

# Amortized Analysis

Thomas Schwarz, SJ

# Amortized Analysis

- Determine the average time of a series of operations
  - Allows to optimize average performance
  - Example: Linear Hashing with fixed load factor  $\alpha$ .
    - Cost of bucket split is divided over  $\frac{1}{\alpha}$  inserts

# Amortized Analysis

- Aggregate Analysis
  - Determine upper bound  $T(n)$  on a sequence of  $n$  operations
  - Average cost is then  $T(n)/n$
- Accounting method:
  - Most operations gets overcharged
  - Accumulated overcharges are used to pay for later operations
- Potential method:
  - Model the “credit” as a potential

# Aggregate Analysis

- Stacks have
  - push
  - pop
  - empty
- Add a multipop( $k$ ) method that pops  $k$  elements (or empties the stack if there are less than  $k$  elements on the stack)
  - standard operations are  $O(1)$
  - multipop is worst case  $O(n)$

# Aggregate Analysis

- Given a sequence of  $m$  stack operations on an initially empty stack
  - Naïve calculation:
    - At most  $m$  elements on the stack at one time
    - Therefore, worst case cost is  $\sim m \times m = O(m^2)$
  - Better analysis:
    - Combined number of steps of pops and multi-pops is smaller or equal to the number of steps of pushes
    - Therefore: combined number of steps is at most  $m$
    - Therefore: amortized costs  $O(1)$

# Aggregate Analysis

- Counters implemented as binary array
  - Interested in calculating the bit-flips

0	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---

- Single increment can flip all the bits
- Array with  $k$  binary registers has  $2^k - 1$  increments
  - Does this mean that number of bit-flips for  $k$  increments is  $\Theta(k)$  per increment?

# Aggregate Analysis

- Write down binary numbers in order
  - Observe
    - Least Significant Bit (LSB) flips every time
    - Second LSB flips every other time
    - Third LSB flips every  $2^2$  times
    - ...
- ```
000000
000001
000010
000011
000100
000101
000110
000111
001000
001001
001010
001011
001100
001101
001110
```

# Aggregate Analysis

- Assume  $n$  increment operations with an arbitrary starting counter value
- Then number of bit-flips is less than

- $\sum_{i=0}^{k-1} \lfloor \frac{n}{2^i} \rfloor$

- $< \sum_{i=0}^{k-1} \frac{n}{2^i}$

- $< n \sum_{i=0}^{\infty} \frac{1}{2^i}$

- $= n \times 0.11111\dots_2 = 2n$



# Aggregate Analysis

- Aggregate cost of  $n$  increment operations is therefore
  - $< 2n$
  - $= \Theta(n)$
- Average cost per increment is 2

# Accounting Method

- Stack costs:
  - Push: 1
  - Pop: 1
  - Multipop( $k$ ):  $\min(k, s)$  where  $s$  is the number of elements in the stack
- Charges:
  - Push: 2
  - Pop: 0
  - Multipop( $k$ ): 0

# Accounting Method

- Show that charges will pay for all operations.
  - push pays for itself and for removing the element
  - since we cannot remove elements that have not been pushed, charged amount is always sufficient

# Accounting Method

- Counter:
    - Charge 2 for every bit set to one
    - This allows us to set the bit and to reset the bit
  - To calculate costs of increment operation:
    - Observe: increment only sets one bit to 1.
  - Therefore:
    - $n$  increments cost at most  $2n$  bit-flips
- ```
000000
000001
000010
000011
000100
000101
000110
000111
001000
001001
001010
001011
001100
001101
001110
001111
010000
010001
010010
...
```

# Potential Method

- Represent charges as potential energy
  - Potential function  $\Phi$  maps each state of the data structure to a number
  - Amortized cost of an operation:
    - actual cost + change in potential
- Implies:
  - Amortized costs of  $n$  operations
    - actual costs of  $n$  operations + change in potential

# Potential Method

- Stack:
  - Potential = number of elements in stack
    - Potential can never be less than zero
  - Amortized cost of a stack operation
    - Push: cost plus change in potential =  $1 + 1$
    - Multipop / pop: cost plus change in potential =  $k - k$

# Potential Method

- Counter:
  - Potential = number of bits set (= number of one bits)
  - Amortized cost of an increment:
    - If  $i^{\text{th}}$  increment resets  $t_i$  bits:
      - Cost is  $1 + t_i$
      - Amortized cost is
        - cost + potential change  
 $\leq (t_i + 1) + (1 - t_i) = 2$

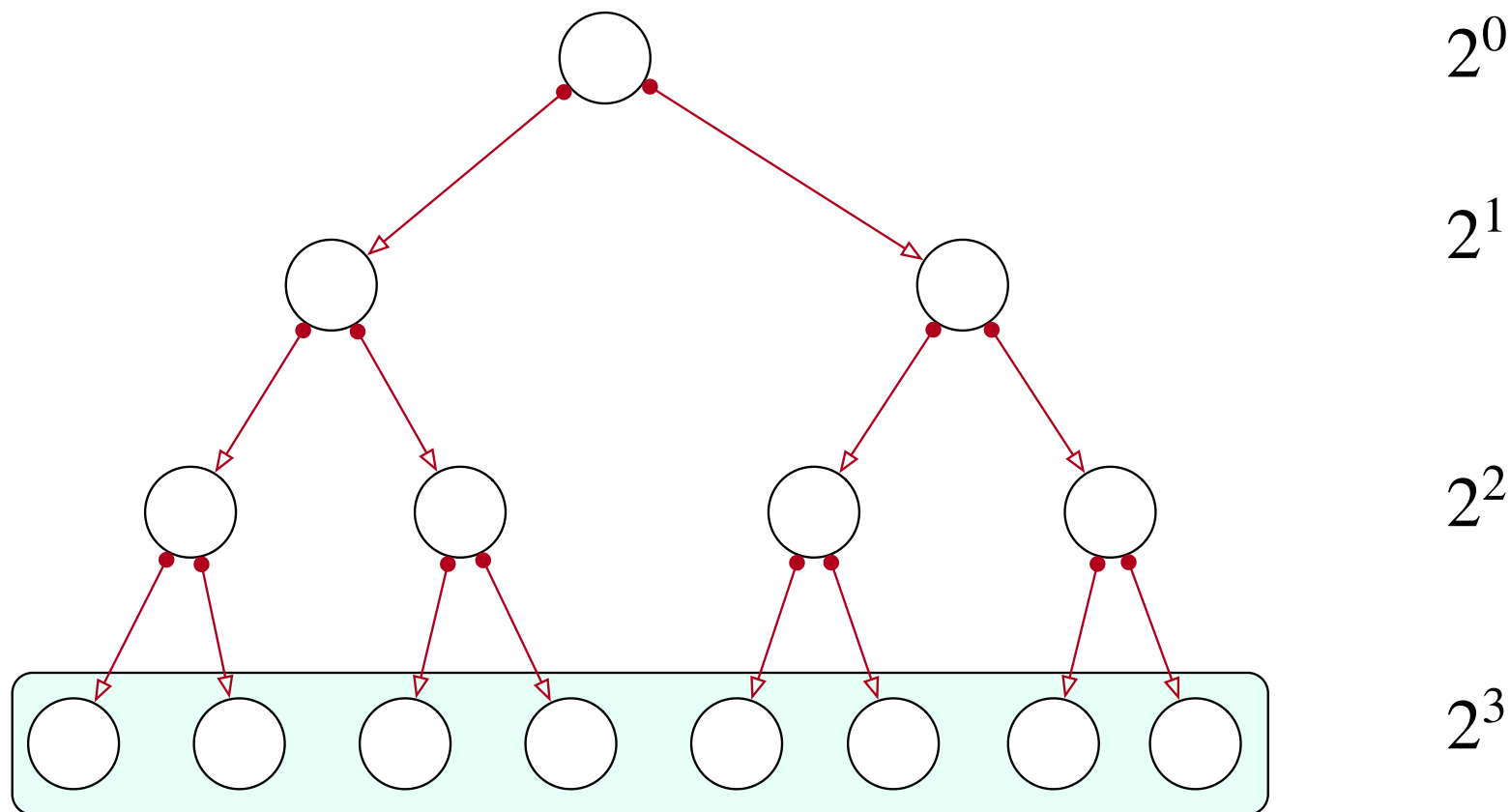
# Binary Trees and Heaps

Thomas Schwarz, SJ



# Behavior of Trees

- A full binary tree of depth  $n$  has
  - $1 + 2 + 2 \cdot 2 + 2 \cdot 2 \cdot 2 + \dots + 2^{n-1}$
  - $= (111\dots1)_2 = 2^n - 1$  places



# Behavior of Trees

- Reversely:
  - To store  $m$  elements in a binary tree:
    - Need a tree of depth  $d$  such that
      - $2^{(d-1)} - 1 \leq m < 2^d - 1$
    - Equivalent to
      - $2^{d-1} \leq m + 1 < 2^d$
      - $d - 1 \leq \log_2(m + 1) < d$
      - $d - 1 = \lfloor \log_2(m + 1) \rfloor$

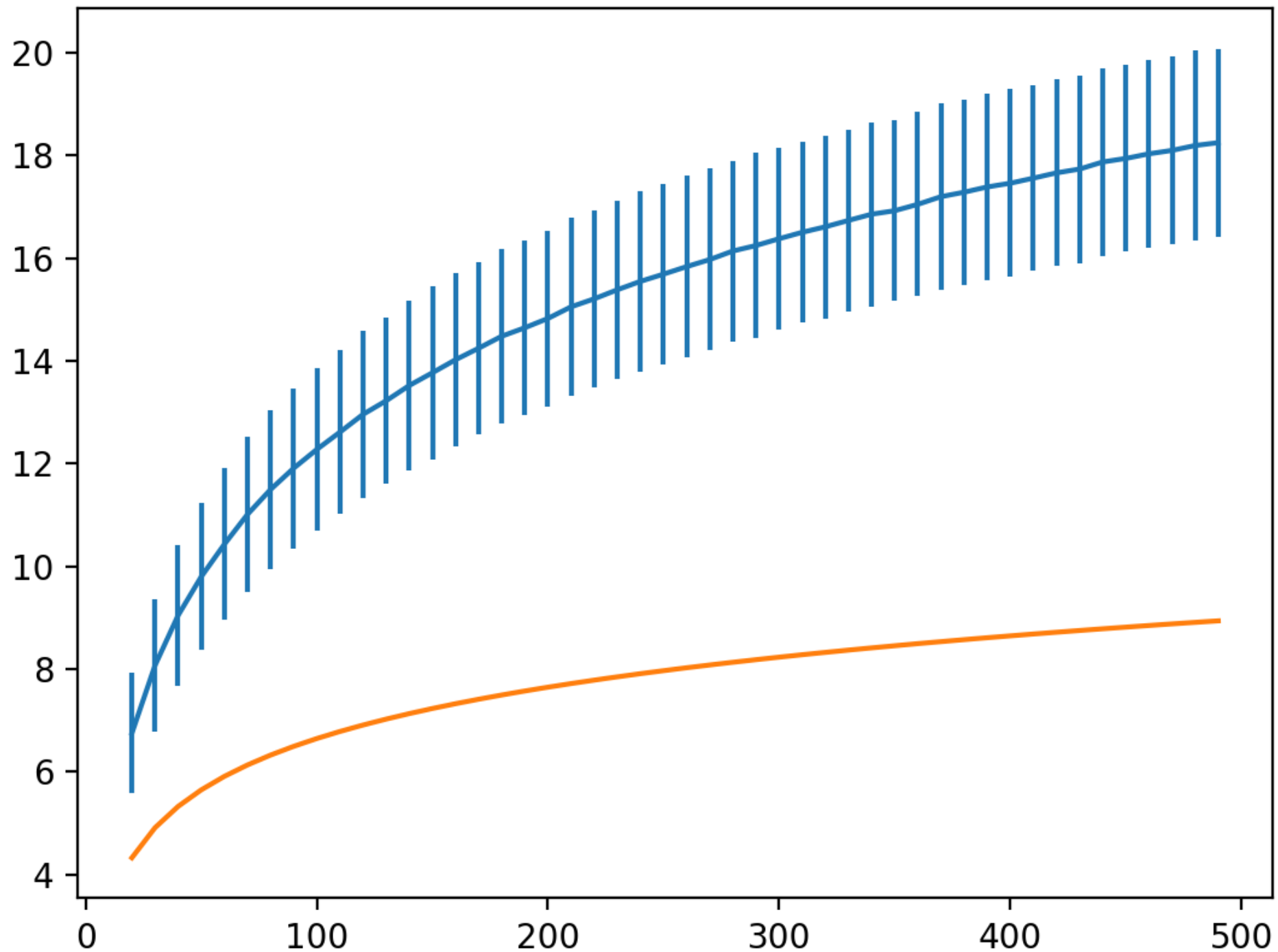
# Behavior of Trees

- This parsimony is not natural
  - Random inserts: Trees have much larger depth
  - Self-modifying trees restructure themselves in order to get closer
- Importance:
  - Searching an element takes time  $\sim$  to depth
  - Inserting an element takes time  $\sim$  to depth

# Behavior of Trees

- Experiment:
  - Insert  $n$  elements into a binary tree
  - Get the depth
  - Repeat 10,000 times
  - Depict mean plus/minus standard deviation

# Behavior of Trees



# Behavior of Trees

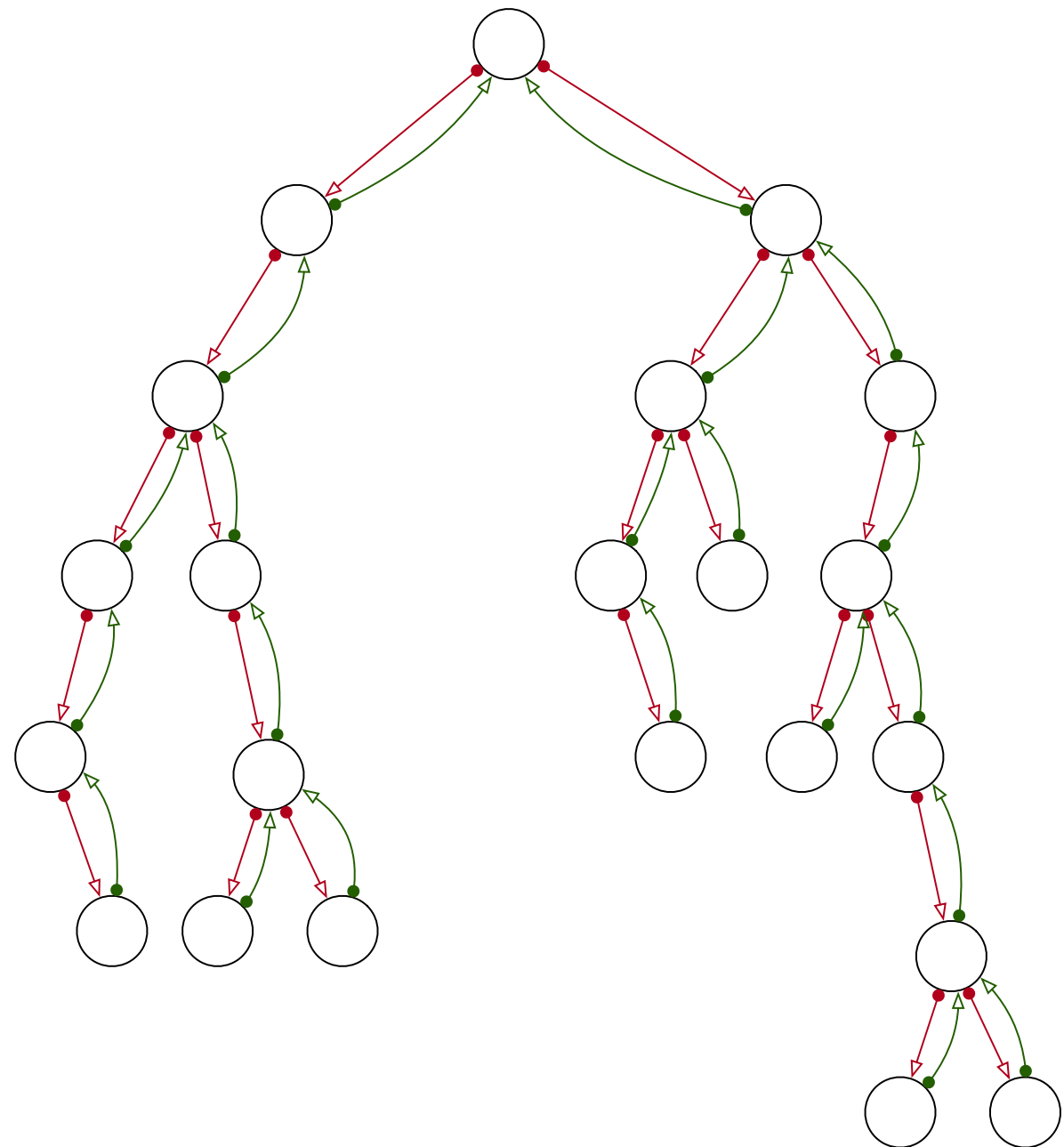
- On average
  - Trees have more than twice necessary depth
- But on average
  - Behavior is still logarithmic
- Theory: Height of a random binary search tree for a random permutation of  $n$  elements is
  - $\alpha \log_e(n)$  with  $\alpha \approx 4.31107$
  - $= 2.98821 \log_2(n)$ 
    - Robson 1979 / Devroye 1986

# Decorating Binary Search Trees

- General principle for Data Structures:
  - Can store more information in order to improve performance
- Example:
  - Removal of elements from a binary search tree
    - Difficult because we need to find parent
    - Can be made simpler by having a parent pointer

# Binary Trees with Parent Link

- Each node stores a link to the parent
- For root, link is None
- Faster deletes at the cost of more storage per node





# Binary Trees with Parent Link

- Expand to a key-value store by adding a field for record
- Add a parent link

```
class Node:
    def __init__(self, value, record):
        self.value = value
        self.record = record
        self.up, self.left, self.right = None, None, None

    def __repr__(self):
        return "Node : {}, Value: {}, Record: {},
                Left: {}, Right: {}, Up: {}".format(
                    hex(id(self)), self.value, self.record,
                    hex(id(self.left)), hex(id(self.right)),
                    hex(id(self.up)))
```

# Binary Trees with Parent Link

- We have to maintain the up link:

```
def insert(self, value, record):
    new_node = Node(value, record)
    if not self.root:
        self.root = new_node
    else:
        current = self.root
        while True:
            if value < current.value:
                if current.left:
                    current = current.left
                else:
                    current.left = new_node
                    new_node.up = current
            return
```

# Binary Trees with Parent Link

- But deleting a record is still not trivial
  - Special case when
    - the tree is empty

```
def remove(self, value):  
    if not self.root:  
        return False
```

# Binary Trees with Parent Link

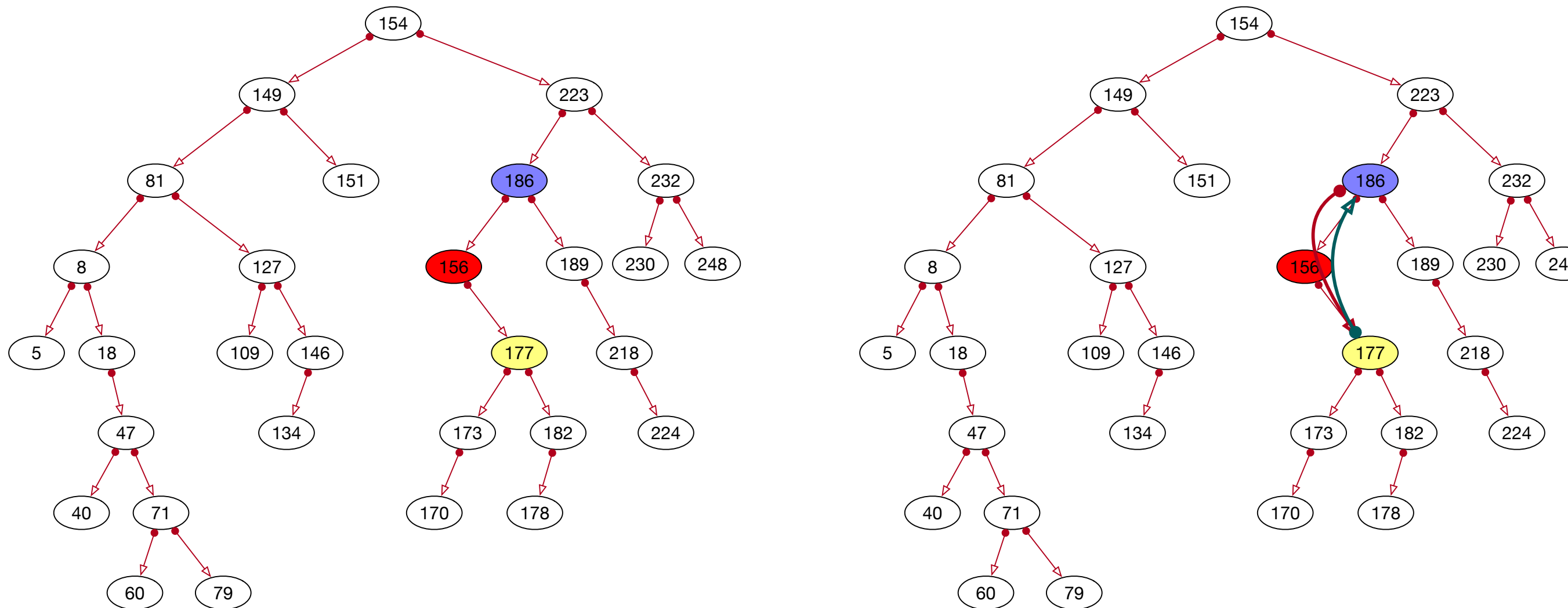
```
def remove(self, value):
    if not self.root:
        return False
    current = self.root
    while True:
        if not current:
            return False
        if value == current.value:
            break
        if value < current.value:
            current = current.left
        else:
            current = current.right
    if current == None:
        return False
    to_delete = current
```

# Binary Trees with Parent Link

- We still need to make additional case distinctions
  - But we no longer need a stack to keep track of the nodes
  - Case distinctions:
    - No children:
      - Just delete (unless we are deleting the root)
    - One child
    - Two children

# Binary Trees with Parent Link

- Removing node with one child
- Move child up and reset **two** links



# Binary Trees with Parent Link

- Special case if parent is root

```
elif not to_delete.left and to_delete.right:
    # node has only a right child
    parent = to_delete.up
    if not parent:
        self.root = to_delete.right
        return True
    else:
        if parent.left == to_delete:
            parent.left = to_delete.right
            to_delete.right.up = parent
        else:
            parent.right = to_delete.right
            to_delete.right.up = parent
    return True
```

# Binary Trees with Parent Link

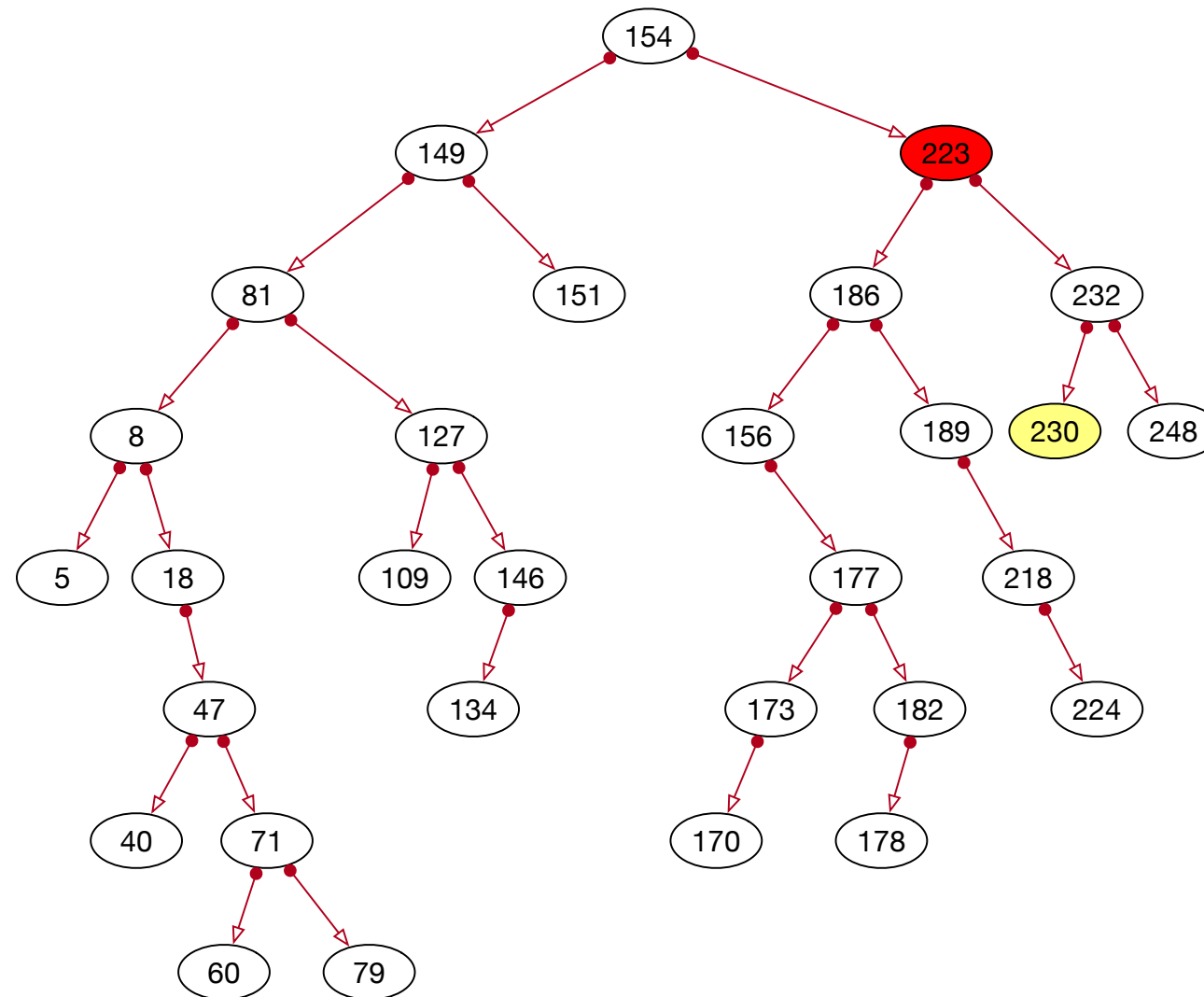
- Otherwise: reset two links

```
elif not to_delete.left and to_delete.right:
    # node has only a right child
    parent = to_delete.up
    if not parent:
        self.root = to_delete.right
        return True
    else:
        if parent.left == to_delete:
            parent.left = to_delete.right
            to_delete.right.up = parent
        else:
            parent.right = to_delete
            to_delete.right.up = parent
        return True
```



