c \geq 1, \text{ we have, } \hat{c}_i \leq 2 + c = \Theta(1)$$

# Amortized Analysis ( Potential Method )

Potential Function,

$$\Phi(D_i) = c \times ( \text{\#trees in the data structure after the } i\text{-th operation} ),$$

where  $c$  is a constant.

**UNION(  $H_1, H_2$  ):**

Suppose the operation scans  $k > 0$  locations of the array of tree pointers, and uses the link operation  $l$  times. Observe that  $k > l \geq 0$ . Each link reduces the number of trees by 1.

actual cost,  $c_i = k$

potential change,  $\Delta_i = \Phi(D_i) - \Phi(D_{i-1}) = -c \times l$

amortized cost,  $\hat{c}_i = c_i + \Delta_i = k - c \times l$

Since  $k = O(\log n)$  and  $l = O(\log n)$ , we have,

$$\hat{c}_i = O(\log n) \text{ for any } c.$$

# Amortized Analysis ( Potential Method )

Potential Function,

$$\Phi(D_i) = c \times ( \text{\#trees in the data structure after the } i\text{-th operation} ),$$

where  $c$  is a constant.

**EXTRACT-MIN(  $H$  ):**

Let in Step 1:  $r$  = rank of the tree with the smallest key

and in Step 4:  $k$  = #locations of pointer array scanned during UNION

$l$  = #link operations during UNION

$t$  = #trees in the heap after the UNION

Then actual cost,  $c_i = 1$  ( step 1 ) + 1 ( step 2 ) +  $r$  ( step 3 )

+  $k$  ( step 4: union ) +  $t$  ( step 4: update *min* ptr )

$$= 2 + k + t + r$$

# Amortized Analysis ( Potential Method )

Potential Function,

$$\Phi(D_i) = c \times ( \text{\#trees in the data structure after the } i\text{-th operation} ),$$

where  $c$  is a constant.

**EXTRACT-MIN(  $H$  ):**

Let in Step 1:  $r =$  rank of the tree with the smallest key

and in Step 4:  $k =$  #locations of pointer array scanned during UNION

$l =$  #link operations during UNION

$t =$  #trees in the heap after the UNION

potential change,  $\Delta_i = \Phi(D_i) - \Phi(D_{i-1})$

$= c \times (r - 1)$  ( removing *min* element in step 1  
removes 1 tree but creates  $r$  new ones )

$-c \times l$  ( linkings in step 4  
reduces #trees by  $l$  )

# Amortized Analysis ( Potential Method )

Potential Function,

$$\Phi(D_i) = c \times ( \text{\#trees in the data structure after the } i\text{-th operation} ),$$

where  $c$  is a constant.

**EXTRACT-MIN(  $H$  ):**

Let in Step 1:  $r$  = rank of the tree with the smallest key

and in Step 4:  $k$  = #locations of pointer array scanned during UNION

$l$  = #link operations during UNION

$t$  = #trees in the heap after the UNION

actual cost,  $c_i = 2 + k + t + r$

potential change,  $\Delta_i = \Phi(D_i) - \Phi(D_{i-1}) = c \times (r - l - 1)$

Then amortized cost,  $\hat{c}_i = c_i + \Delta_i = 2 + k + t + r + c \times (r - l - 1)$

Since  $k = O(\log n)$ ,  $l = O(\log n)$ ,  $t = O(\log n)$  &  $r = O(\log n)$ ,