# Binary Trees with Parent Link

- Two children:
  - Identify the next node in-order traversal



# Binary Trees with Parent Link

- Two children:
  - Find the next node in in-order traversal:
    - Go to the right: `current.right`
    - Then go always to the left

```
def min_value_node(a_node):  
    current = a_node  
    while current.left:  
        current = current.left  
    return current
```

# Binary Trees with Parent Link

- Two nodes

```
elif to_delete.left and to_delete.right:
    #node has two children
    leaf = Binary_Tree.min_value_node(
                                     to_delete.right)
    save_value = leaf.value
    save_record = leaf.record
    self.remove(leaf.value)
    to_delete.value = save_value
    to_delete.record = save_record
```

# Binary Trees with Parent Link

- Safe the values of the resulting leaf

```
elif to_delete.left and to_delete.right:
    #node has two children
    leaf =
        Binary_Tree.min_value_node(to_delete.right)
    print('leaf', leaf)
    save_value = leaf.value
    save_record = leaf.record
    self.remove(leaf.value)
    to_delete.value = save_value
    to_delete.record = save_record
```

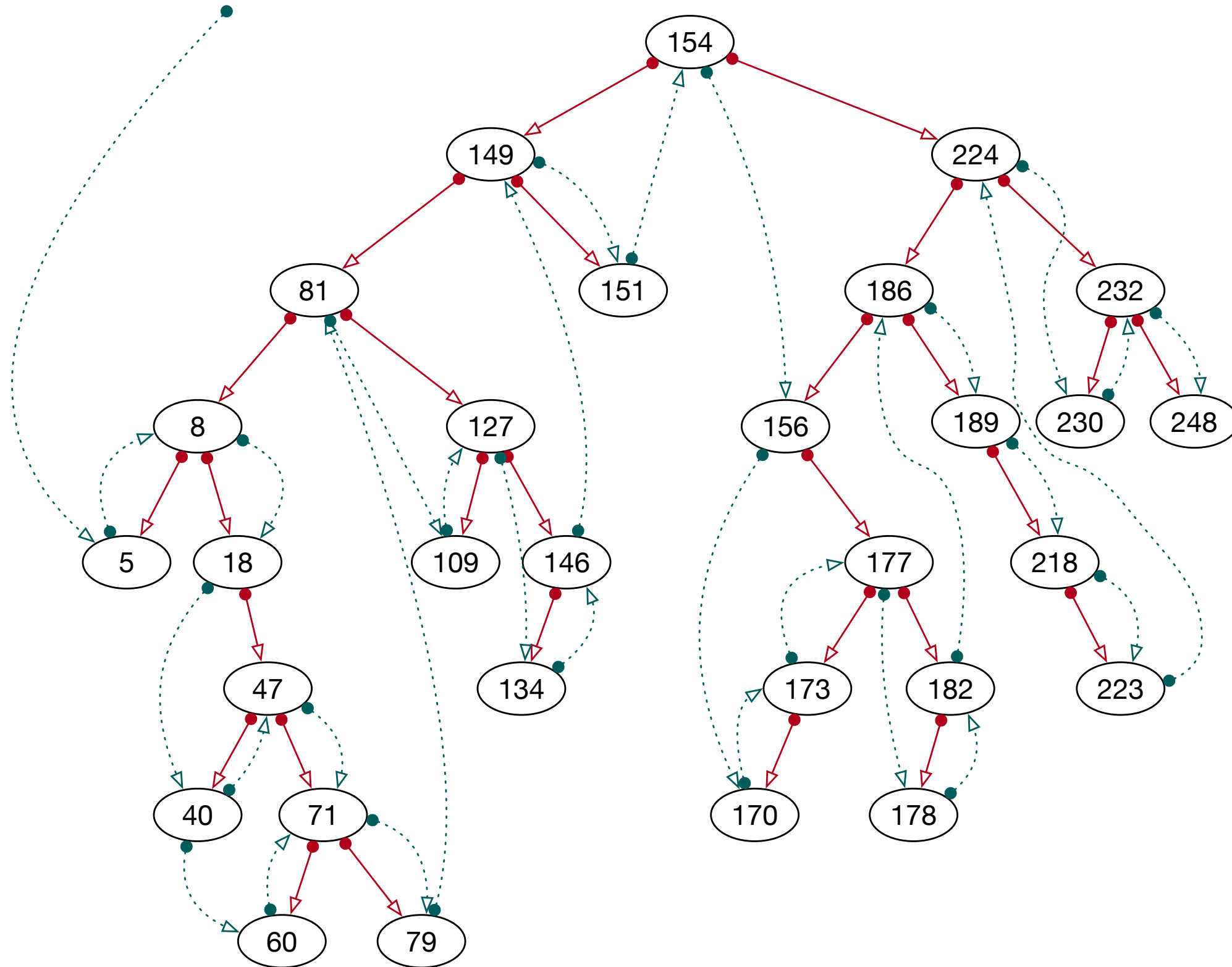
# Binary Trees with Parent Link

- Then delete the leaf
  - I cheat by using recursion

```
elif to_delete.left and to_delete.right:
    #node has two children
    leaf =
        Binary_Tree.min_value_node(to_delete.right)
    print('leaf',leaf)
    save_value = leaf.value
    save_record = leaf.record
    self.remove(leaf.value)
    to_delete.value = save_value
    to_delete.record = save_record
```

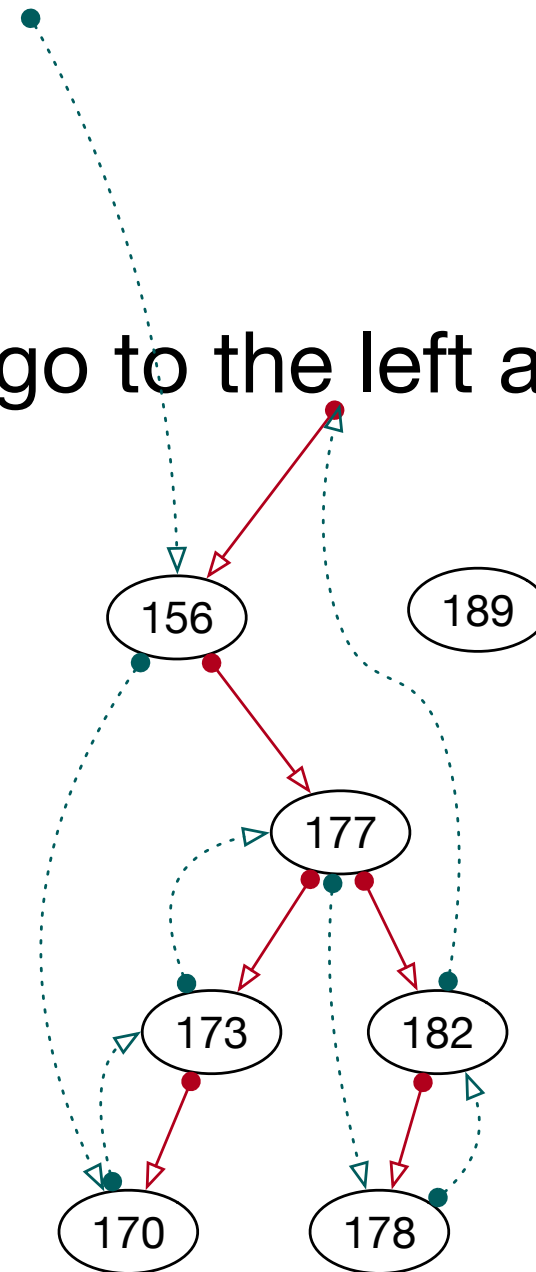
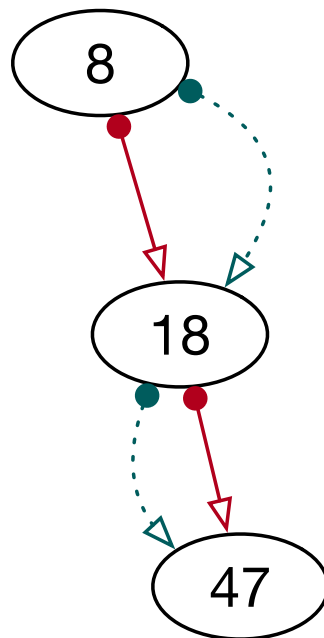
# Binary Trees with Parent Link

- Non-recursive in-order traversal
  - Here is a tree with an additional set of links for in-order traversal



# Binary Trees with Parent Link

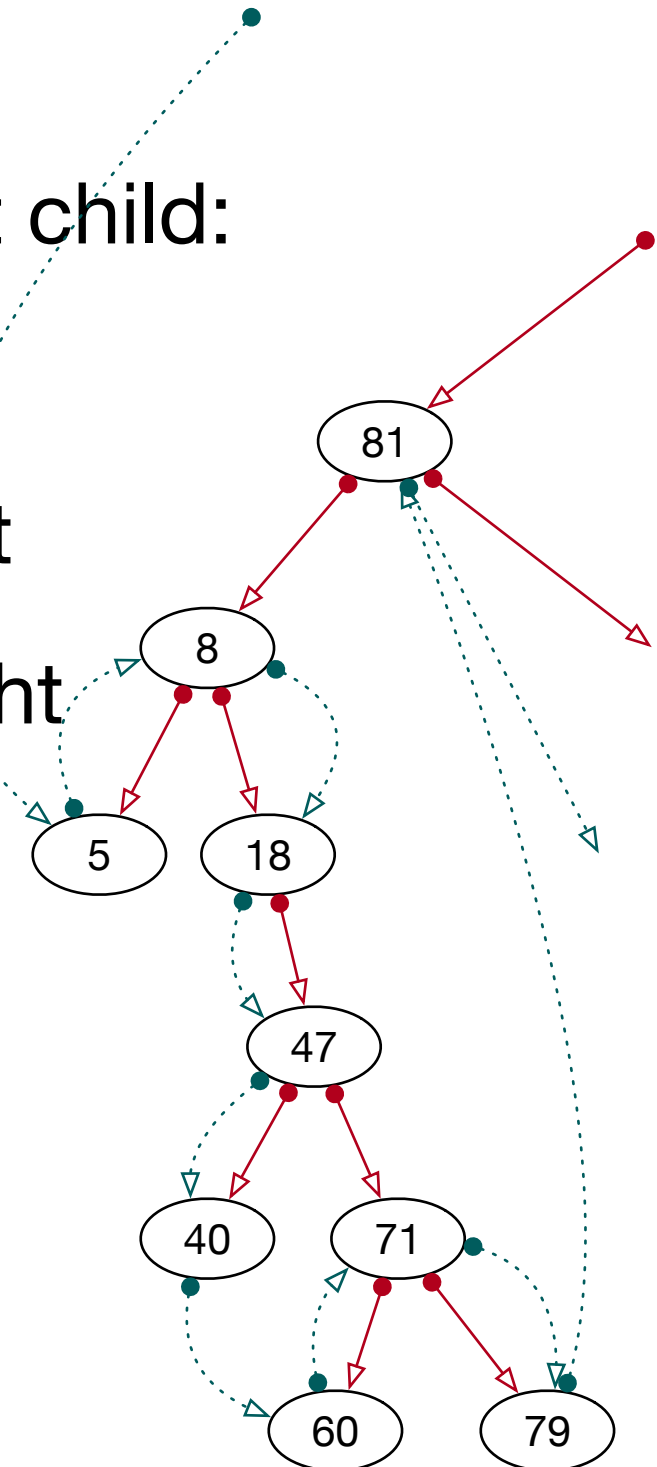
- What is the next node:
  - If the node has a right child:
    - Go one to the right, then go to the left as much as possible





# Binary Trees with Parent Link

- What is the next node if there is no right child:
  - If parent is to the left:
    - Follow parents if they are to the left
    - Then take the first parent to the right



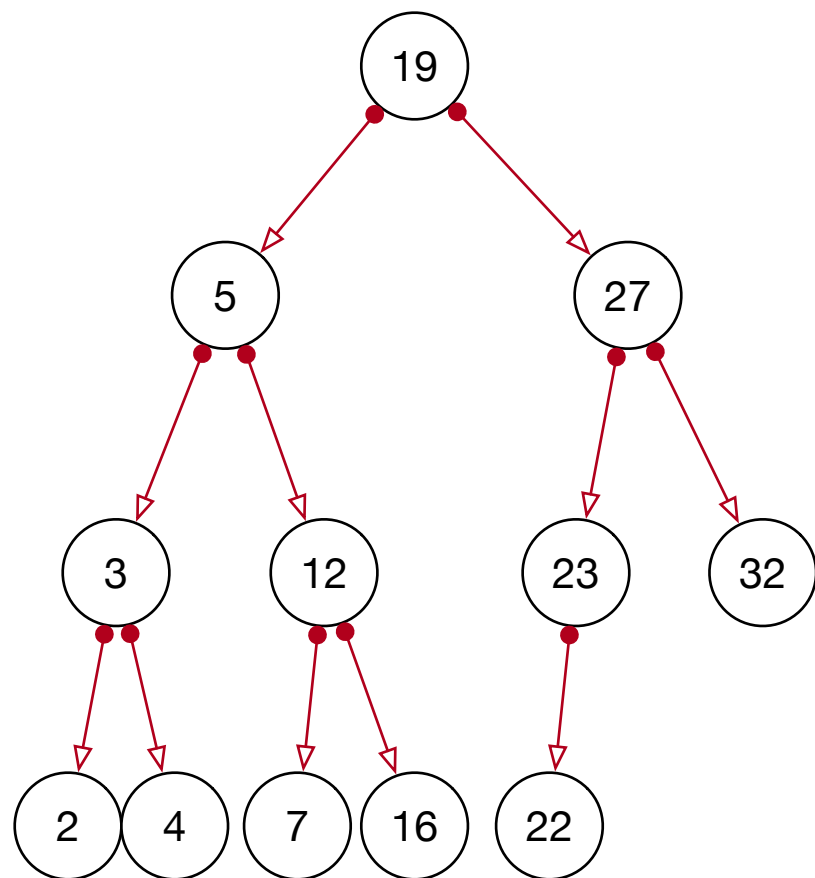
# Binary Trees with Parent Link

- Thus:
  - Can do in-order traversal without a stack or recursion

# Binary Trees using Arrays

# Using Arrays

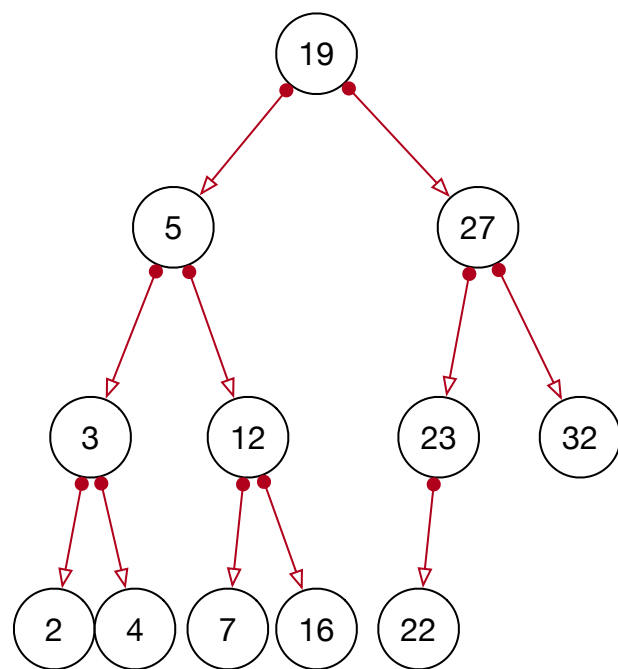
- In a tree, each node has up to two children
  - Can organize nodes in an array
    - Leave first spot open



—	19	5	27	3	12	23	32	2	4	7	16	22
	1	2	3	4	5	6	7	8	9	10	11	12

# Using Arrays

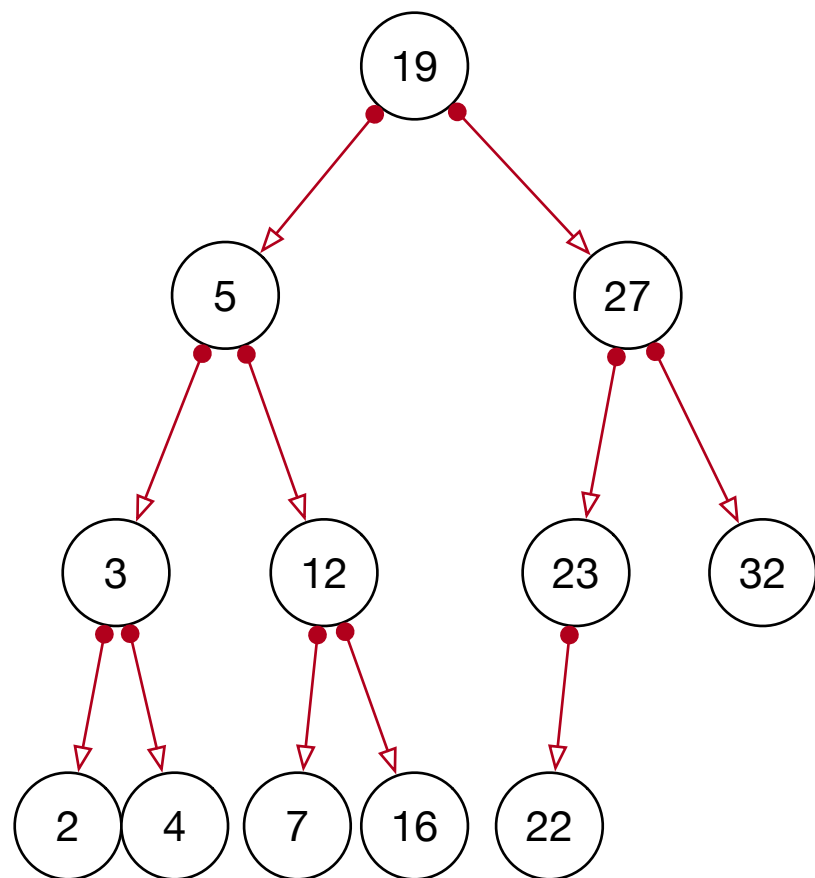
- Left child of node at index  $i$ 
  - Located at index  $2i$
- Right child of node at index  $i$ 
  - Located at index  $2i + 1$



—	19	5	27	3	12	23	32	2	4	7	16	22
	1	2	3	4	5	6	7	8	9	10	11	12

# Using Arrays

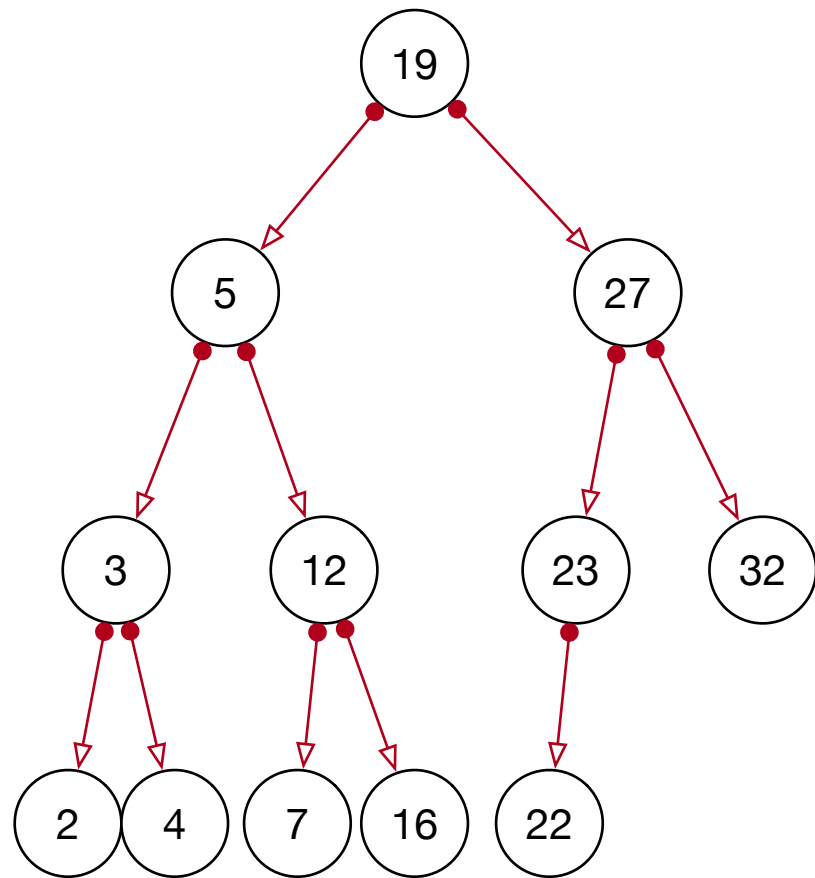
- Parent of node at index  $i$  is located at index  $i//2$ 
  - Mathematical notation:  $\lfloor \frac{i}{2} \rfloor$



—	19	5	27	3	12	23	32	2	4	7	16	22
	1	2	3	4	5	6	7	8	9	10	11	12

# Using Arrays

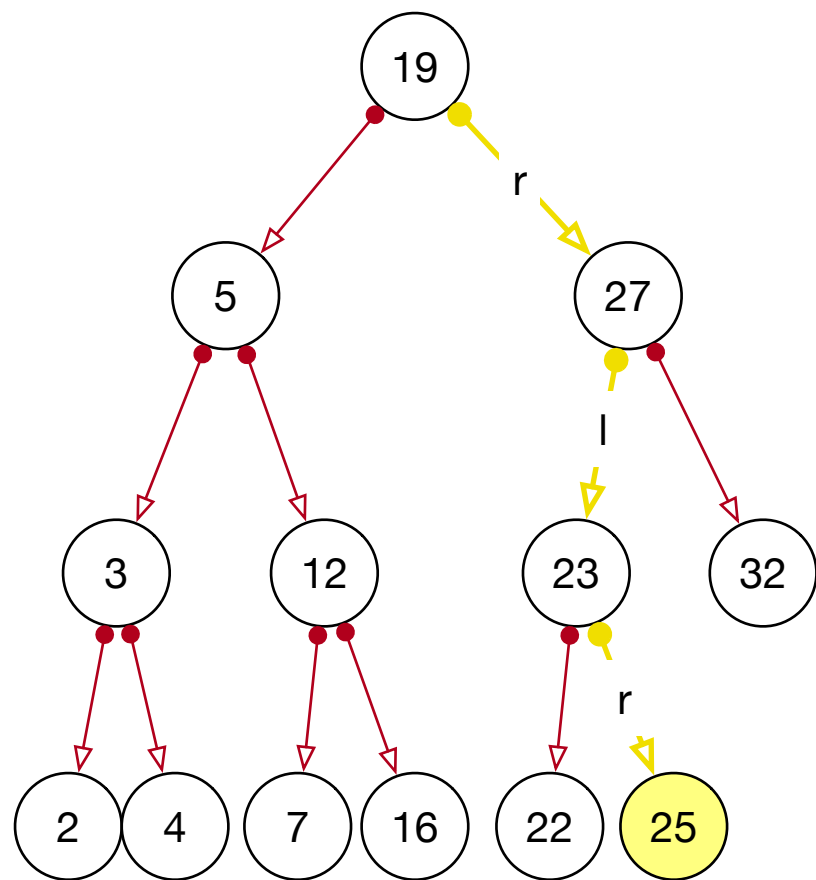
- Right children are at odd indices, left children are even indices