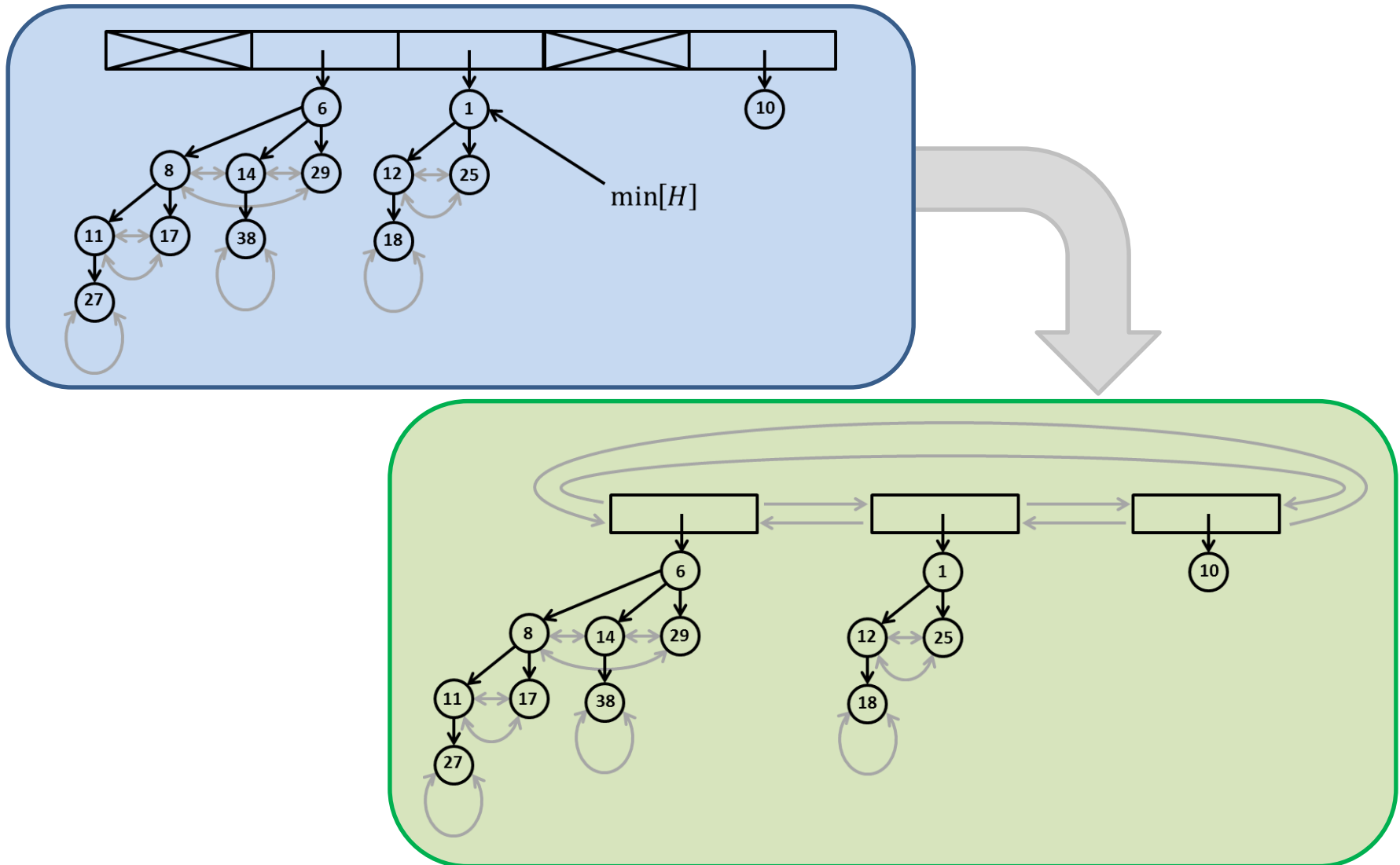
we have,  $\hat{c}_i = O(\log n)$  for any  $c$ .

# Binomial Heap Operations

Heap Operation	Worst-case	Amortized
MAKE-HEAP	$\Theta(1)$	$\Theta(1)$
INSERT	$O(\log n)$	$\Theta(1)$
MINIMUM	$\Theta(1)$	$\Theta(1)$
EXTRACT-MIN	$O(\log n)$	$O(\log n)$
UNION	$O(\log n)$	$O(\log n)$

# Binomial Heaps with Lazy Union

We maintain pointers to the trees in a doubly linked circular list ( instead of an array ), but do not maintain a *min* pointer.





# Binomial Heap Operations with Lazy Union

We maintain the following invariant:

$$\bigwedge_{B_j \in H} \text{credit}(B_j) = 2$$

**MAKE-HEAP(  $x$  )**: Create a singleton heap as before. Hence, amortized cost =  $\Theta(1)$ .

**LINK(  $B_k^{(1)}, B_k^{(2)}$  )**: The two input trees have 4 units of saved credits of which 1 unit will be used to pay for the actual cost of linking, and 2 units will be saved as credit for the newly created tree. So, linking is still free, and it has 1 unused credit that can be used to pay for additional work if necessary.

**UNION(  $H_1, H_2$  )**: Simply concatenate the two root lists into one, and update the min pointer. Clearly, amortized cost =  $\Theta(1)$ .

**INSERT(  $H, x$  )**: This is MAKE-HEAP followed by a UNION. Hence, amortized cost =  $\Theta(1)$ .

# Binomial Heap Operations with Lazy Union

We maintain the following invariant:

$$\bigwedge_{B_j \in H} \text{credit}(B_j) = 2$$

**EXTRACT-MIN(  $H$  ):** Unlike in the array version, in this case we may have several trees of the same rank.

We create an array of length  $\lfloor \log_2 n \rfloor + 1$  with each location containing a *nil* pointer. We use this array to transform the linked list version to array version.

We go through the list of trees of  $H$ , inserting them one by one into the array, and linking and carrying if necessary so that finally we have at most one tree of each rank. We also create a min pointer.

We now perform EXTRACT-MIN as in the array case.

Finally, we collect the nonempty trees from the array into a doubly linked list, and return.

# Binomial Heap Operations with Lazy Union

We maintain the following invariant:

$$\bigwedge_{B_j \in H} \text{credit}(B_j) = 2$$

**EXTRACT-MIN(  $H$  ):** We only need to show that converting from linked list version to array version takes  $O(\log n)$  amortized time.

Suppose we start with  $t$  trees, and perform  $l$  links. So, we spend  $O(t + l)$  time overall.

As each link decreases the number of trees by 1, after  $l$  links we end up with  $t - l$  trees. Since at that point we have at most one tree of each rank, we have  $t - l \leq \lfloor \log_2 n \rfloor + 1$ .

Thus  $t + l = 2l + (t - l) = O(l + \log n)$ .

The  $O(l)$  part can be paid for by the  $l$  extra credits from  $l$  links.

We only charge the  $O(\log n)$  part to EXTRACT-MIN.

# Binomial Heap Operations with Lazy Union

We use exactly the same potential function as in the previous version,

$$\Phi(D_i) = c \times (\text{\#trees in the data structure after the } i\text{-th operation}),$$

where  $c$  is a constant.

As before, clearly,  $\Phi(D_0) = 0$

and for all  $i > 0$ ,  $\Phi(D_i) \geq 0$

**MAKE-HEAP(  $x$  ):**

actual cost,  $c_i = 1$  ( for creating the singleton heap )

potential change,  $\Delta_i = \Phi(D_i) - \Phi(D_{i-1}) = c$

( as #trees increases by 1 )

amortized cost,  $\hat{c}_i = c_i + \Delta_i = 1 + c = \Theta(1)$

# Binomial Heap Operations with Lazy Union

We use exactly the same potential function as in the previous version,

$$\Phi(D_i) = c \times (\text{\#trees in the data structure after the } i\text{-th operation}),$$

where  $c$  is a constant.

**UNION(  $H_1, H_2$  ):**

actual cost,  $c_i = 1$  ( for merging the two doubly linked lists )

potential change,  $\Delta_i = \Phi(D_i) - \Phi(D_{i-1}) = 0$

( no new tree is created or destroyed )

amortized cost,  $\hat{c}_i = c_i + \Delta_i = 1 = \Theta(1)$

# Binomial Heap Operations with Lazy Union

We use exactly the same potential function as in the previous version,

$$\Phi(D_i) = c \times (\text{\#trees in the data structure after the } i\text{-th operation}),$$

where  $c$  is a constant.

**INSERT(  $H, x$  ):**

Constant amount of work is done by MAKE-HEAP and UNION, and MAKE-HEAP creates a new tree.

$$\text{actual cost, } c_i = 1 + 1 = 2$$

$$\text{potential change, } \Delta_i = \Phi(D_i) - \Phi(D_{i-1}) = c$$

$$\text{amortized cost, } \hat{c}_i = c_i + \Delta_i = 2 + c = \Theta(1)$$

# Binomial Heap Operations with Lazy Union

We use exactly the same potential function as in the previous version,

$$\Phi(D_i) = c \times (\text{\#trees in the data structure after the } i\text{-th operation}),$$

where  $c$  is a constant.

## **EXTRACT-MIN( $H$ ):**

Cost of creating the array of pointers is  $\lfloor \log_2 n \rfloor + 1$ .

Suppose we start with  $t$  trees in the doubly linked list, and perform  $l$  link operations during the conversion from linked list to array version.

So we perform  $t + l$  work, and end up with  $t - l$  trees.

Cost of converting to the linked list version is  $t - l$ .

actual cost,  $c_i = \lfloor \log_2 n \rfloor + 1 + (t + l) + (t - l) = 2t + \lfloor \log_2 n \rfloor + 1$

potential change,  $\Delta_i = \Phi(D_i) - \Phi(D_{i-1}) = -c \times l$

# Binomial Heap Operations with Lazy Union

We use exactly the same potential function as in the previous version,

$$\Phi(D_i) = c \times (\text{\#trees in the data structure after the } i\text{-th operation}),$$

where  $c$  is a constant.

**EXTRACT-MIN(  $H$  ):**

actual cost,  $c_i = \lfloor \log_2 n \rfloor + 1 + (t + l) + (t - l) = 2t + \lfloor \log_2 n \rfloor + 1$

potential change,  $\Delta_i = \Phi(D_i) - \Phi(D_{i-1}) = -c \times l$

amortized cost,  $\hat{c}_i = c_i + \Delta_i = 2(t - l) + \lfloor \log_2 n \rfloor + 1 - (c - 2) \times l$

But  $t - l \leq \lfloor \log_2 n \rfloor + 1$  (as we have at most one tree of each rank)

So,  $\hat{c}_i \leq 3\lfloor \log_2 n \rfloor + 3 - (c - 2) \times l$

$\leq 3\lfloor \log_2 n \rfloor + 3$  (assuming  $c \geq 2$ )

$= O(\log n)$



# Binomial Heap Operations

Heap Operation	Worst-case	Amortized ( Eager Union )	Amortized ( Lazy Union )
MAKE-HEAP	$\Theta(1)$	$\Theta(1)$	$\Theta(1)$
INSERT	$O(\log n)$	$\Theta(1)$	$\Theta(1)$
MINIMUM	$\Theta(1)$	$\Theta(1)$	$\Theta(1)$
EXTRACT-MIN	$O(\log n)$	$O(\log n)$	$O(\log n)$
UNION	$O(\log n)$	$O(\log n)$	$\Theta(1)$