—	19	5	27	3	12	23	32	2	4	7	16	22
	1	2	3	4	5	6	7	8	9	10	11	12

# Using Arrays

- We can calculate the index if we are given a sequence of directions



—	19	5	27	3	12	23	32	2	4	7	16	22	25
	1	2	3	4	5	6	7	8	9	10	11	12	13

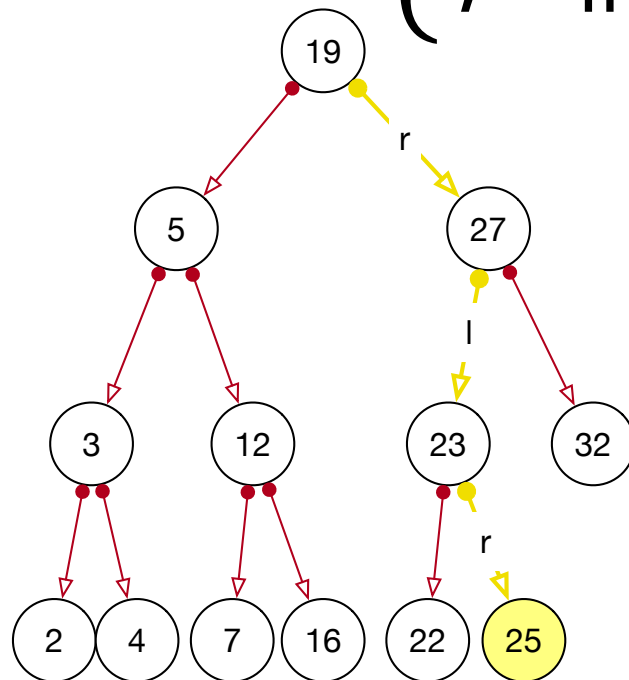
rlr  $((1*2+1)*2)*2+1 = 13$



# Using Arrays

- Define  $r(n) := 2n + 1$ ,  $l(n) := 2n$
- Then node is at index  $(o_m \circ o_{m-1} \circ \dots \circ o_2 \circ o_1)(1)$

- where  $o_i = \begin{cases} l & \text{if we go left in step } i \\ r & \text{if we go right in step } i \end{cases}$



—	19	5	27	3	12	23	32	2	4	7	16	22	25
	1	2	3	4	5	6	7	8	9	10	11	12	13

rlr  $((1*2+1)*2)*2+1 = 13$

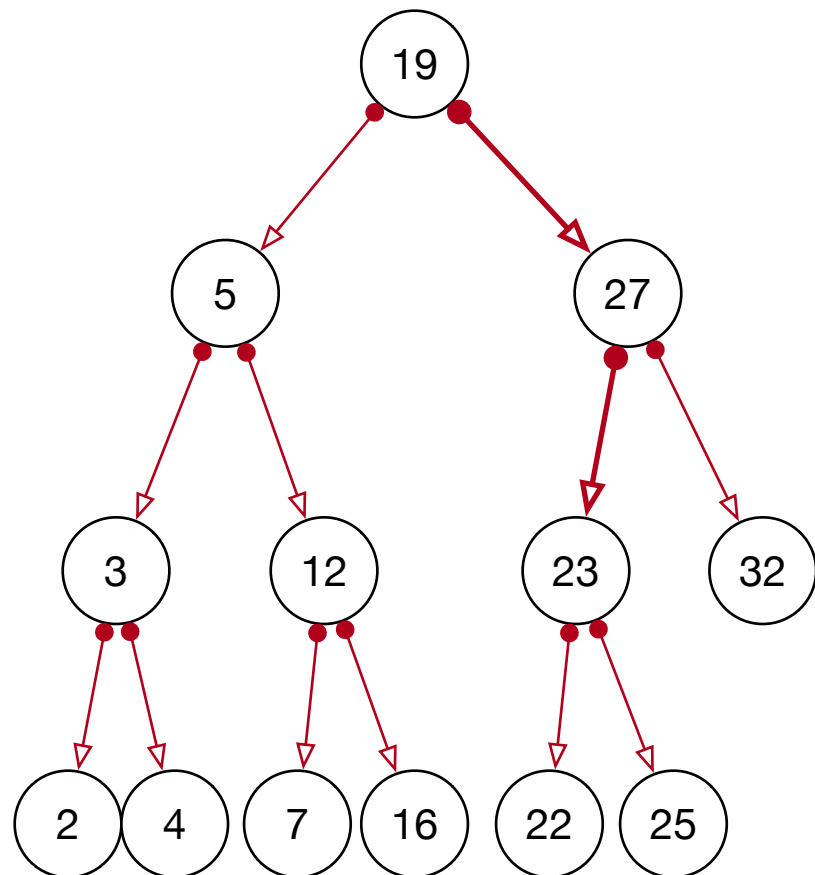
$r \circ l \circ r(1)$

# Using Arrays

- Can we do something about the unused first element in the array?
  - We just need to adjust the index: by adding 1 and subtracting 1

# Using Arrays

- Children of node  $i$  are now  $2 \cdot (i + 1) - 1 = 2 \cdot i + 1$  and  $(2 \cdot (i + 1) + 1) - 1 = 2 \cdot i + 2$

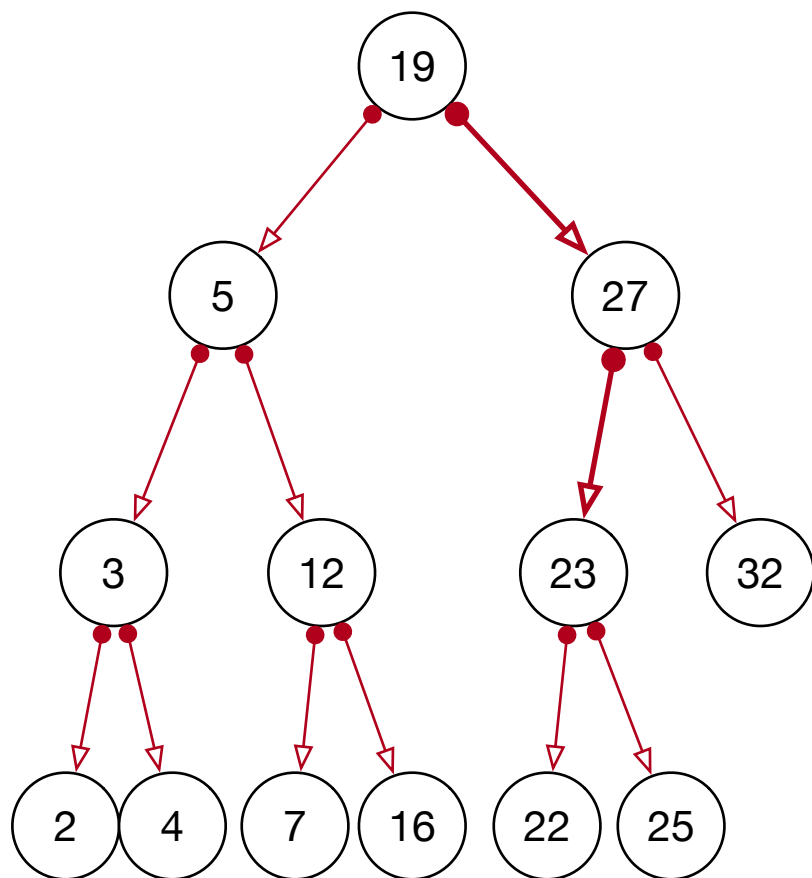


19	5	27	3	12	23	32	2	4	7	16	22	25
0	1	2	3	4	5	6	7	8	9	10	11	12

# Using Arrays

- Parent of a node located at index  $i$  is located

- at index  $\lfloor \frac{i+1}{2} \rfloor - 1$



19	5	27	3	12	23	32	2	4	7	16	22	25
0	1	2	3	4	5	6	7	8	9	10	11	12

# Using Arrays

- One advantage:
  - We automatically have a way to find the parent

# Priority Queue

- ADS with
  - Insertion
  - Popping maximum element
- Example: insert 5, insert 4, insert 10, pop, insert 7, insert 3, pop, insert 2, pop, pop
  - Returns on insert 5, insert 4, insert 10, **pop**, insert 7, insert 3, pop, insert 2, pop, pop: 10
  - Returns on insert 5, insert 4, insert 10, pop, insert 7, insert 3, **pop**, insert 2, pop, pop: 7
  - Returns on insert 5, insert 4, insert 10, pop, insert 7, insert 3, pop, insert 2, **pop**, pop: 5
  - Returns on insert 5, insert 4, insert 10, pop, insert 7, insert 3, pop, insert 2, pop, **pop**: 4

# Priority Queues

- Simplistic implementation
  - A list
    - Whenever we look for an element, we look for the maximum of the list
    - Run time: Proportional to the length of the list

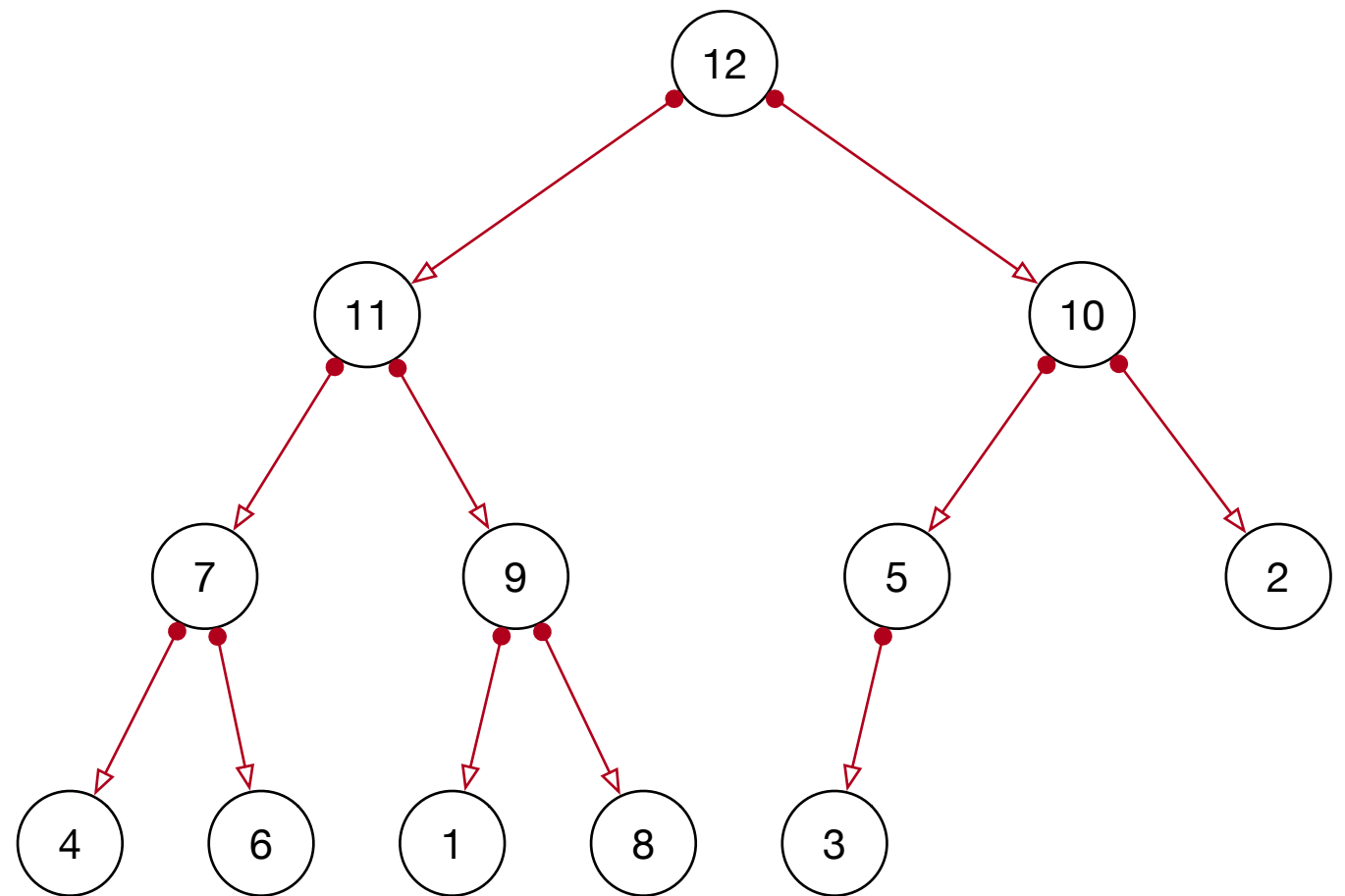
# Priority Queues

- Favorite implementation:
  - Heap:
    - A ***complete*** binary tree
      - Tree is maximum balanced
    - That is **partially** ordered



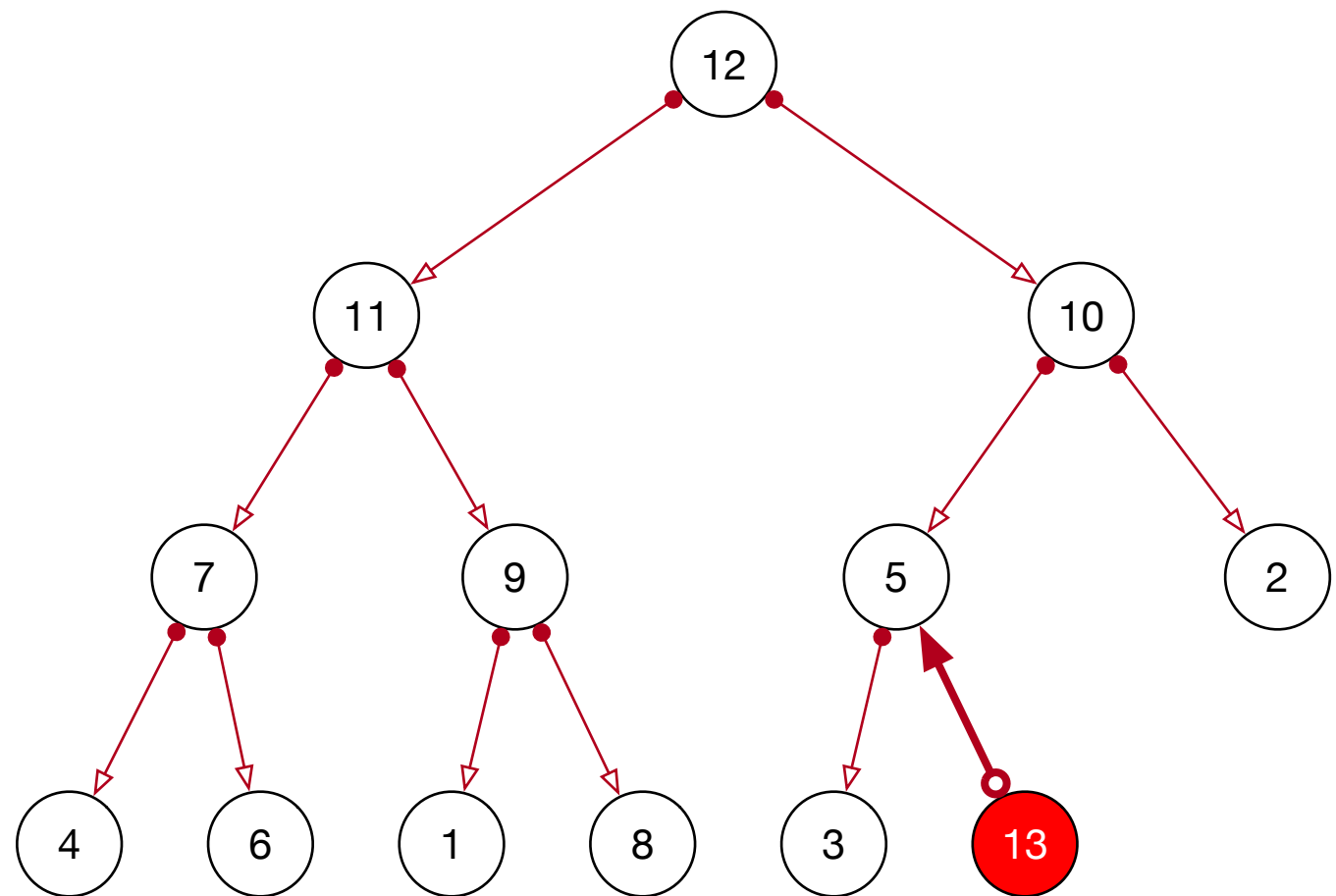
# Priority Queues

- Heaps as binary tree
  - Complete:
    - No nodes missing
    - Last generation filled from left
  - Partially ordered:
    - parent has larger value than child



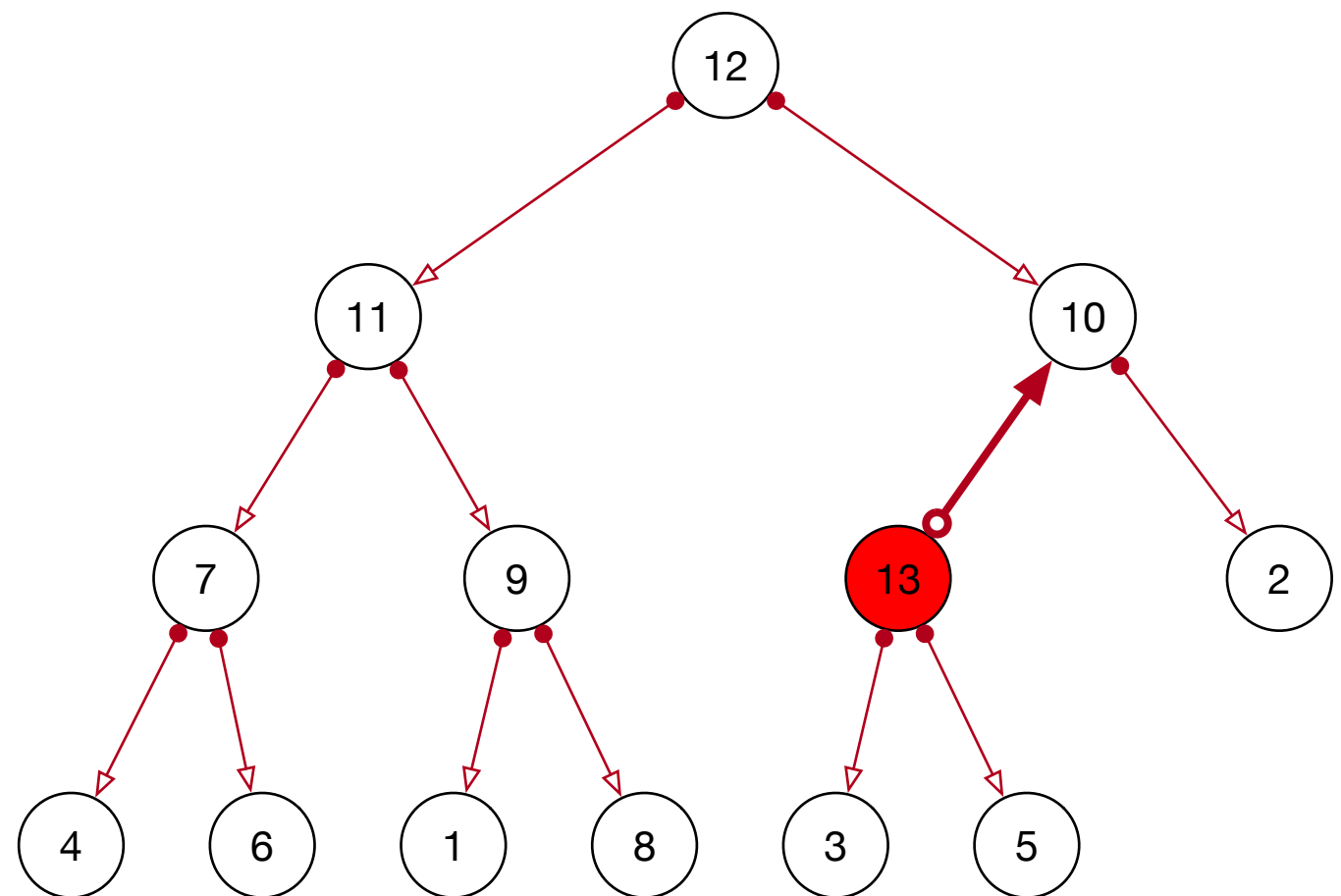
# Priority Queues

- Operations: Insertion
  - Insert at the next spot
  - If the new node is larger than the parent:
    - swap with parent



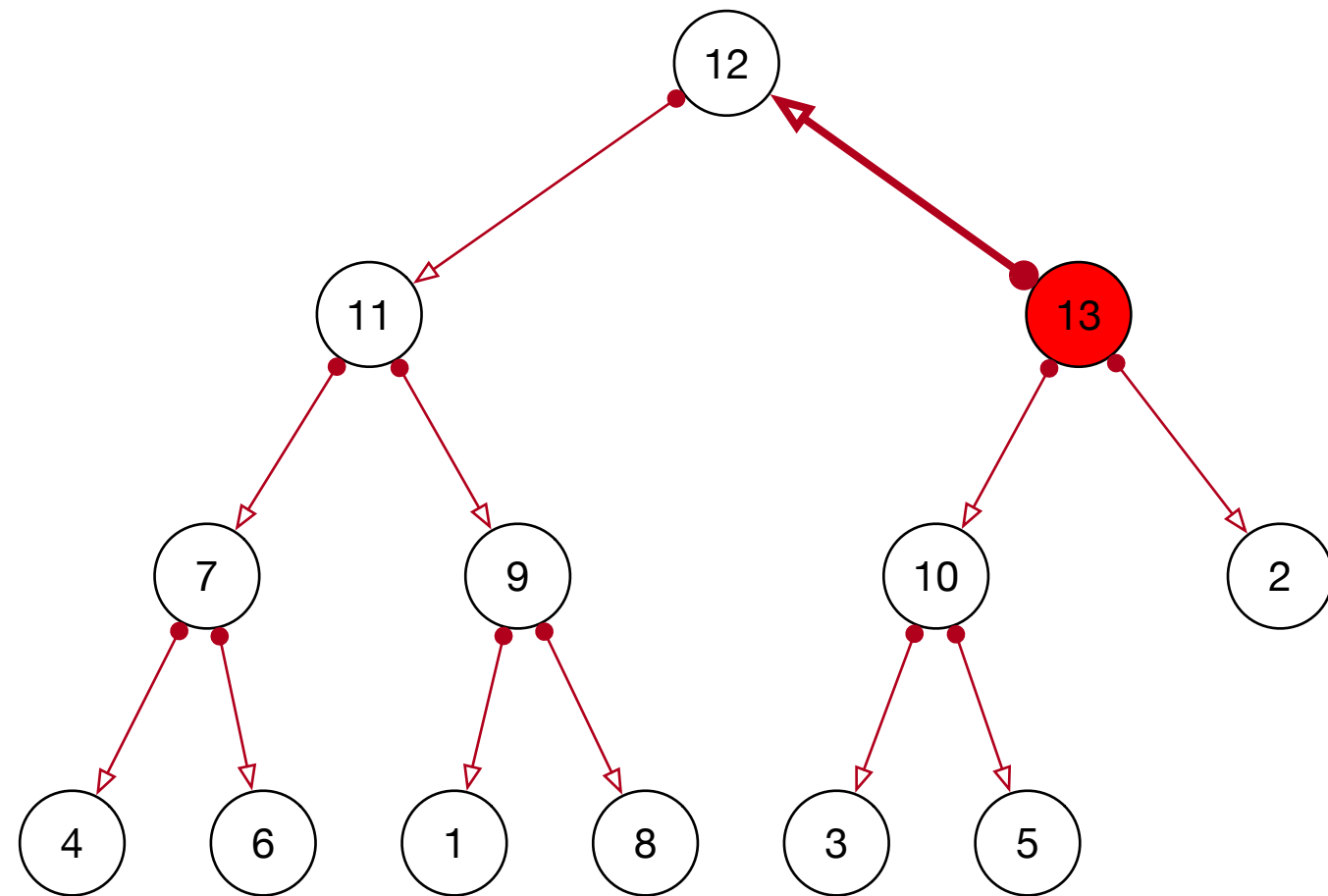
# Priority Queues

- This is repeated
  - if necessary



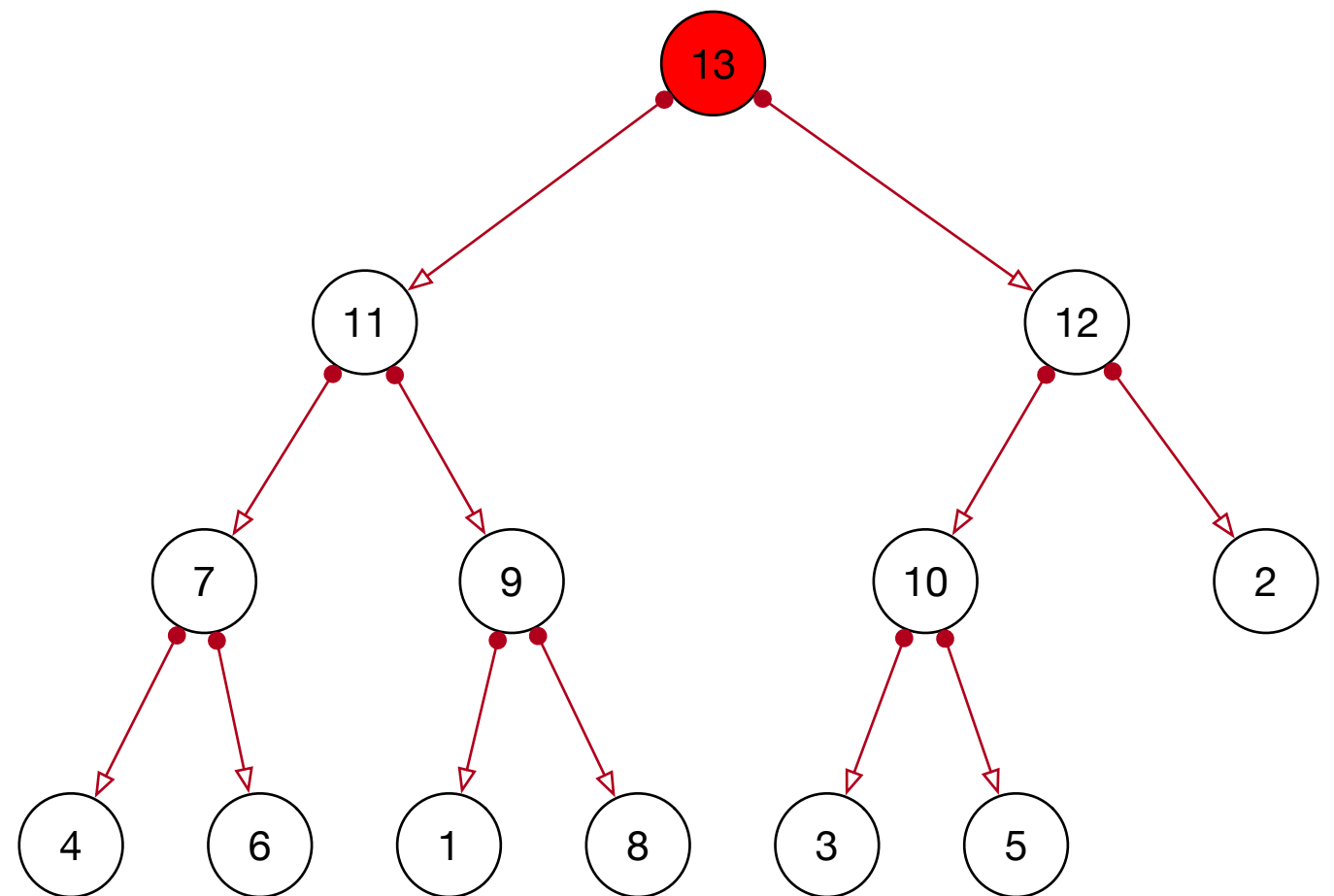
# Priority Queues

- Notice:
  - The only violation of order can be with parent



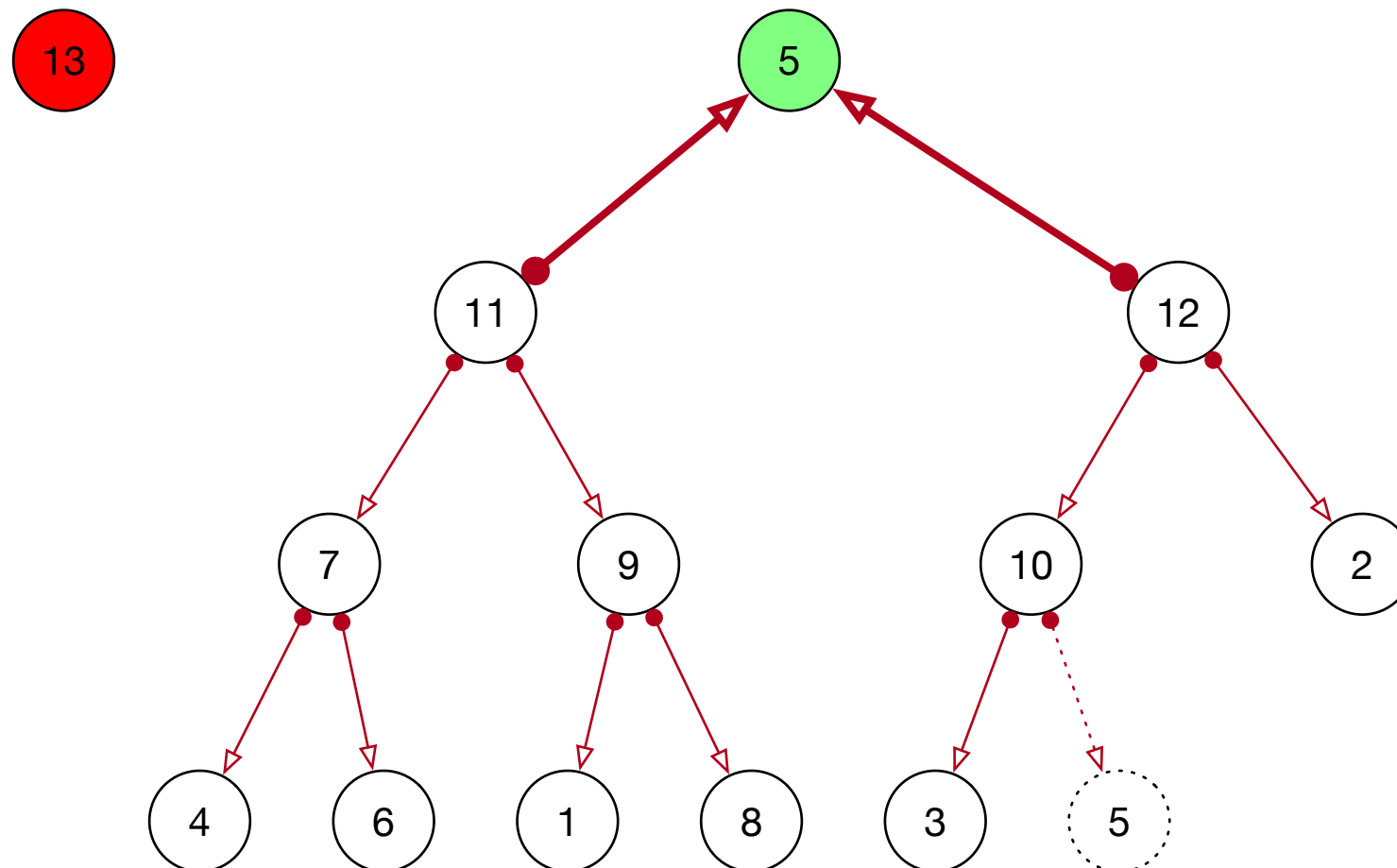
# Priority Queues

- There are at most  $\log_2(n)$  swaps
- Compared to  $n$



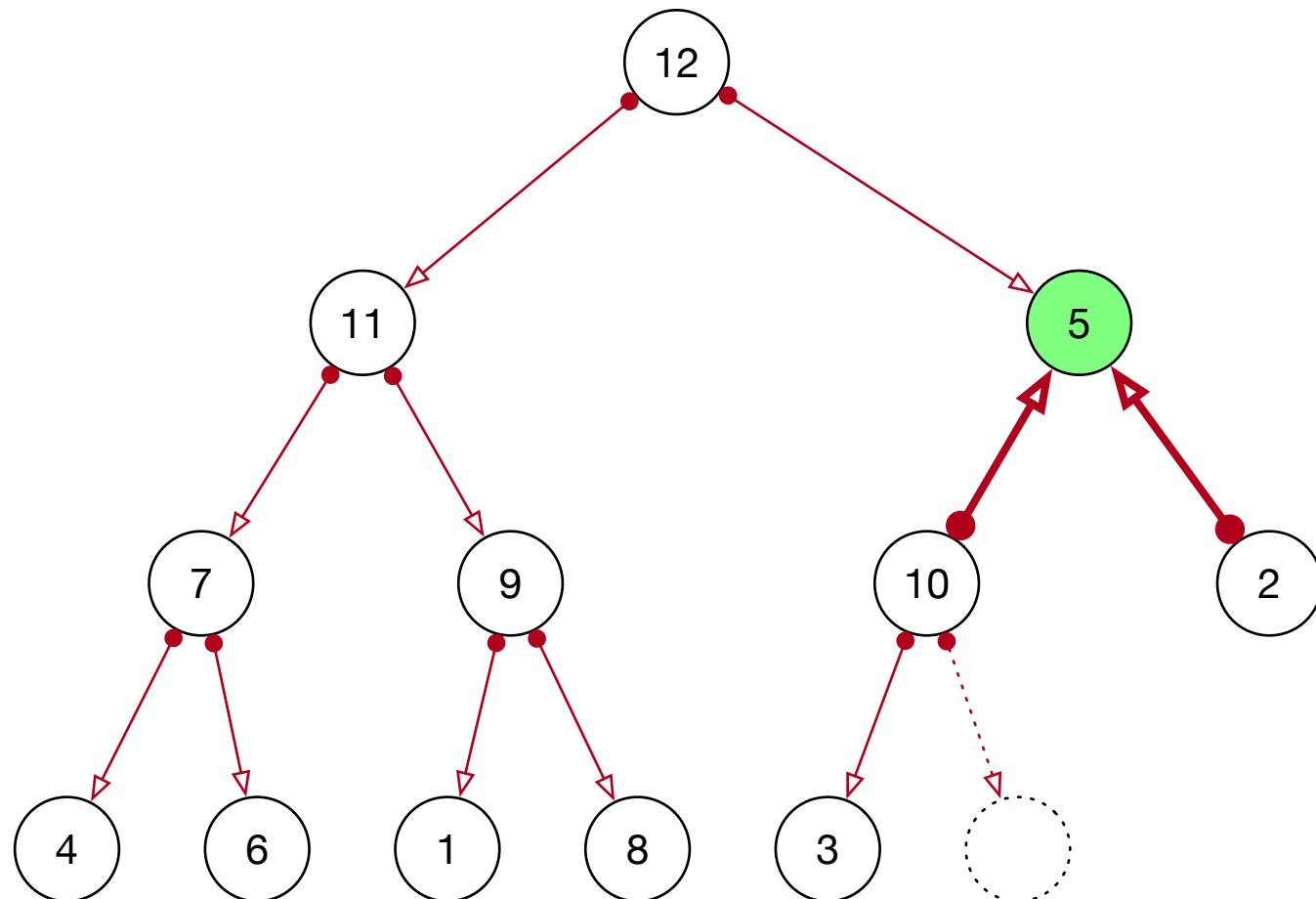
# Priority Queues

- Remove Maximum:
  - Maximum is at the top, remove it
  - Move last element into the top position



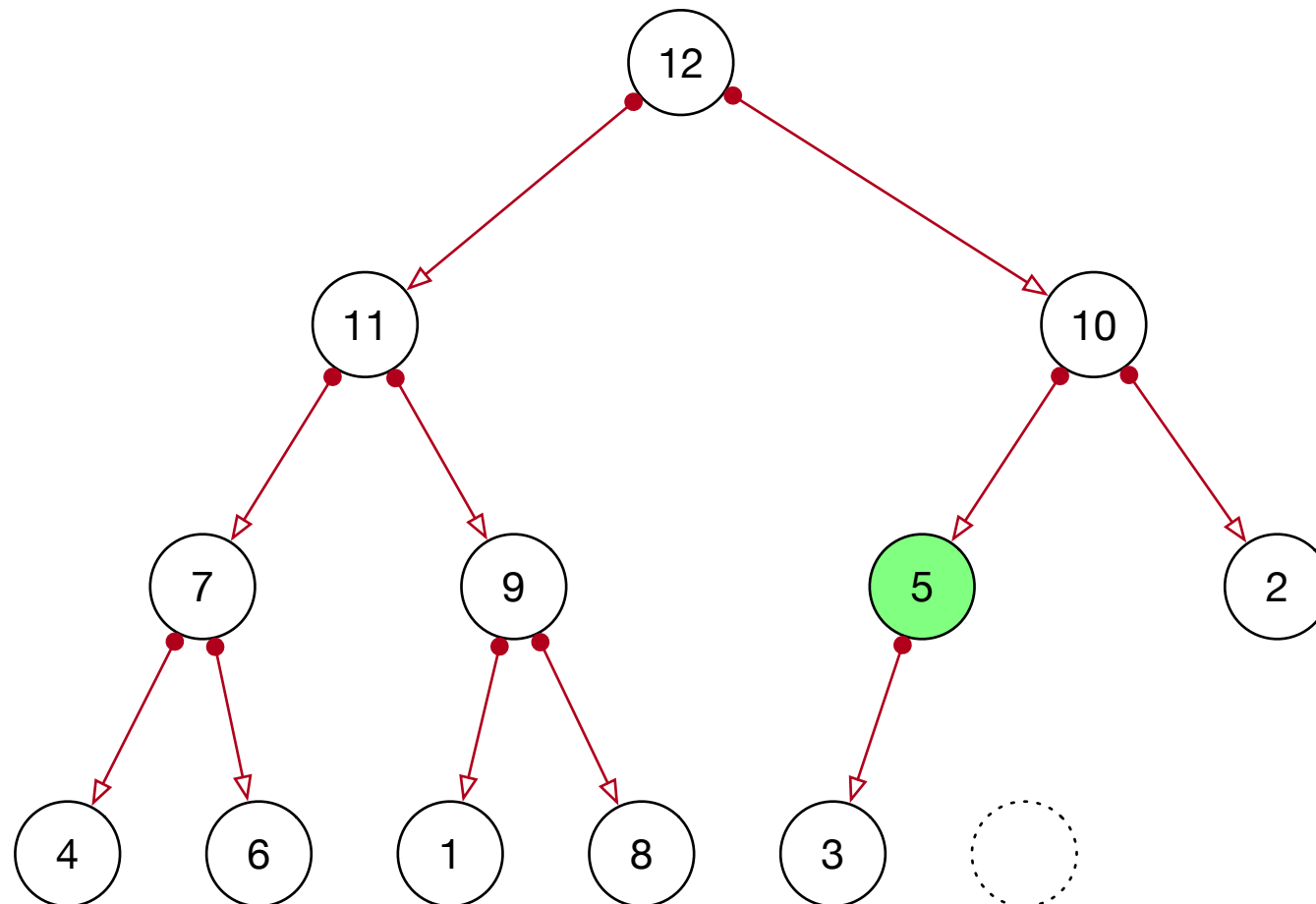
# Priority Queues

- Then restore the heap property
  - Move up the *larger* sibling



# Priority Queues

- Until there is no violation





# Priority Queues

- Implementation:
  - Need to implement two "heapify" operations
    - Going up for insert
    - Going down for extract maximum

# Priority Queues

- Define a class PQ with class methods for index calculation

```
class PQ:
    def __init__(self):
        self.array = []
    def up(index):
        return (index+1)//2-1
    def left(index):
        return 2*index + 1
    def right(index):
        return 2*index + 2
```

# Priority Queues

- Insert at the end of the array
- but note the index

```
def insert(self, value):  
    n = len(self.array)  
    self.array.append(value)  
    while n>0:  
        parent = PQ.up(n)  
        print(n, parent, 'indices')  
        if self.array[parent] < value:  
            self.array[n], self.array[parent] =  
                self.array[parent], self.array[n]  
            n = parent  
        else:  
            return
```

# Priority Queues

- Adjust by swapping with parent
  - Index of current element is  $n$

```
def insert(self, value):
    n = len(self.array)
    self.array.append(value)
    while n>0:
        parent = PQ.up(n)
        print(n, parent, 'indices')
        if self.array[parent] < value:
            self.array[n], self.array[parent] =
                self.array[parent], self.array[n]
            n = parent
        else:
            return
```

# Priority Queues

- Calculate the parent node

```
def insert(self, value):
    n = len(self.array)
    self.array.append(value)
    while n > 0:
        parent = PQ.up(n)

        if self.array[parent] < value:
            self.array[n], self.array[parent] =
                self.array[parent], self.array[n]
            n = parent
        else:
            return
```

# Priority Queues

- And swap if necessary

```
def insert(self, value):
    n = len(self.array)
    self.array.append(value)
    while n > 0:
        parent = PQ.up(n)

        if self.array[parent] < value:
            self.array[n], self.array[parent] =
                self.array[parent], self.array[n]
            n = parent
    else:
        return
```

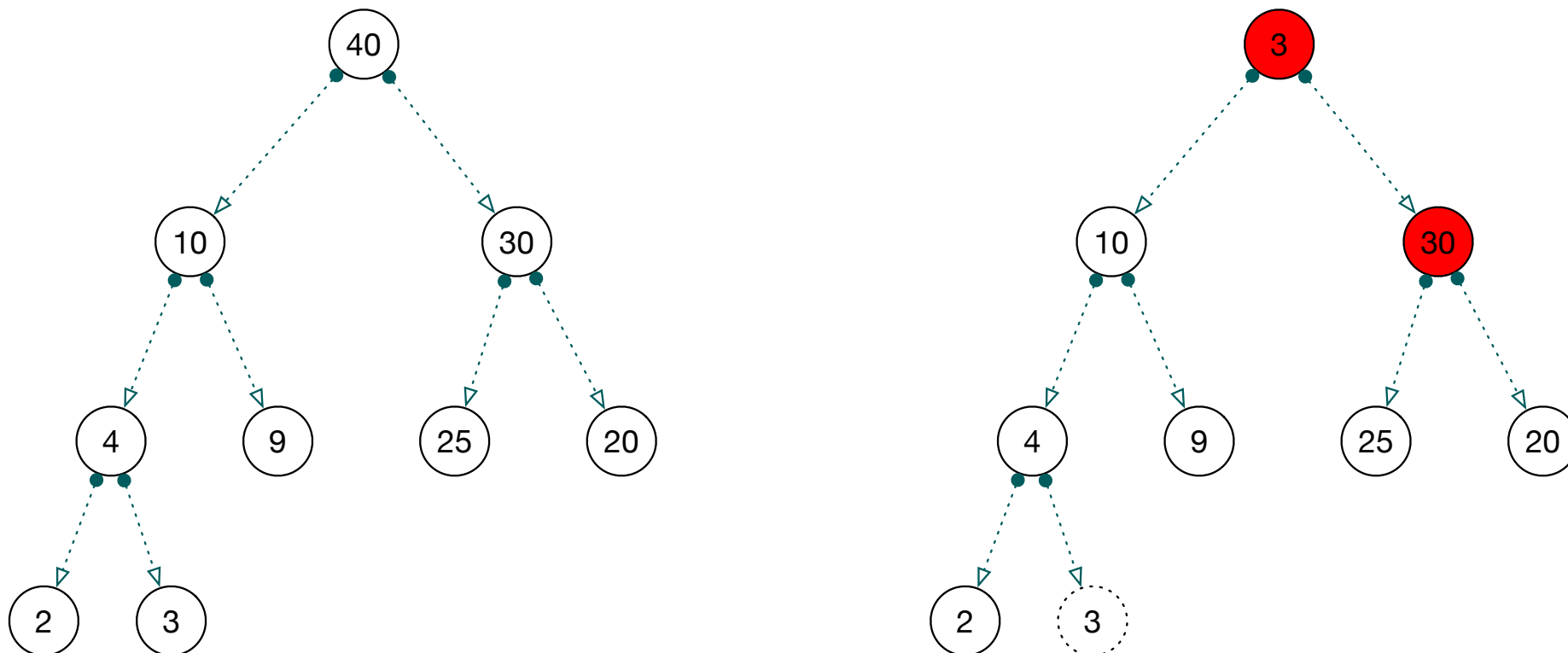
# Priority Queues

- Then reset the index

```
def insert(self, value):
    n = len(self.array)
    self.array.append(value)
    while n > 0:
        parent = PQ.up(n)
        print(n, parent, 'indices')
        if self.array[parent] < value:
            self.array[n], self.array[parent] =
                self.array[parent], self.array[n]
            n = parent
        else:
            return
```

# Priority Queues

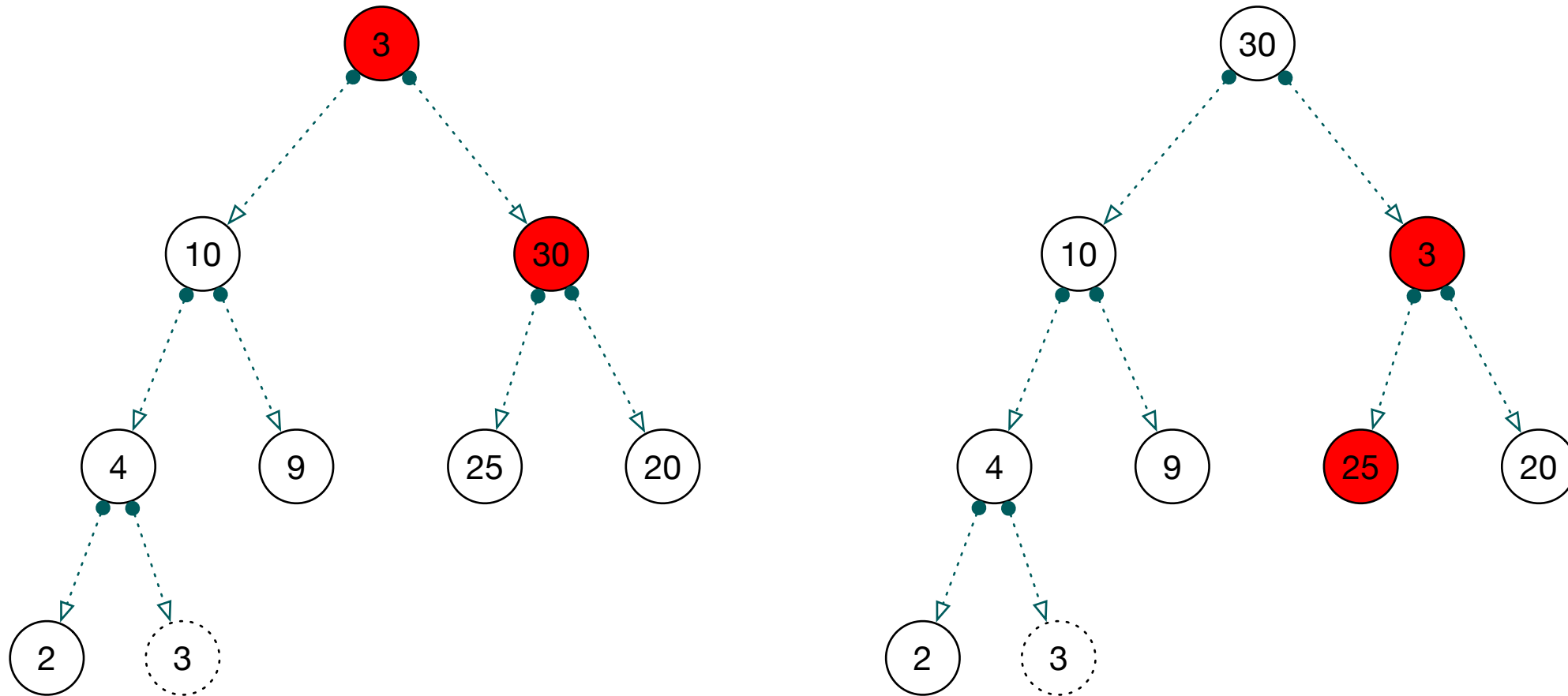
- Extract maximum:
  - Maximum is always at position 0
  - Swap its value with the last element in the array
  - Then heapify:



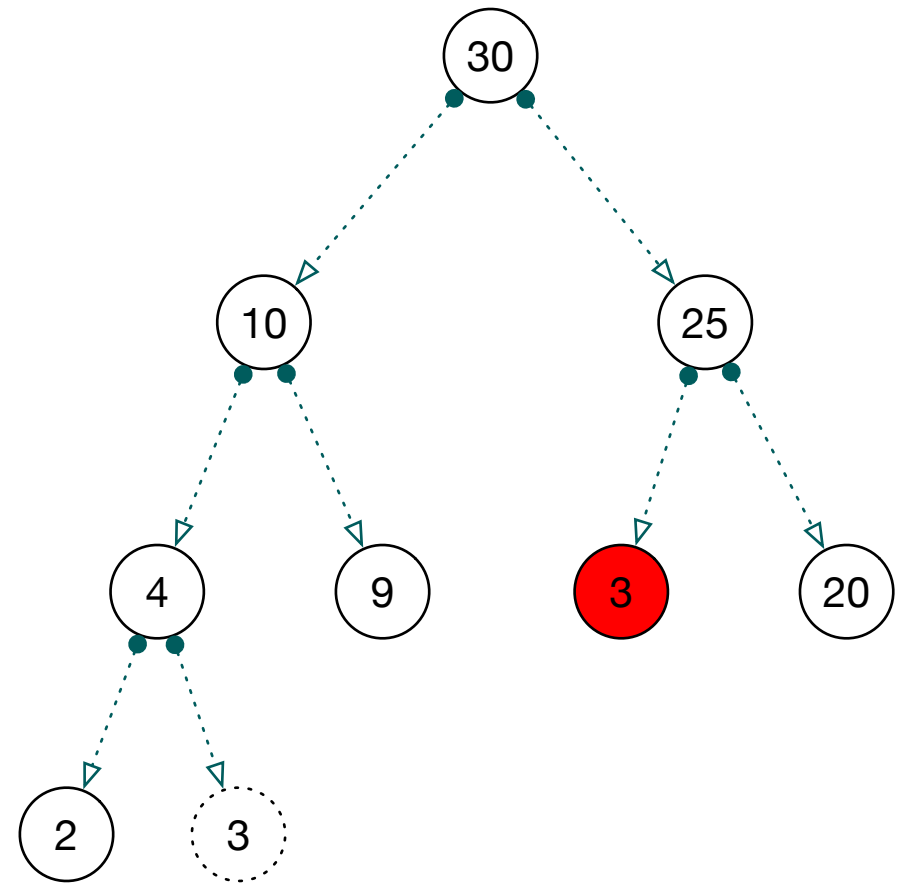
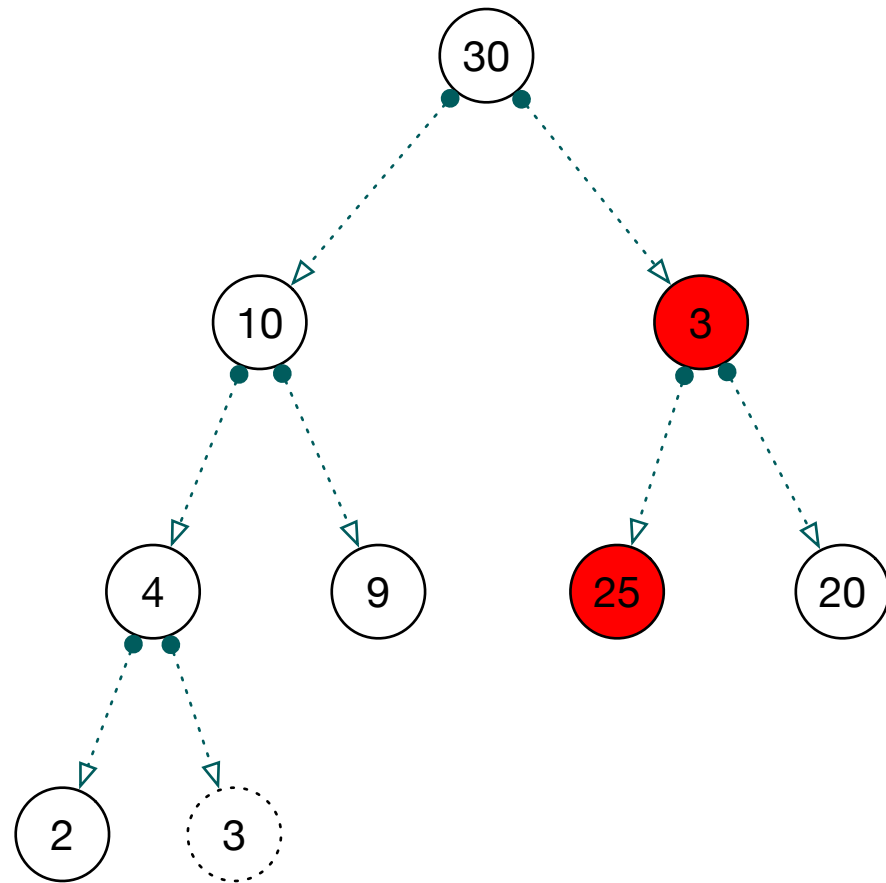


# Priority Queues

- This is also recursive, but proceeds from top to bottom



# Priority Queues



# Priority Queues

- Swap last and first node
- Delete from node

```
def get_max(self):  
    ret_val = self.array[0]  
    last = self.array[-1]  
    del self.array[-1]  
    self.array[0] = last  
    n=0
```

# Priority Queues

- Now recursively recover the heap property
  - Make case distinctions according to whether
    - both children exist
    - only the left child exist
    - no children present

# Priority Queues

- Both children exist

```
def get_max(self):
    ...
    while n < len(self.array):
        left = PQ.left(n)
        right = PQ.right(n)
        if right < len(self.array):
            if self.array[n] > self.array[left] and
                self.array[n] > self.array[right]:
                return ret_val
            if self.array[left] < self.array[right]:
                m = right
            else:
                m = left
            self.array[n], self.array[m] = self.array[m],
self.array[n]
            n = m
```

# Priority Queues

- Heap property is not violated

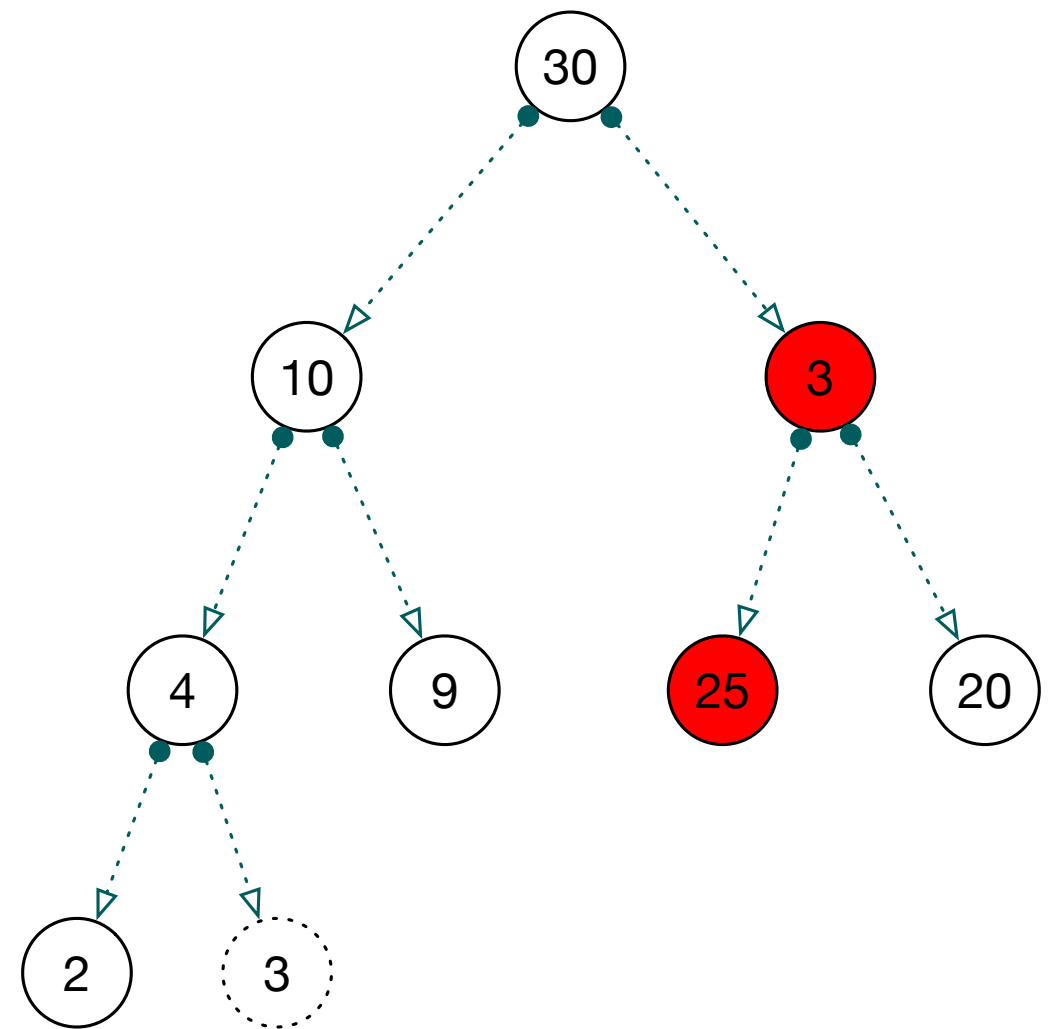
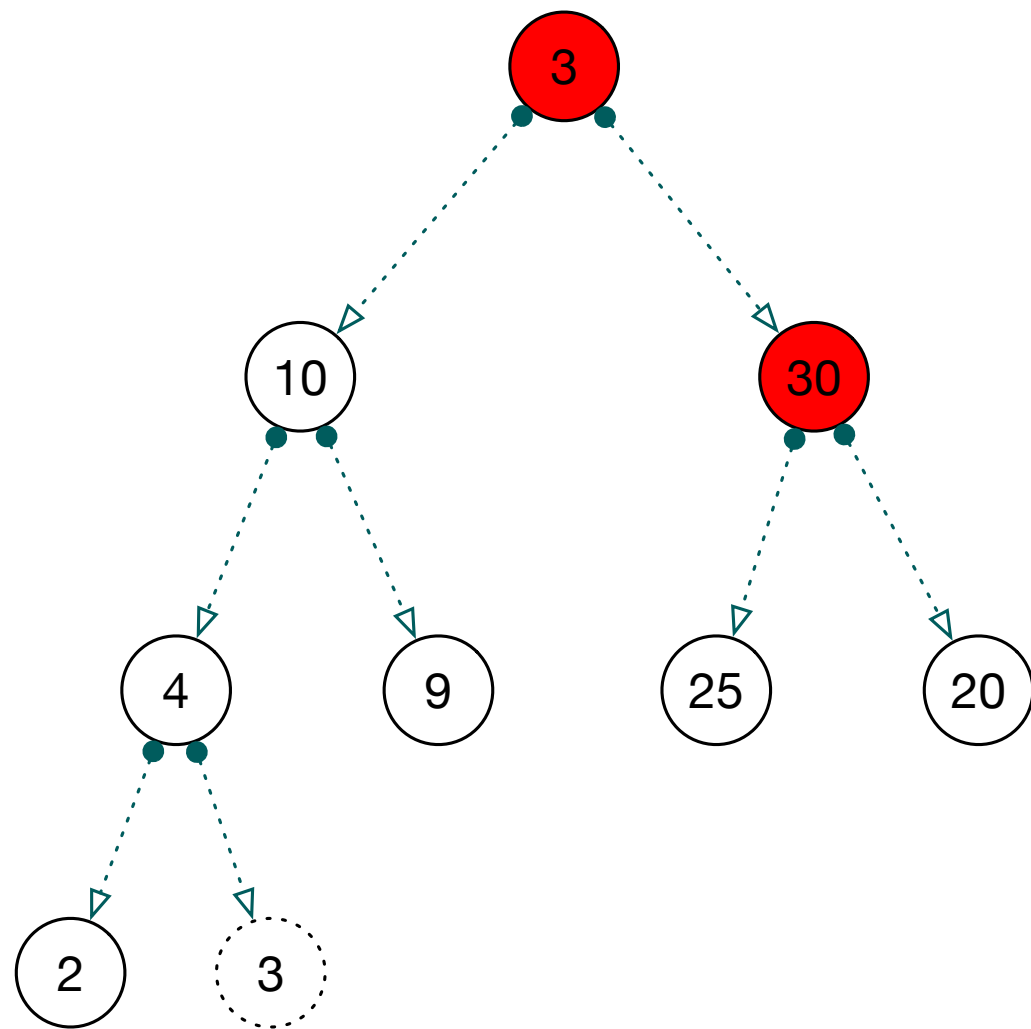
```
def get_max(self):
    ...
    while n < len(self.array):
        left = PQ.left(n)
        right = PQ.right(n)
        if right < len(self.array):
            if self.array[n] > self.array[left] and
                self.array[n] > self.array[right]:
                return ret_val
            if self.array[left] < self.array[right]:
                m = right
            else:
                m = left
            self.array[n], self.array[m] = self.array[m],
self.array[n]
            n = m
```

# Priority Queues

- Select the larger of the two children for swapping

```
def get_max(self):
    ...
    while n < len(self.array):
        left = PQ.left(n)
        right = PQ.right(n)
        if right < len(self.array):
            if self.array[n] > self.array[left] and
                self.array[n] > self.array[right]:
                return ret_val
            if self.array[left] < self.array[right]:
                m = right
            else:
                m = left
            self.array[n], self.array[m] =
                self.array[m], self.array[n]
        n = m
```

# Priority Queues





# Priority Queues

- Swap

```
def get_max(self):
    ...
    while n < len(self.array):
        left = PQ.left(n)
        right = PQ.right(n)
        if right < len(self.array):
            if self.array[n] > self.array[left] and
                self.array[n] > self.array[right]:
                return ret_val
            if self.array[left] < self.array[right]:
                m = right
            else:
                m = left
            self.array[n], self.array[m] =
                self.array[m], self.array[n]
        n = m
```

# Priority Queues

- Swap

```
def get_max(self):
    ...
    while n < len(self.array):
        left = PQ.left(n)
        right = PQ.right(n)
        if right < len(self.array):
            if self.array[n] > self.array[left] and
                self.array[n] > self.array[right]:
                return ret_val
            if self.array[left] < self.array[right]:
                m = right
            else:
                m = left
            self.array[n], self.array[m] =
                self.array[m], self.array[n]
        n = m
```

# Priority Queues

- And do not forget to set yourself up for recursion

```
def get_max(self):
    ...
    while n < len(self.array):
        left = PQ.left(n)
        right = PQ.right(n)
        if right < len(self.array):
            if self.array[n] > self.array[left] and
                self.array[n] > self.array[right]:
                return ret_val
            if self.array[left] < self.array[right]:
                m = right
            else:
                m = left
            self.array[n], self.array[m] =
                self.array[m], self.array[n]
        n = m
```

# Priority Queues

- Only one child can exist (but then it has to be the left one)
- Heap property might not be violated

```
elif left < len(self.array):  
    if self.array[n] > self.array[left]:  
        return ret_val  
    m = left  
    self.array[n], self.array[m] =  
        self.array[m], self.array[n]  
    n = m
```

# Priority Queues

- Only one child can exist (but then it has to be the left one)
  - But if it is, we have only one candidate for swapping

```
elif left < len(self.array):  
    if self.array[n] > self.array[left]:  
        return ret_val  
m = left  
self.array[n], self.array[m] =  
        self.array[m], self.array[n]  
n = m
```

# Priority Queues

- Per defensive programming, we pretend that we might have to go on:

```
elif left < len(self.array):  
    if self.array[n] > self.array[left]:  
        return ret_val  
    m = left  
    self.array[n], self.array[m] =  
        self.array[m], self.array[n]  
n = m
```

# Priority Queues

- Difficult Homework:
  - Extract Maximum and insertion of a new element are sometimes combined
  - In this case, we can save work by:
    - inserting the new element at the beginning of the array
    - work ourselves downwards to restore the heap property
  - Implement this

# Priority Queues

- Other operations:
  - peek
    - returns the maximum, but does not remove it
  - is\_empty
    - checks whether the array is empty



# Priority Queues

- Costs of operations
  - Priority queue with  $n$  elements uses  $\log_2(n)$  steps in order to heapify
  - Peek and is\_empty run in constant time

# Priority Queues

- Python implementation of priority queues
  - heapq implements a minimum heap
  - Uses a Python list

```
heapq.heappush(lista, element)
```

```
heapq.heappop(lista)
```

# Priority Queues

- This is an efficient implementation
  - We can "kludge" a max heap implementation for integers by observing that the maximum of numbers is the negative of the negative integers

```
def smallpush(lista, element):  
    heapq.heappush(lista, -element)  
def smallpop(lista):  
    return -heapq.heappop(lista)
```

# Running Medians

- Task:
  - We are given a stream of numbers
    - At any time, want to be able to determine the median of these numbers
- Example:
  - We get 5, 3, 1, 10, 2
  - Median is now 3
  - We then get 12, 1, 2
    - We have seen 1,1,2,2,3,5,10,12
  - Median is now 2.5 (mean of 2 and 3)

# Running Medians

- Naïve implementation
  - Just keep an ordered list around
- Better way:
  - Keep two sublists of equal size
    - Small and Big
    - All elements in Small are smaller than all elements in Big
    - Use heaps in order to easily extract the maximum of Small and the minimum of Big

# Running Medians

- Adding a new number:
  - If the left heap is smaller, then insert there
  - If the left and right heap have equal size, insert in the right heap
  - But need to maintain the invariant:
    - All elements in the left heap are smaller (or equal) than all elements in the right heap

# Running Medians

- Example: Inserting 5 into
  - Left: 0, 1, 1, 2, 2      Right: 3, 4, 6, 7, 7, 9
- We need to insert into Left, but this violates the invariant
  - Extract the minimum from right (3)
  - Add the minimum to the left
  - Add 5 to right
  - Left: 0, 1, 1, 2, 2, 3      Right: 4, 5, 6, 7, 7, 9

# Running Medians

- Insert another 5:
  - Left: 0, 1, 1, 2, 2, 3      Right: 4, 5, 6, 7, 7, 9
- Rule say insert to the Right:
  - Since  $\max(\text{left}) < 5$ :
    - No problem:
  - Left: 0, 1, 1, 2, 2, 3      Right: 4, 5, 5, 6, 7, 7, 9



# Running Medians

- Insert another 5:
  - Insert into Left:
    - But  $\min(\text{right}) = 4$  which is smaller than 5
  - Inserting 5 into left violates the invariant
    - Need to do something about it:
      - Extract minimum from Right
      - Insert this minimum into Left
      - Insert new element into Right
- Left: 0, 1, 1, 2, 2, 3, 4      Right: 5, 5, 6, 7, 7, 9

# Running Medians

- Calculating medians:
  - If  $\text{len}(\text{Left}) < \text{len}(\text{Right})$ :
    - Median is  $\text{peek}(\text{Right})$
  - Otherwise:
    - Median is  $(\text{peek}(\text{Right}) + \text{peek}(\text{Left})) / 2$

# Tiered Bitvectors

- Given a finite universe indexed by  $\{0, 1, \dots, n - 1\}$ :
  - $U = \{x_i : i \in \{0, 1, \dots, n - 1\}\}$
- A bit vector represents subsets of  $U$ 
  - If  $\delta_{i,S} = \begin{cases} 0 & \text{if } i \in S \\ 1 & \text{if } i \notin S \end{cases}$  (Kronecker delta)
  - $S \subset U$  corresponds to the bit-vector
    - $(\delta_{i,S} : i \in \{0, 1, \dots, n - 1\})$

# Tiered Bitvectors

- To insert  $x_i$  into the set:
  - Set bit  $i$  to 1
- To delete  $x_i$  from the set:
  - Set bit  $i$  to 0
- To lookup whether  $x_i \in S$ :
  - Check value of bit  $i$

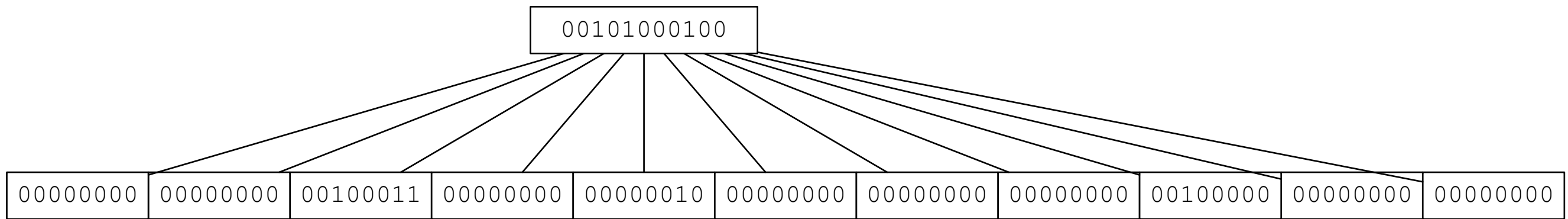
# Tiered Bitvectors

- MinIndex, MaxIndex, PredecessorIndex, SuccessorIndex are extremely slow
  - $\Theta(|U|)$

# Tiered Bitvectors

- If memory is (as is typical) accessed via cache-lines
  - Words of length  $L = 64\text{B}$
- Break the universe into  $|U|/L$  pieces of size  $L$ 
  - $U_0, U_1, U_2, \dots, U_{n/L}$
- Introduce a master bitvector by
  - $\Delta_{i,S} = \begin{cases} 0 & \text{if } S \cap U_i = \emptyset \\ 1 & \text{if } S \cap U_i \neq \emptyset \end{cases}$

# Tiered Bitvectors



# Tiered Bitvectors

- Ordered dictionary operations run in time
  - $O(U/L + L)$
- This is minimized when  $L = \sqrt{U}$



# Tiered Bitvectors

- What are the operations?

# Tiered Bitvectors

- Inserting to a set:
  - Set master bitvector bit
  - Set partial bitvector bit
- Deleting from a set
  - Reset partial bitvector bit
  - If partial bitvector is empty: Reset master bitvector bit

# Tiered Bitvectors

- is-empty:
  - Check master bitvector only has entries 0
- min:
  - One min operation on the master bitvector, one min operation on partial bitvector

# Tiered Bitvectors

- Obviously, we can extend this from two tiers to many tiers.
  - Result is a tree
  -

# Fibonacci Heaps

# Mergeable Heaps

- Mergeable Heaps are ADS with
  - Make-Heap
  - $\text{Insert}(H, x)$
  - $\text{Minimum}(H)$
  - $\text{Extract-Min}(H)$
  - $\text{Union}(H_1, H_2)$
- Fibonacci heaps in addition have
  - $\text{Decrease-Key}(H, x, \text{new\_val})$
  - $\text{Delete}(H, x)$

# Mergeable Heaps

Procedure	Binary Heap (worst-case)	Fibonacci Heap (amortized)
Make-Heap	$\Theta(1)$	$\Theta(1)$
Insert	$\Theta(\log n)$	$\Theta(1)$
Minimum	$\Theta(1)$	$\Theta(1)$
Extract-Min	$\Theta(\log n)$	$O(\log n)$
Union	$\Theta(n)$	$\Theta(1)$
Decrease-key	$\Theta(\log n)$	$\Theta(1)$
Delete	$\Theta(\log n)$	$O(\log n)$

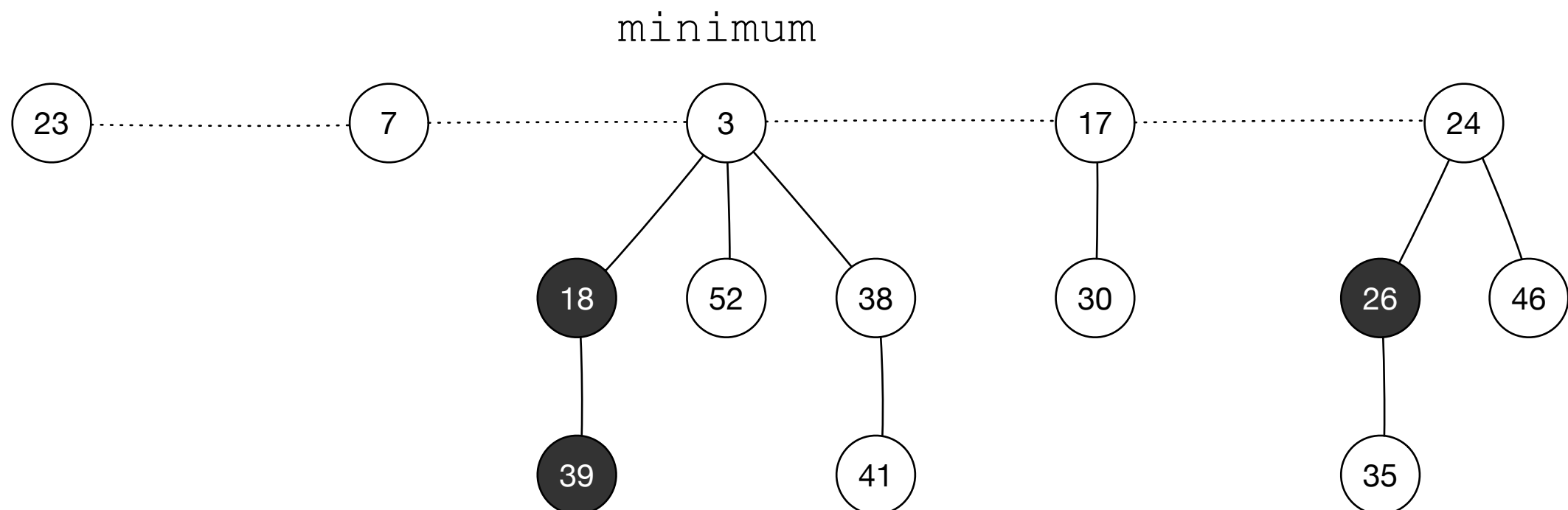
# Fibonacci Heaps

- Fibonacci heaps:
  - Useful when Extract-Min and Delete are rare
  - E.g. graph algorithms where we use decrease-key in order to update edges
    - Minimum spanning trees
    - Single-source shortest paths
-



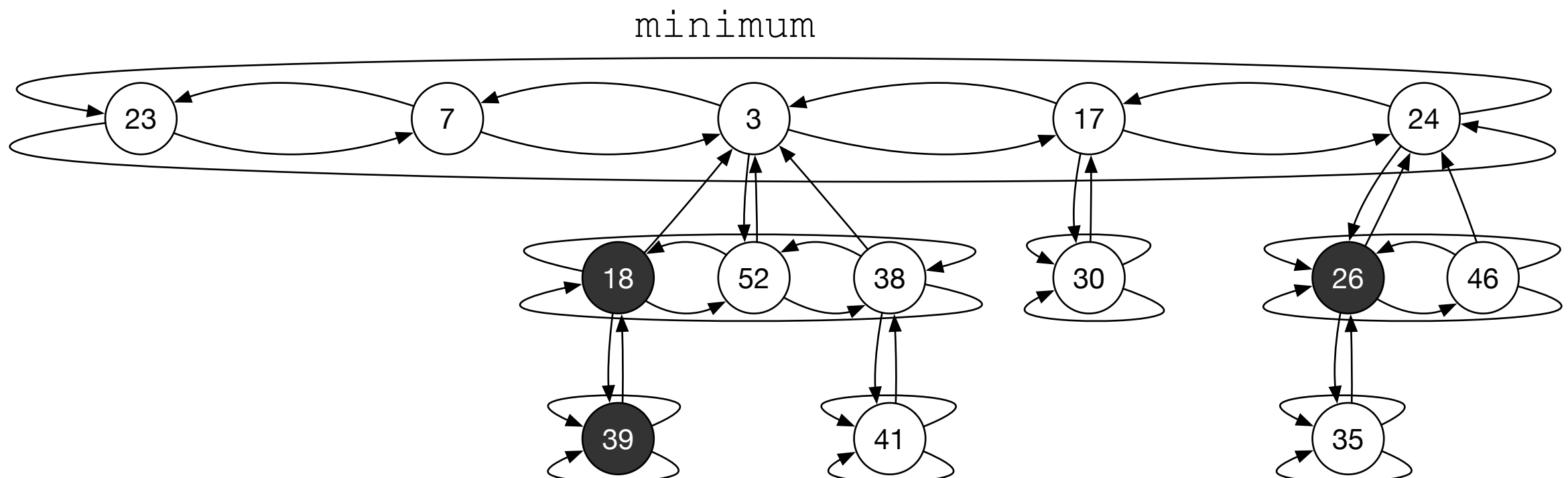
# Fibonacci Heaps

- Fibonacci heap:
  - Collection of rooted trees that are min-heap ordered:
    - Every child's key is larger than its parent's key
- Example:



# Fibonacci Heaps

- Each node has a link to their parent
- All children (including root list) are in a double linked child list
- Parent has one link to the child list



# Fibonacci Heaps

- Double linked list:
  - Allows constant time inserts and fusion

# Fibonacci Heaps

- Each node has attributes
  - Number of children in degree
  - Node  $x$  has mark  $x.mark$  to indicate whether node  $x$  has lost a child since the last time  $x$  was made a child of another node
    - Newly created nodes are unmarked
    - Node  $x$  becomes unmarked whenever it is made child of another node
    - Used for DecreaseKey operation

# Fibonacci Heaps

- Access to a Fibonacci node through pointer to minimum key node
- Each Fibonacci node stores
  - Links to left and right sibling
  - Link to parent unless root
  - Link into the child list

# Fibonacci Heaps

- Use potential method
  - $t(H)$  : number of trees in the root list
  - $m(H)$ : number of marked nodes
- Potential  $\Phi(H) = t(H) + 2m(H)$ 
  - Unit of potential can pay for all constant time operations

# Fibonacci Heap Operations

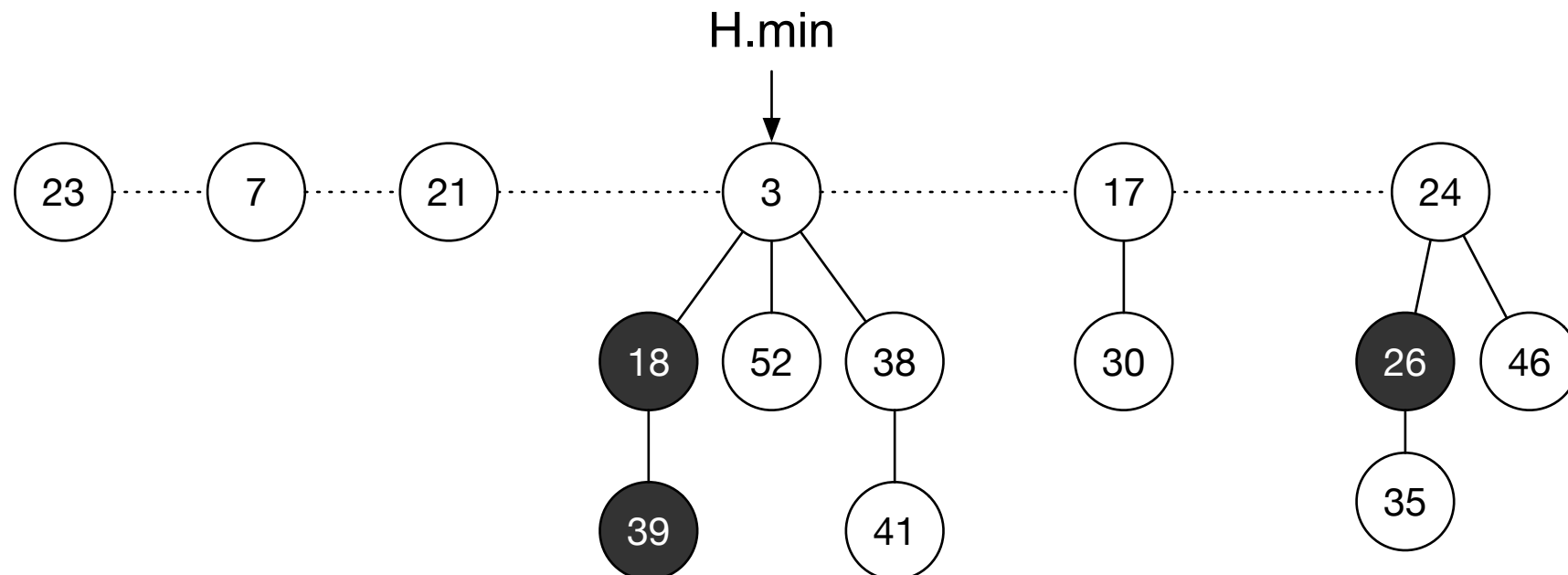
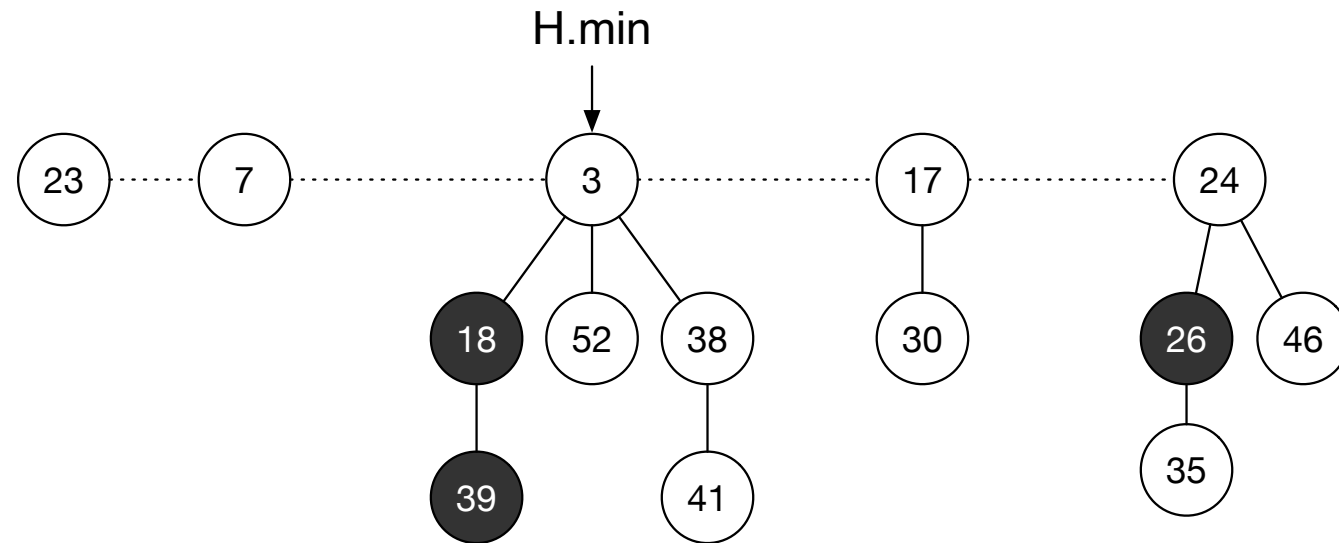
- Creating a new Fibonacci Heap
  - $H.n = 0$
  - $H.min = \text{Null}$
  - $\Phi(H) = 0$
  - amortized cost is  $O(1)$

# Fibonacci Heaps

- Insert( $H, x$ )
  - Create an otherwise empty tree with  $x$  as root
  - If there is no element in the heap ( $H.min = \text{Null}$ ):
    - Create a root list for  $H$  containing  $x$  only
  - Otherwise
    - Insert the new tree into the root list
    - Check whether  $H.min$  needs to be updated
- Amortized costs:  $O(1) + 1 = O(1)$



# Fibonacci Heaps



# Fibonacci Heaps

- Minimum
  - Just returns a pointer to the minimum node
- Amortized costs is  $O(1)$

# Fibonacci Heaps

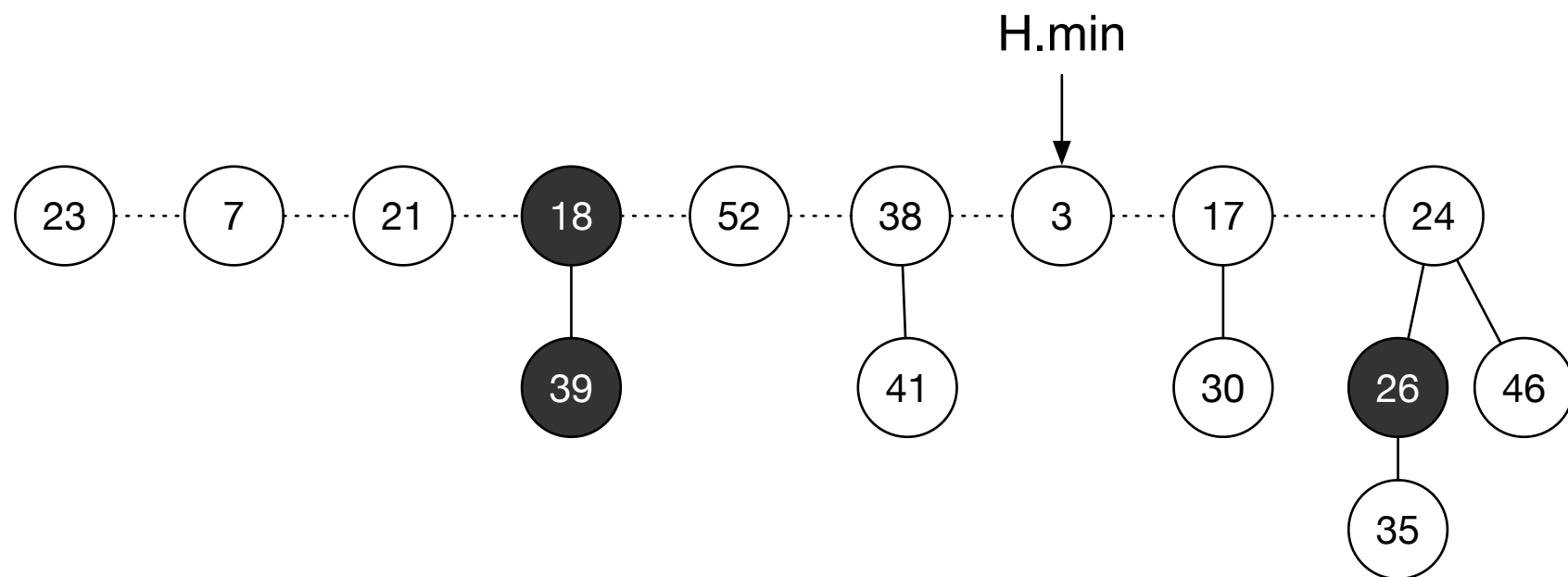
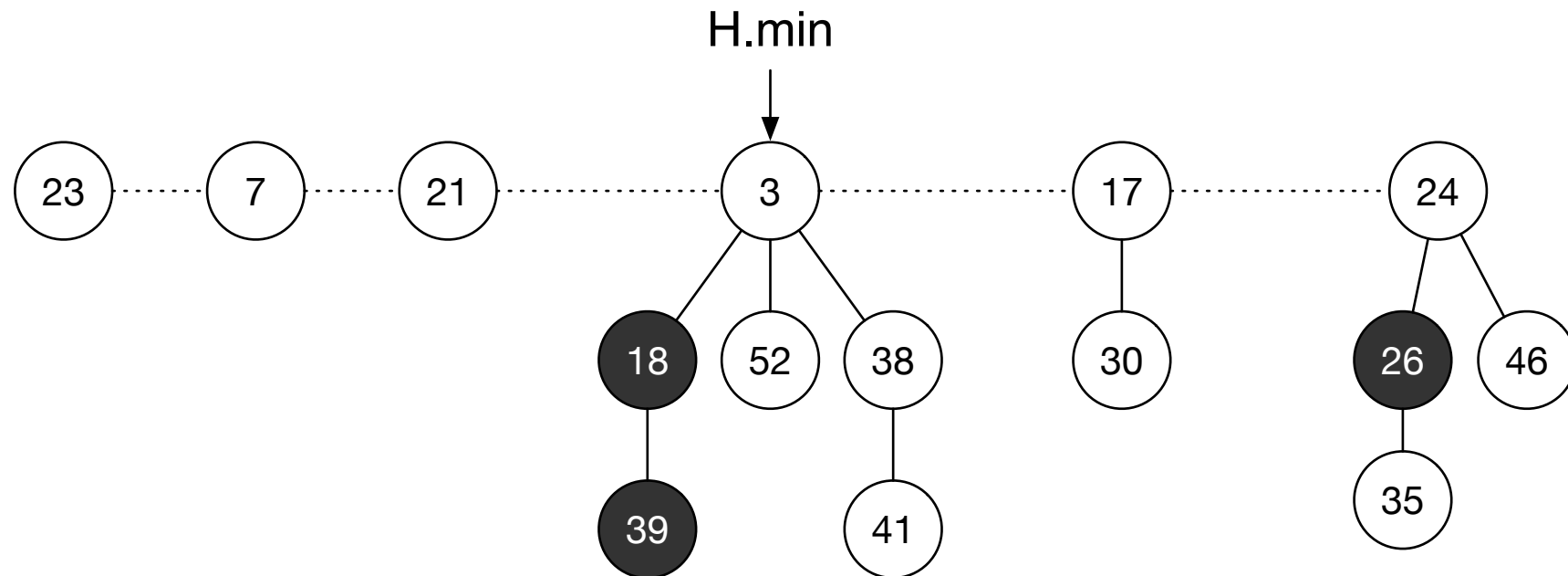
- Uniting two heaps
  - Concatenate the root lists
- Change in potential is zero
- Amortized costs is  $O(1)$

# Fibonacci Heaps

- Extracting the minimum
  - This is where we consolidate

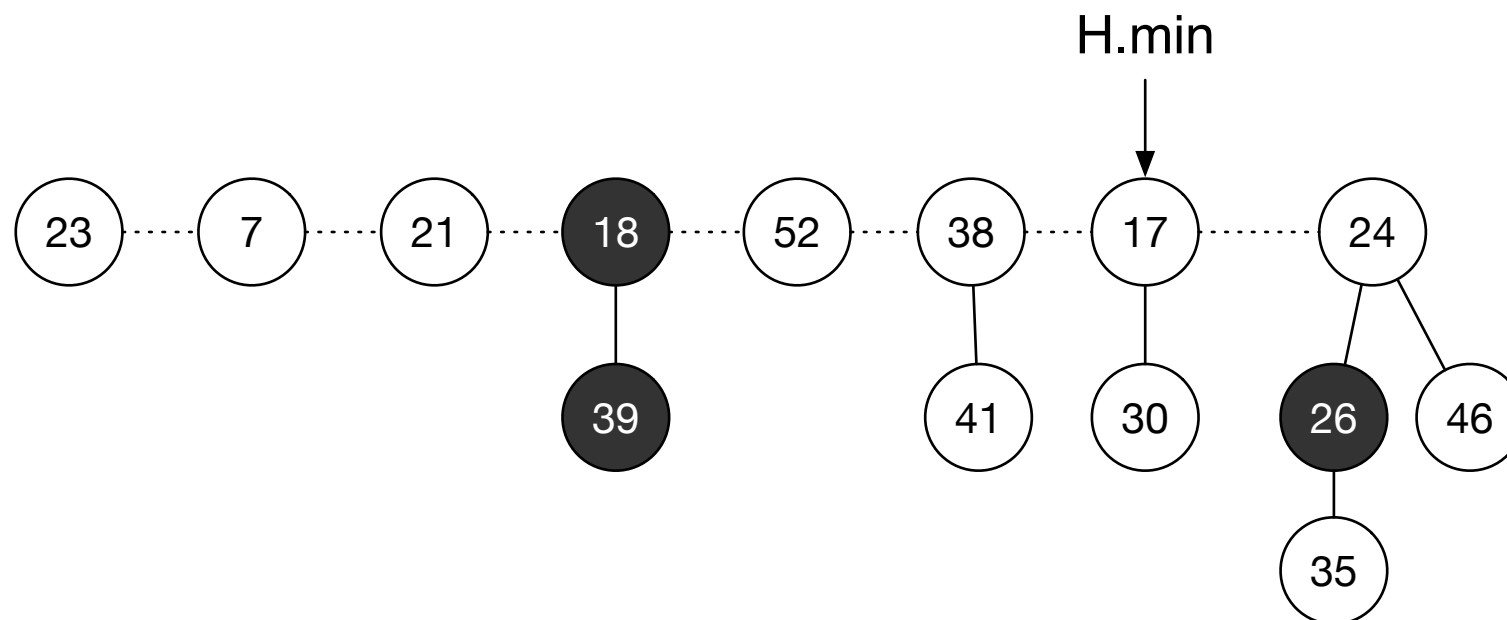
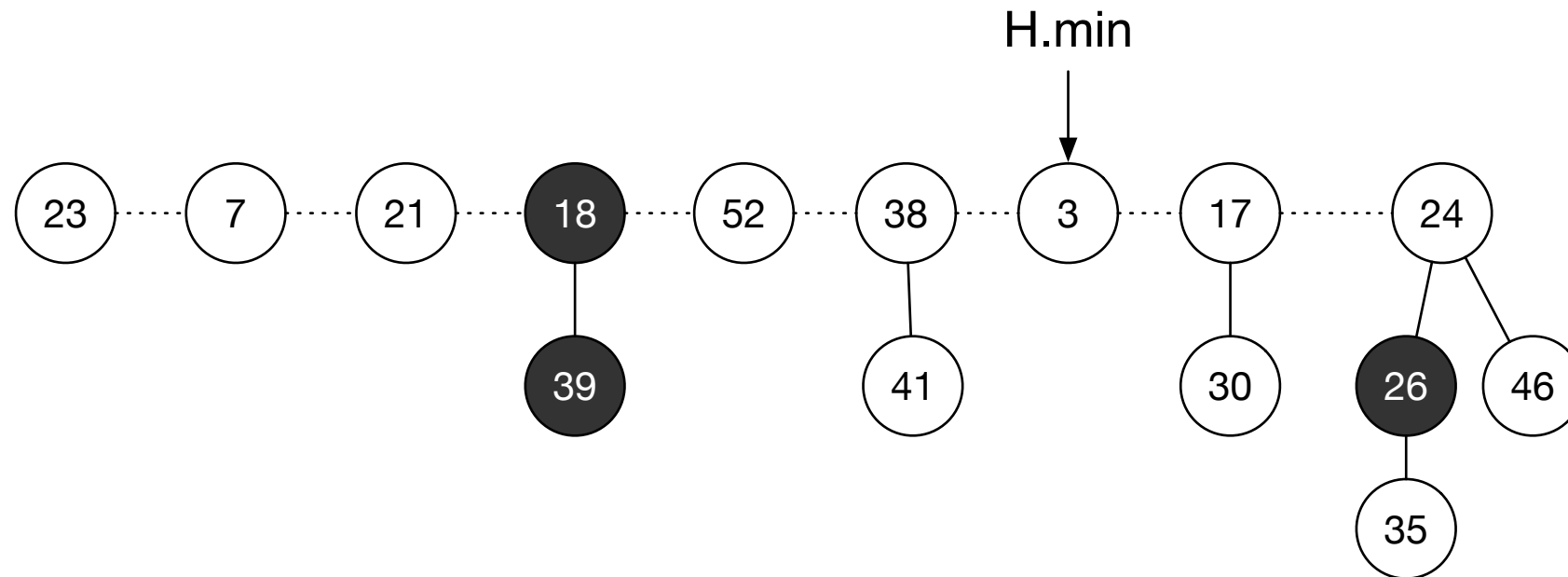
```
z=H.min
if z ≠ NULL:
    for each child x of z:
        add x to the root list of H
        x.p = NULL
    remove z from root list
    if z == z.right: #z only node
        H.min = NULL
    else:
        H.min = z.right
        CONSOLIDATE(H)
    H.n -= 1
return z
```

# Fibonacci Heaps



Moving children of H.min into the root list

# Fibonacci Heaps



Removing and returning H.min

# Fibonacci Heaps

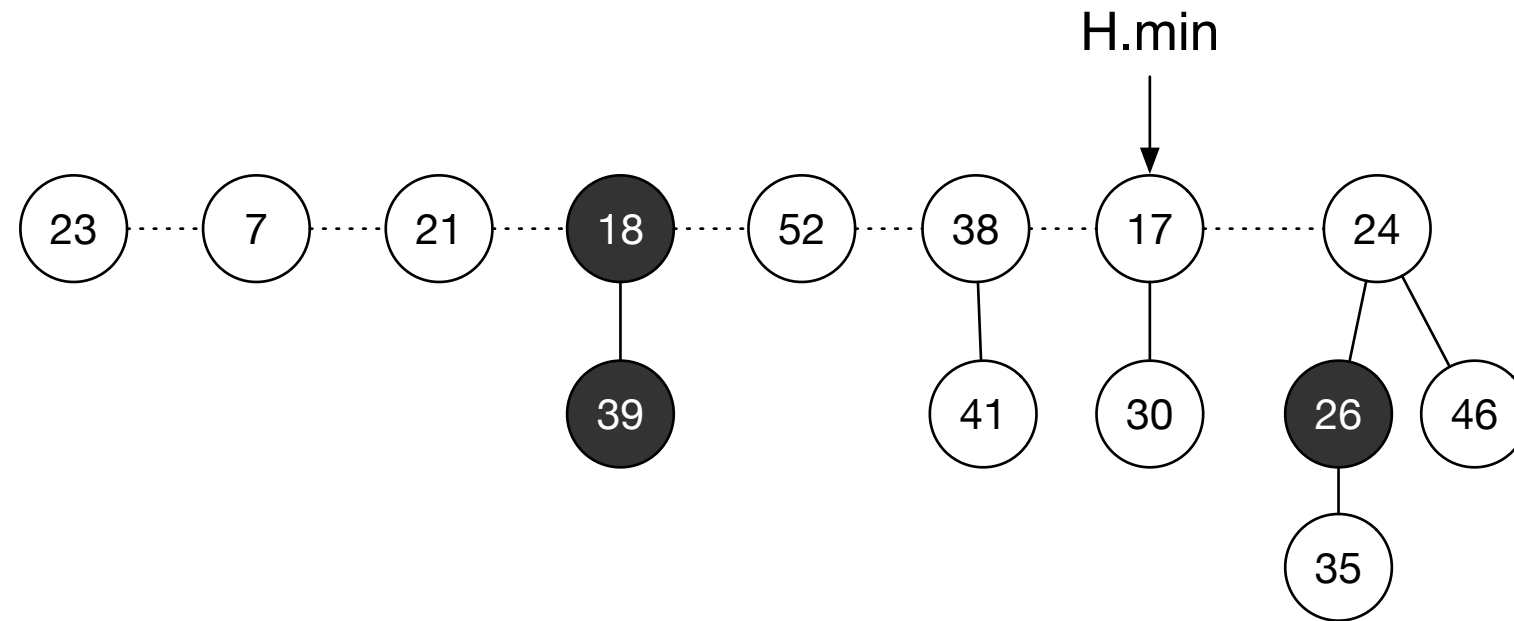
- Consolidation:
  - Repeat:
    - Find two nodes in the root list with trees of the same height
      - Starting at the current link  $H.min$
    - Unify the two trees making the smaller one the root
      - Clears mark on the loser

# Fibonacci Heaps

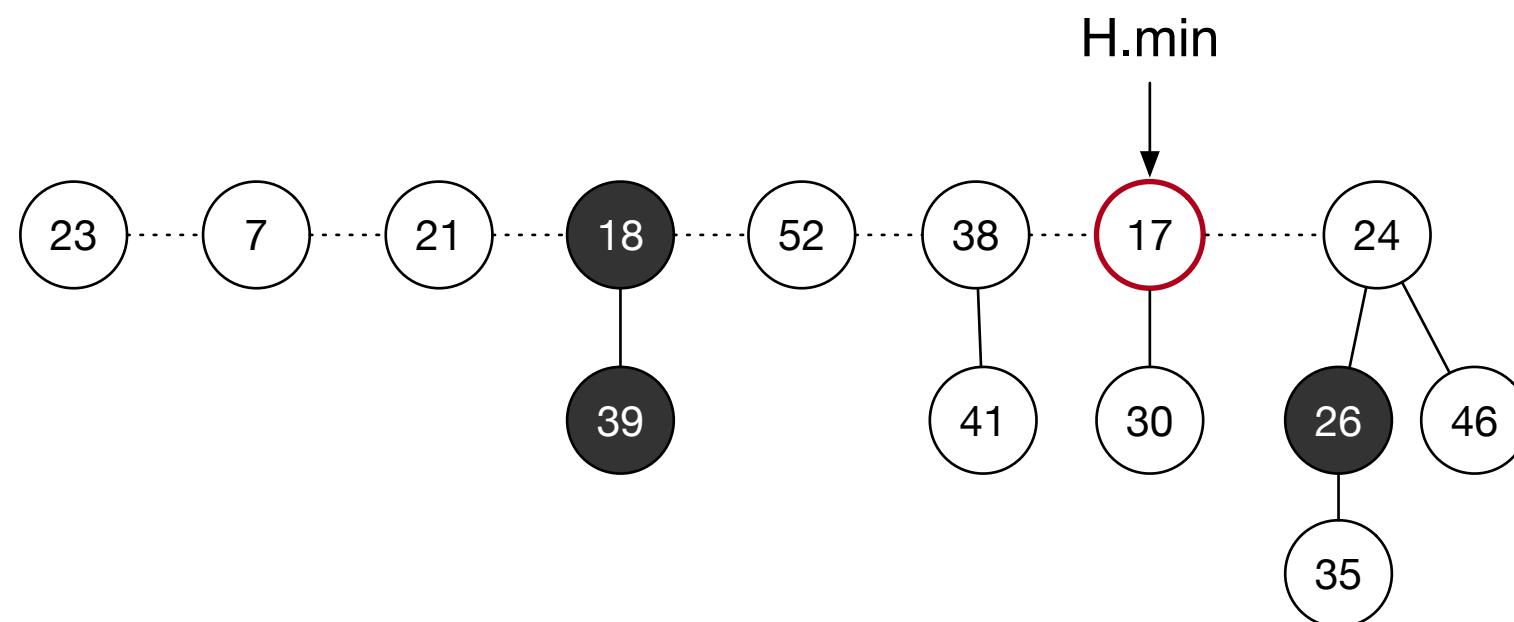
- Consolidation:
  - To find nodes in the root list that can be merged
    - Create an array  $A[0 \dots D(H.n)]$  of nodes in the root tree
      - $D(H.n)$  is the maximum degree of a tree rooted in a node in the root list
    - Fill into  $A$
    -



# Fibonacci Heaps

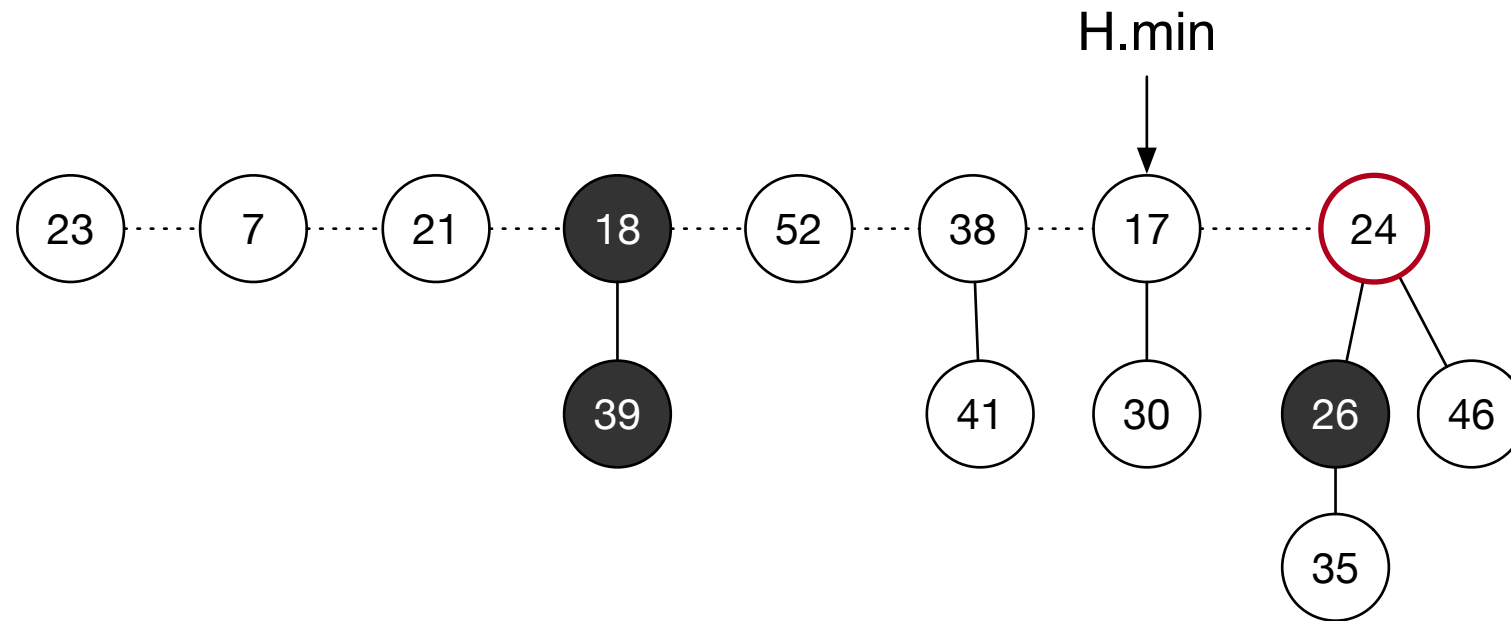


A[0]: NULL  
A[1]: NULL  
A[2]: NULL  
A[3]: NULL



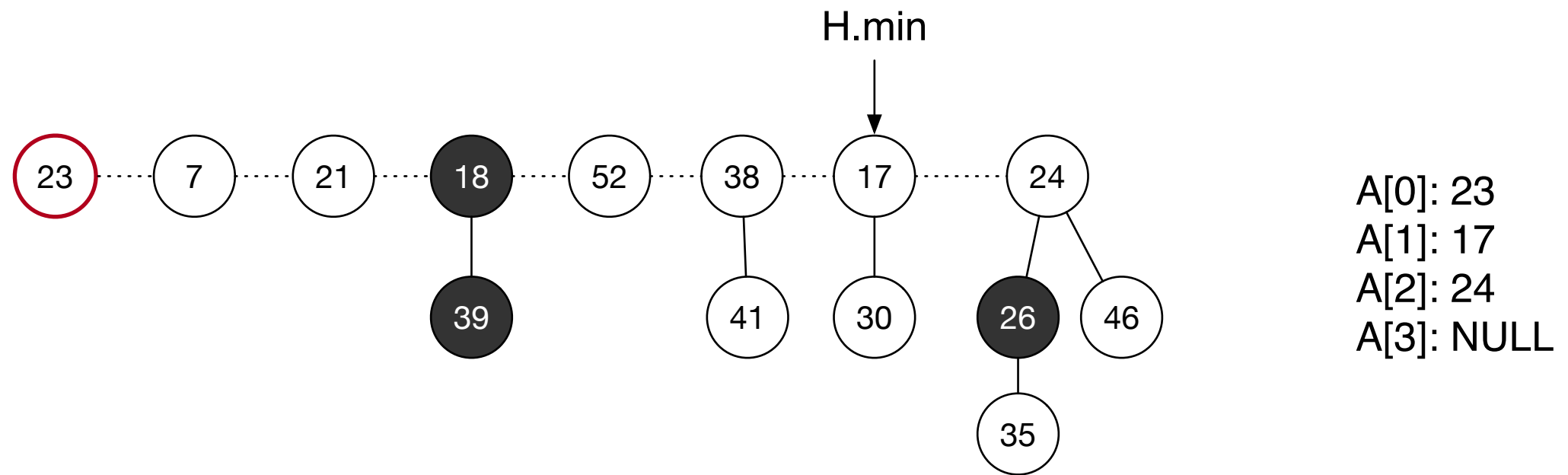
A[0]: NULL  
A[1]: 17  
A[2]: NULL  
A[3]: NULL

# Fibonacci Heaps

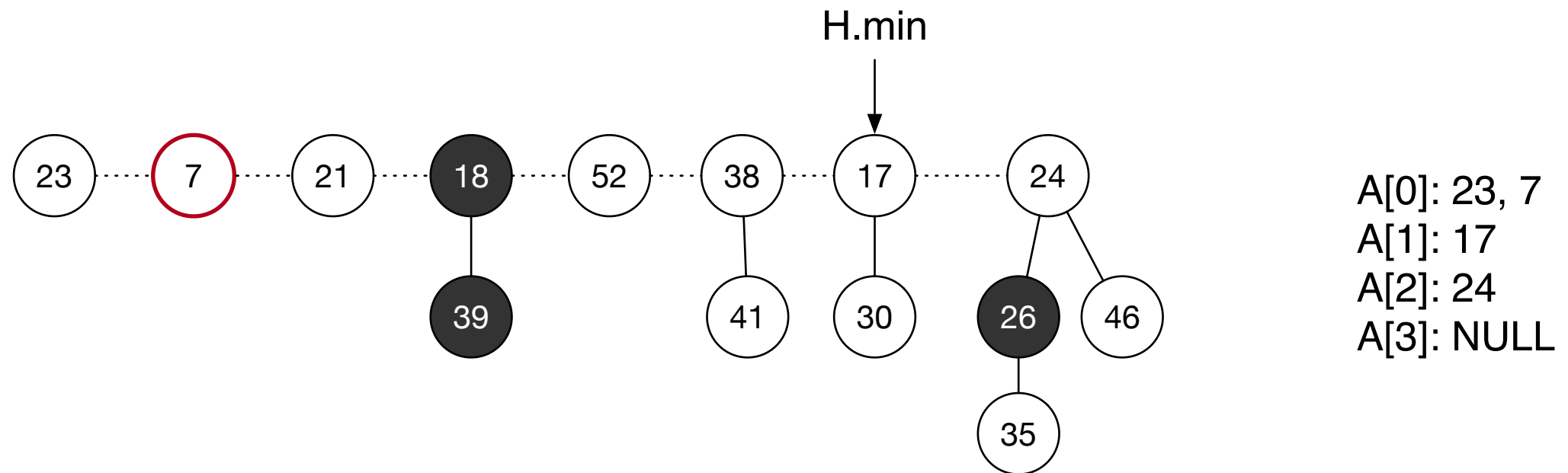


A[0]: NULL  
A[1]: 17  
A[2]: 24  
A[3]: NULL

# Fibonacci Heaps

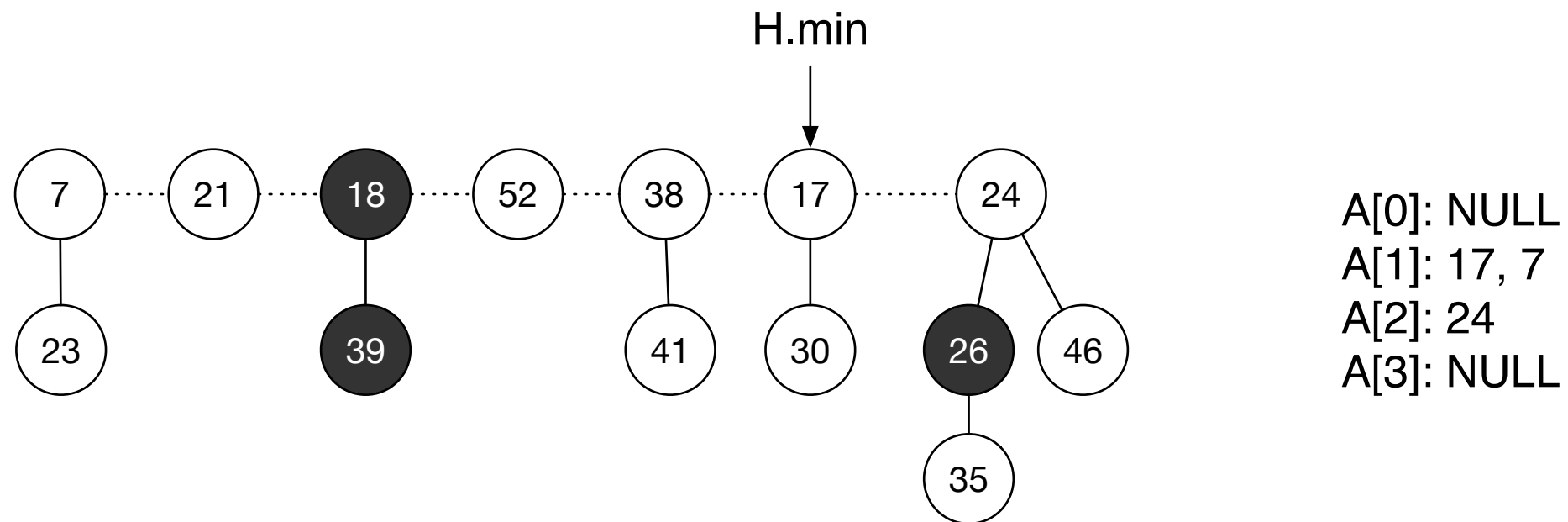


# Fibonacci Heaps



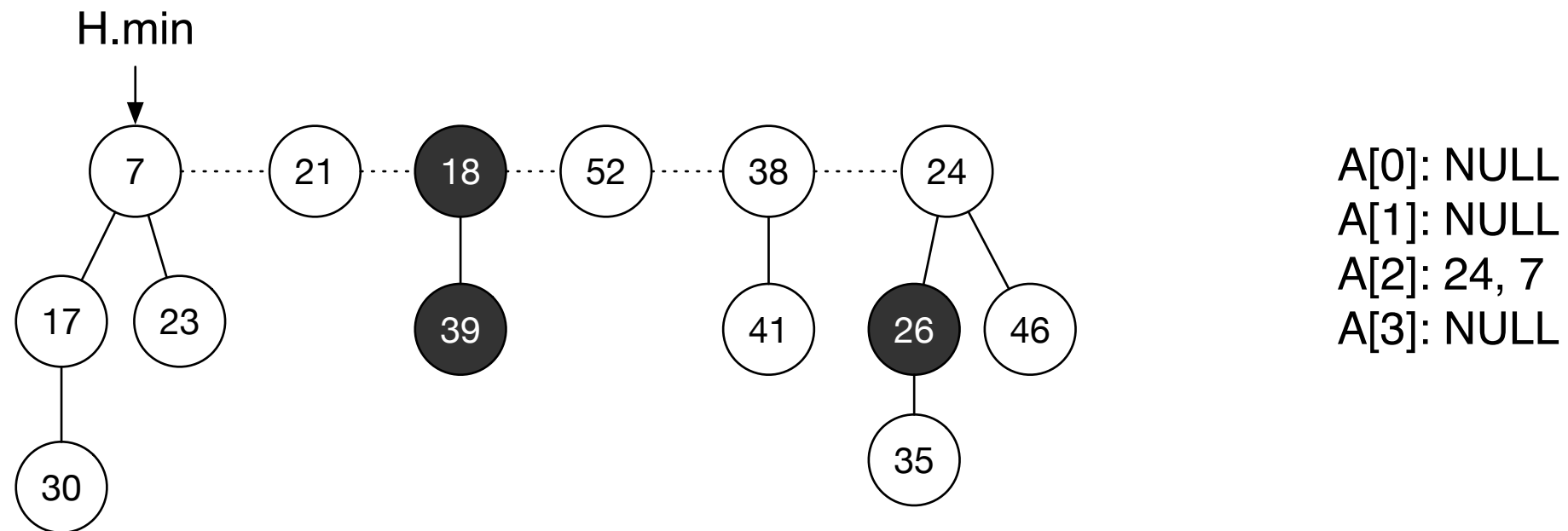
We cannot insert 7 into A[0], so we combine

# Fibonacci Heaps



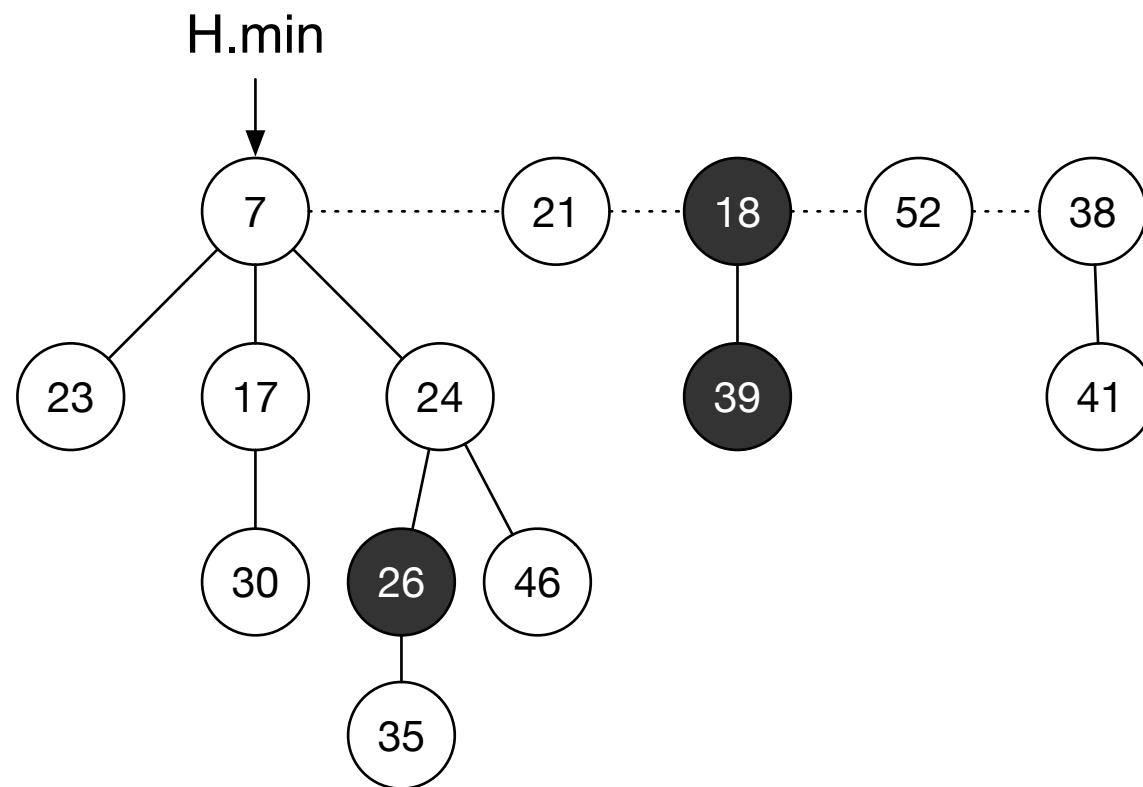
We remove 7 from the array. We then merge. After merging, we try to insert 7 into A[1]. Since there is already an occupant there, we find another merge candidate

# Fibonacci Heaps



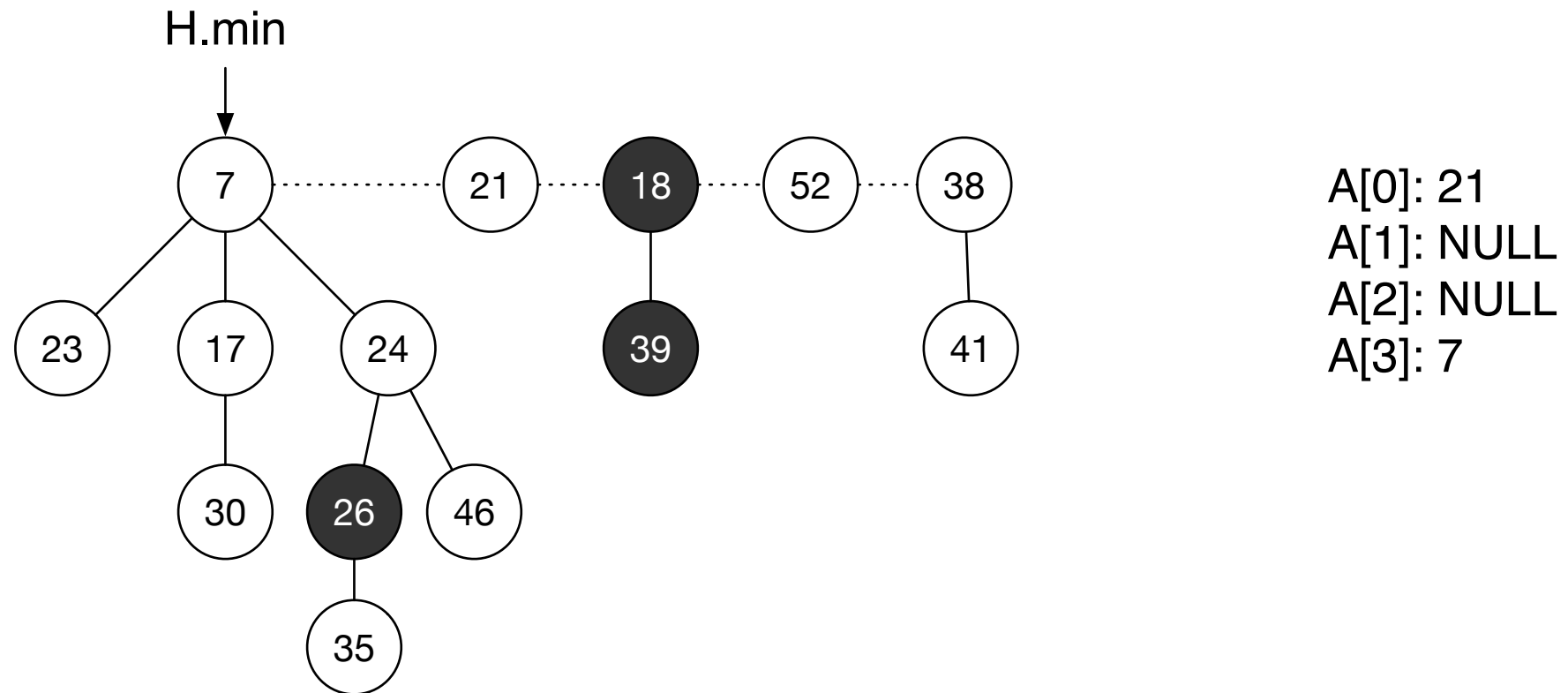
Inserting the new tree into A gives us another merge candidate

# Fibonacci Heaps



A[0]: NULL  
A[1]: NULL  
A[2]: NULL  
A[3]: 7

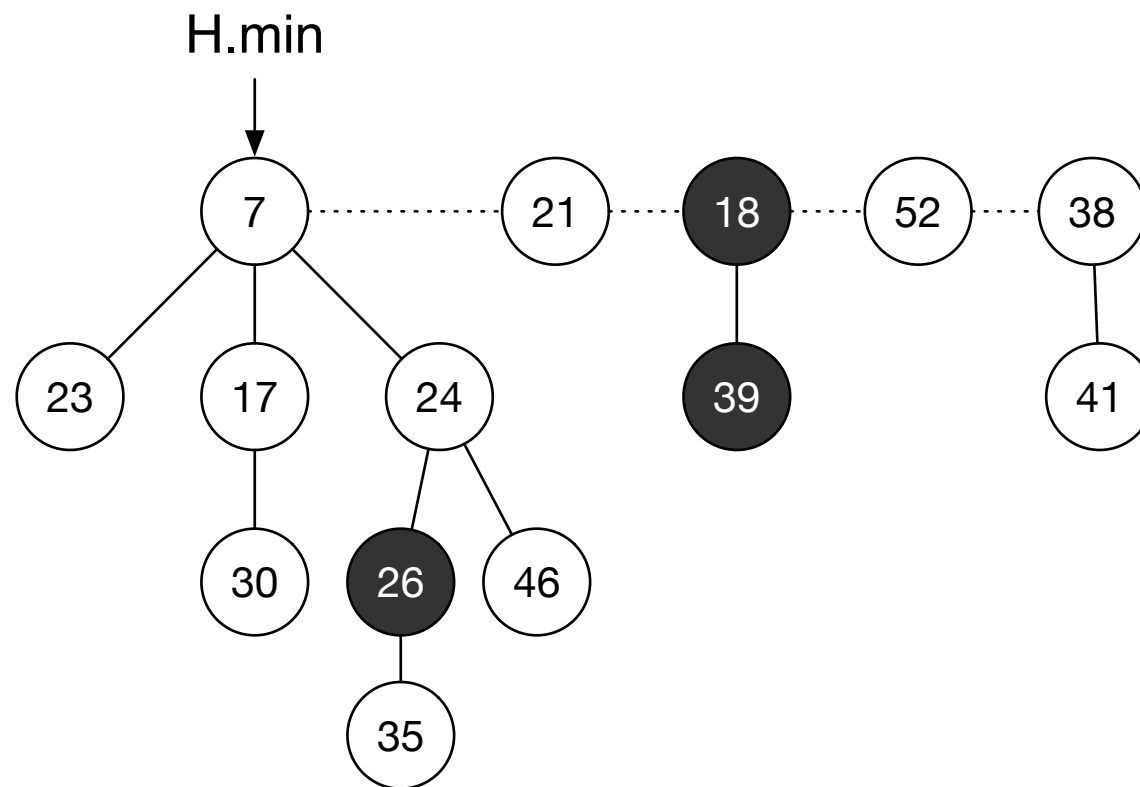
# Fibonacci Heaps



We insert 21 into the array.



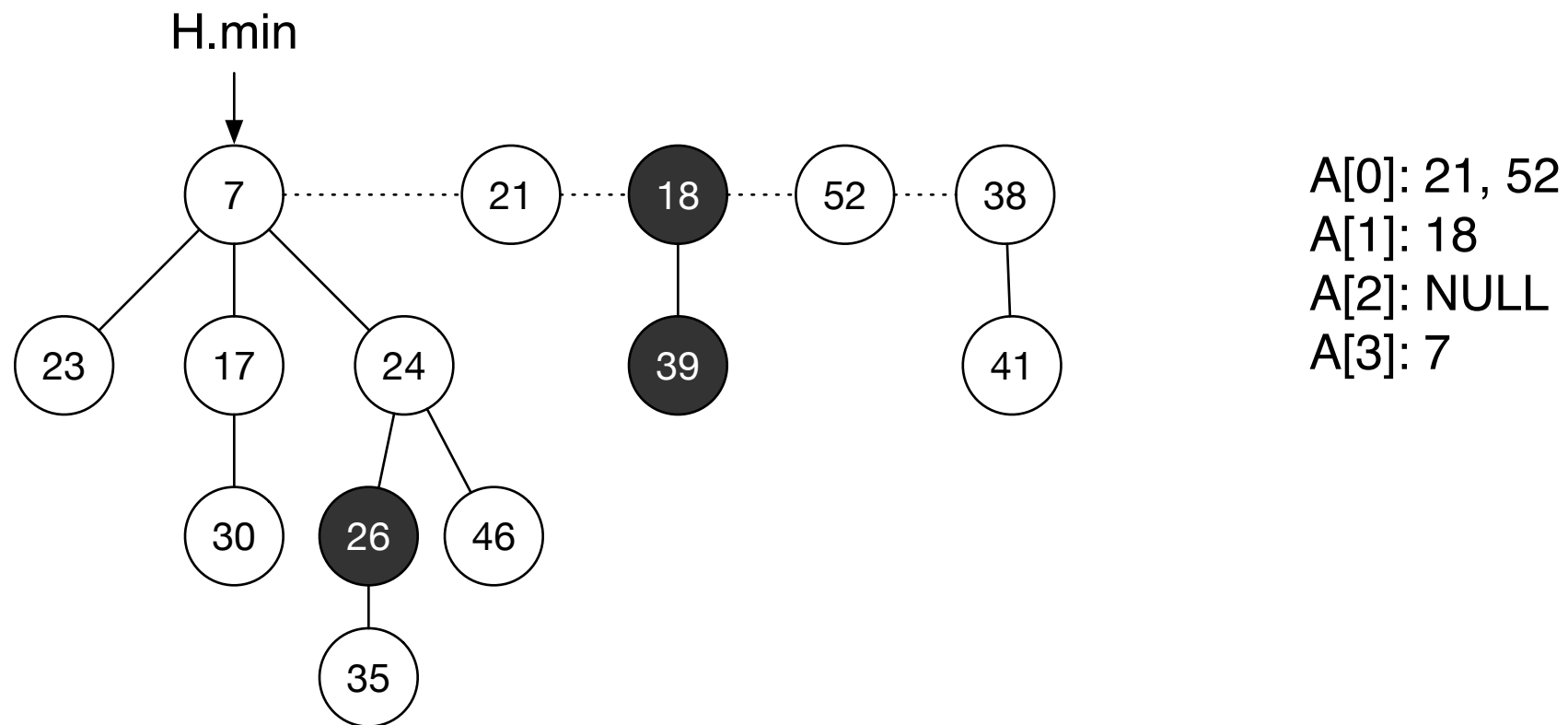
# Fibonacci Heaps



A[0]: 21  
A[1]: 18  
A[2]: NULL  
A[3]: 7

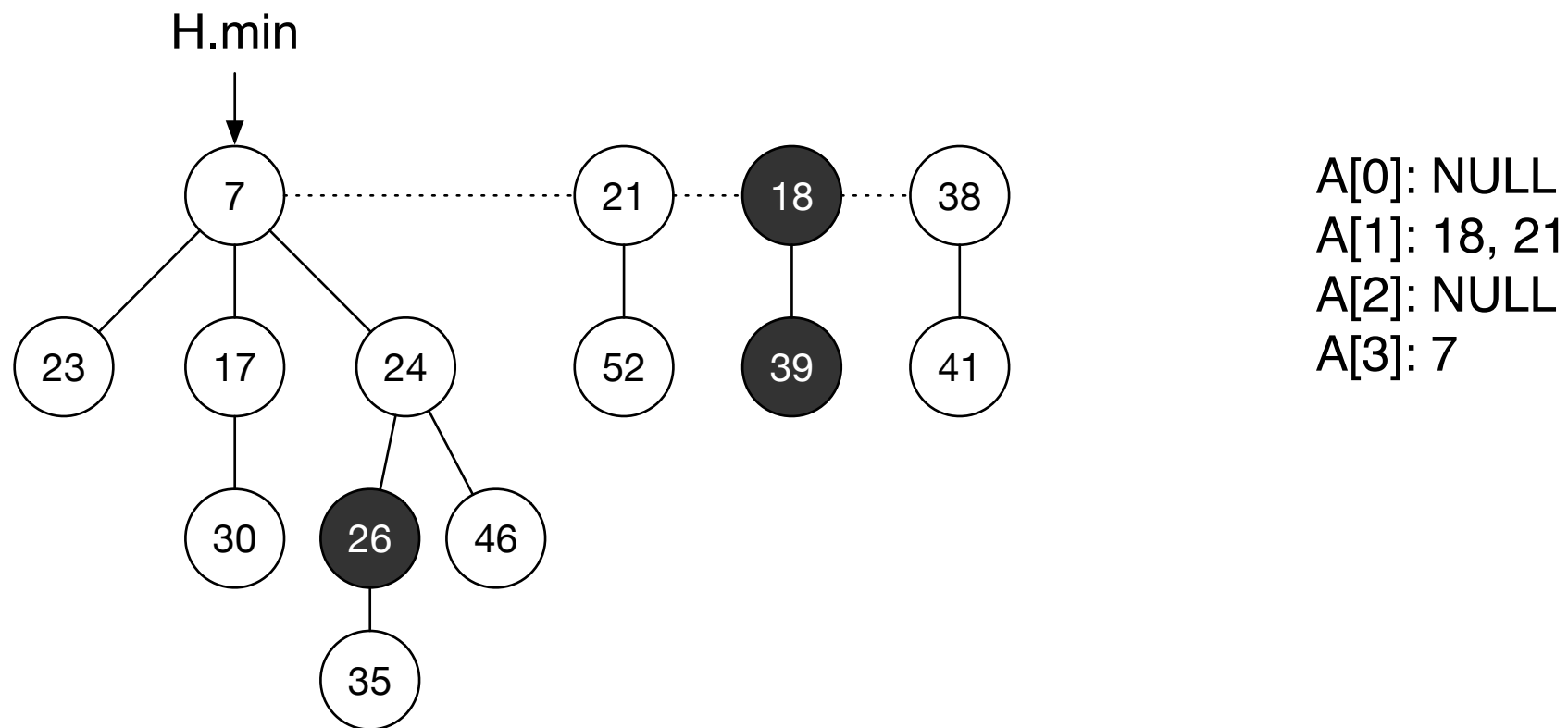
And then insert 18.

# Fibonacci Heaps



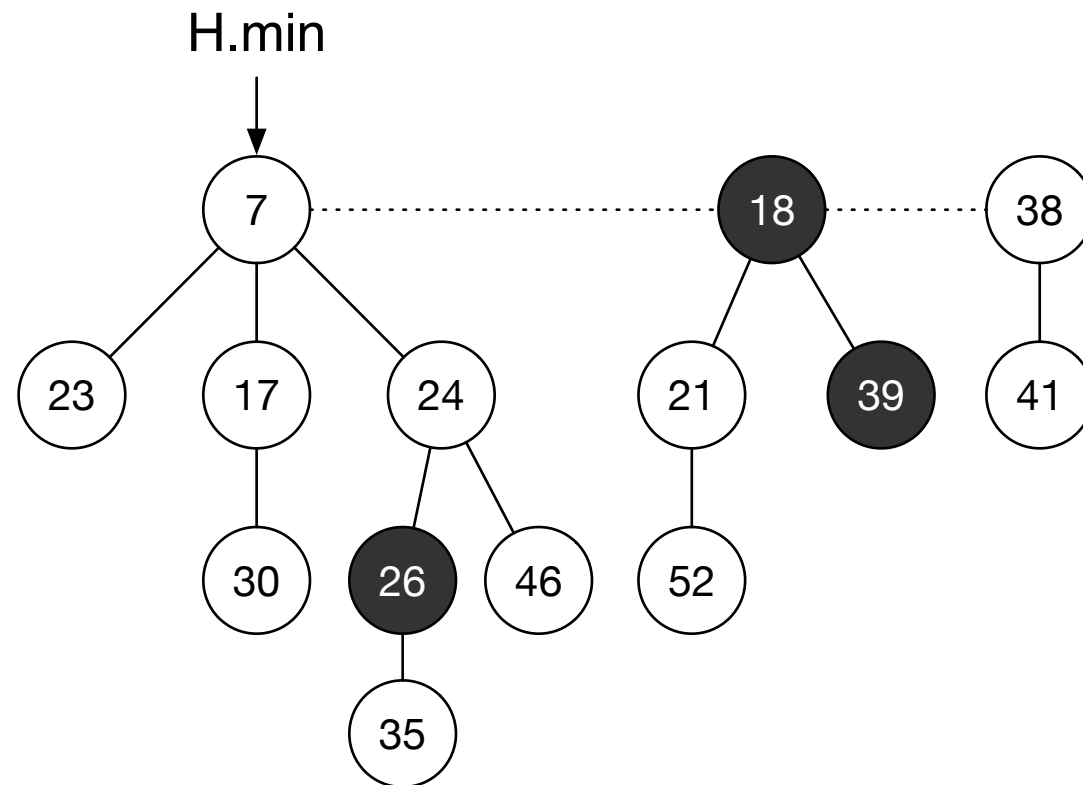
Inserting 52 gives us another merge candidate

# Fibonacci Heaps



Which leads to another merger

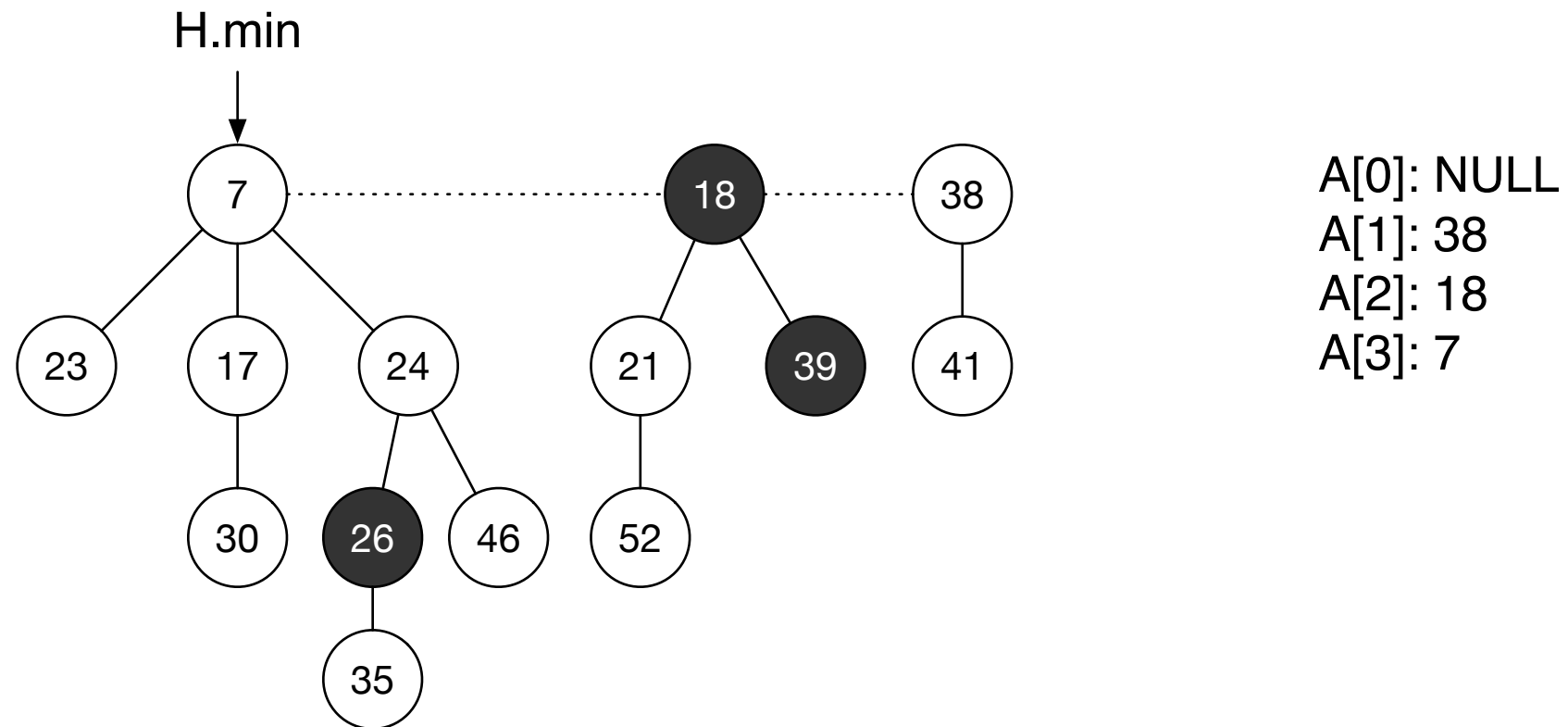
# Fibonacci Heaps



A[0]: NULL  
A[1]: NULL  
A[2]: 18  
A[3]: 7

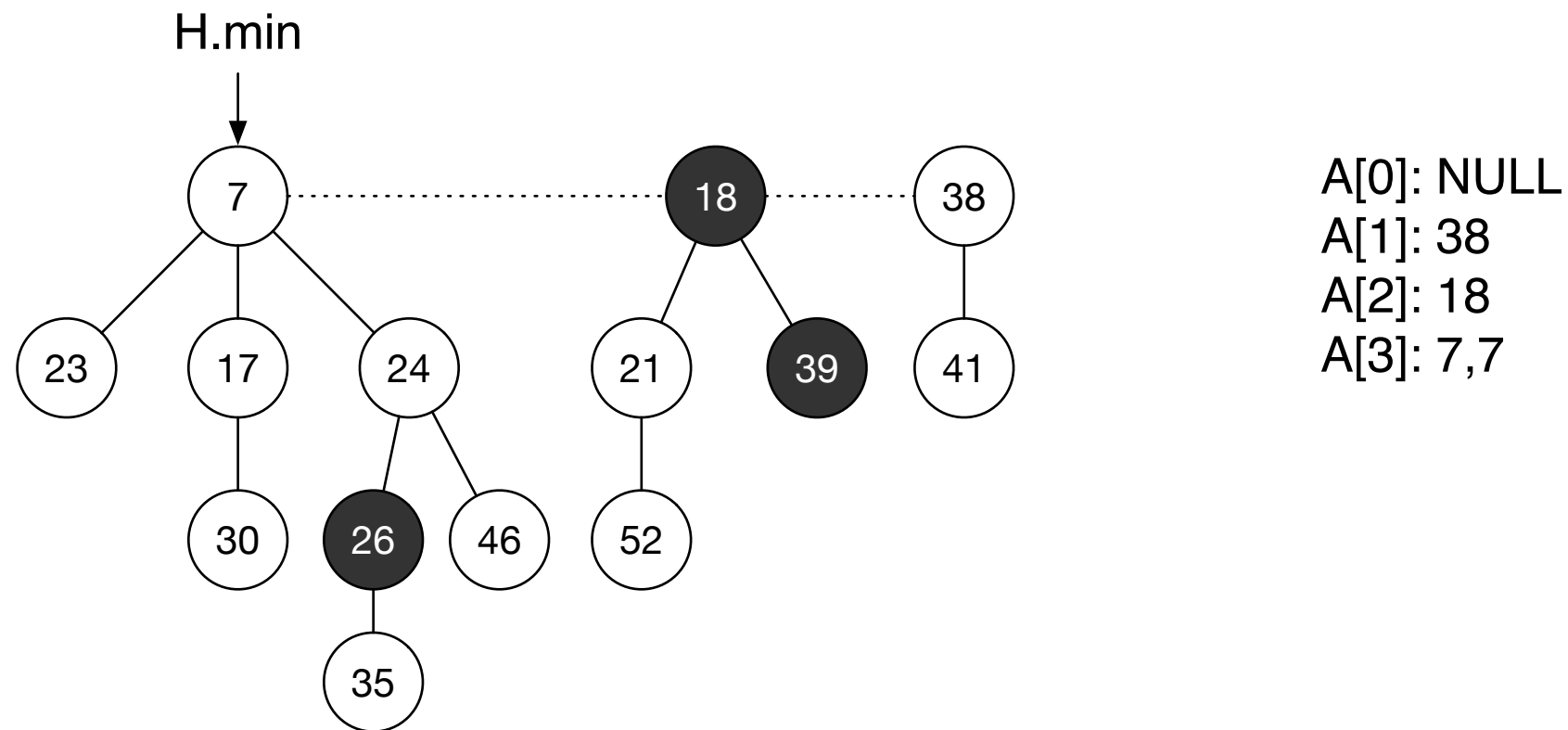
We move on to the next node

# Fibonacci Heaps



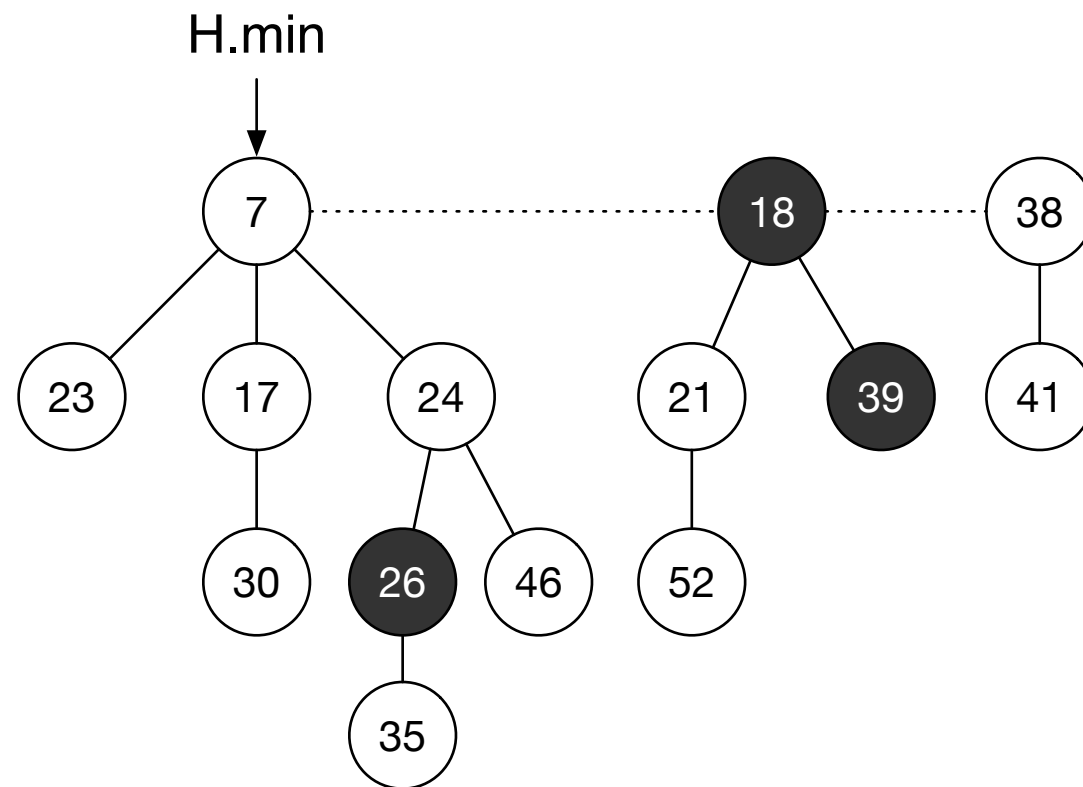
We insert it into the array.

# Fibonacci Heaps



We then move on to the next node. When we insert, we see that this one is already in A and that we are done merging.

# Fibonacci Heaps



A[0]: NULL  
A[1]: 38  
A[2]: 18  
A[3]: 7

We now find the new minimum.

# Fibonacci Heaps

- Combining two nodes
  - $x.key < y.key$

```
def fib_heap_link(H, y, x):  
    Assert x.key < y.key  
    remove y from root list of H  
    y.mark = False
```



# Fibonacci Heaps

- Consolidate:

```
def consolidate(H):
    A = [None for i in range(D(H.n))]
    for w in H.root_list:
        while (A[w.degree]):
            degree = A[w.degree]
            y = A[degree] # candidate for merger
            if w.key > y.key:
                w, y = y, w
            fib_heap_link(H, y, w)
            A[degree] = Null
            degree += 1
        A[w.degree] = w
    reset H.min
```

# Fibonacci Heaps

- Costs:
  - Potential:
    - Potential before is  $t(H) + 2m(H)$
    - Potential after is  $D(n) + 1 + 2m(H)$ 
      - because  $A$  has  $D(n)$  entries
      - because nobody gets marked
  - Work done:
    - $t(H)$  for going through the root list
    - $\leq D(n)$  for inserting the children of minimum
  - Amortized costs:
    - $\leq O(D(n) + t(H)) + (D(n) + 1 + 2m(H) - t(H) - 2m(H)) = O(D(n))$

# Fibonacci Heaps

- Decrease a key
  - Find parent  $p$ 
    - if  $p$  and  $x.key < p.key$ :
      - $CUT(H,x,y)$
      - $CASCADING-CUT(H,y)$
  - If necessary, adjust  $H.min$

# Fibonacci Heaps

- Cut(H, x, y)

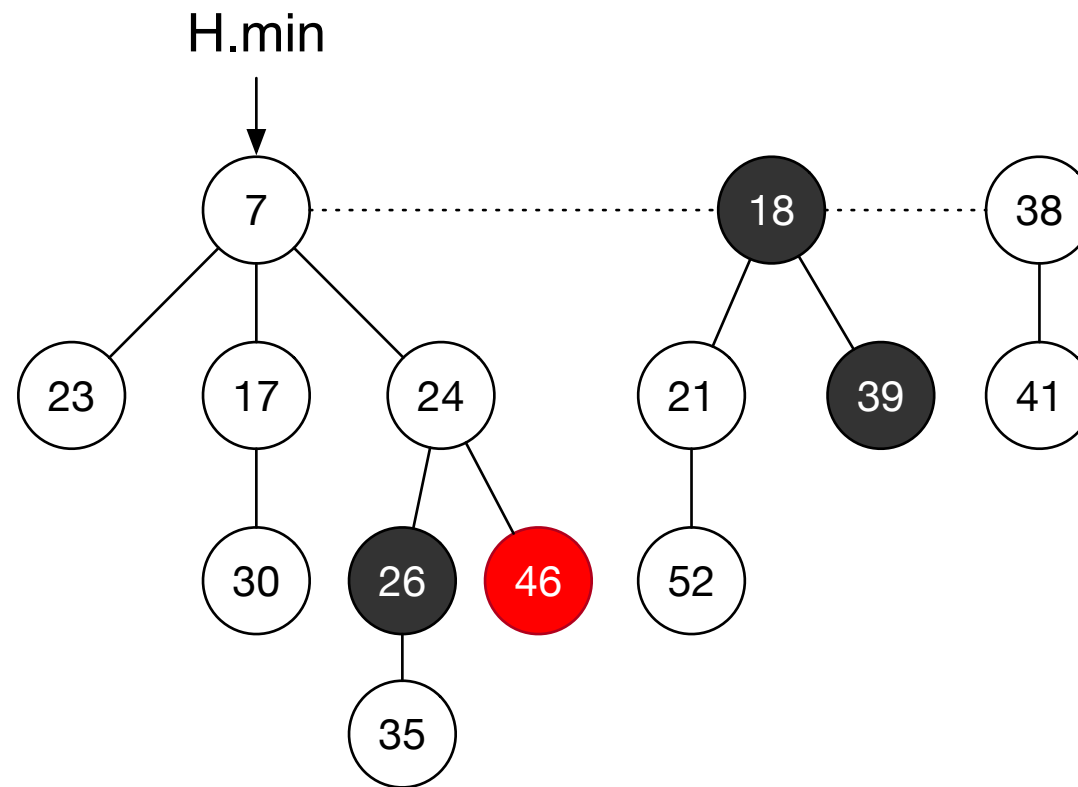
```
def cut(H, x, y):  
    remove x from child list of y  
    y.degree -= 1  
    add x to root list of H  
    x.p = Null  
    x.mark = False
```

# Fibonacci Heaps

- Cascading Cut

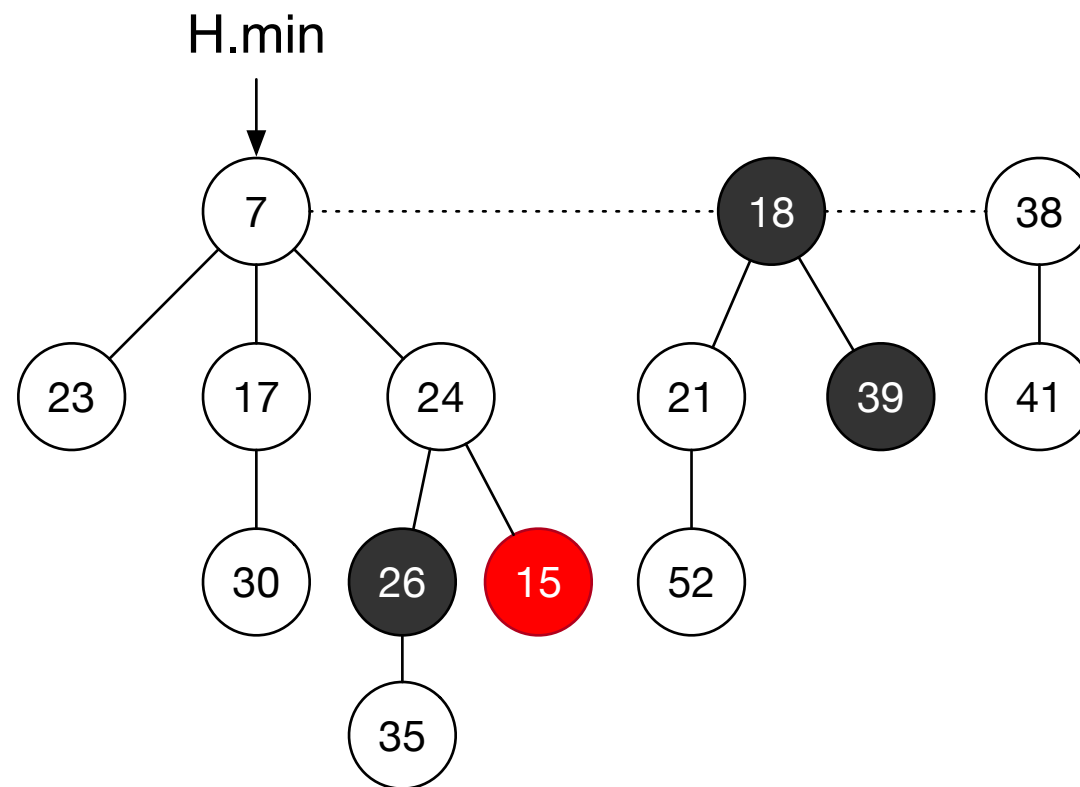
```
def cascading_cut(H, y):  
    z = y.p # parent of y  
    if z:  
        if y.mark == False:  
            y.mark = True  
        else:  
            cut(H, y, z)  
            cascading_cut(H, z)
```

# Fibonacci Heaps

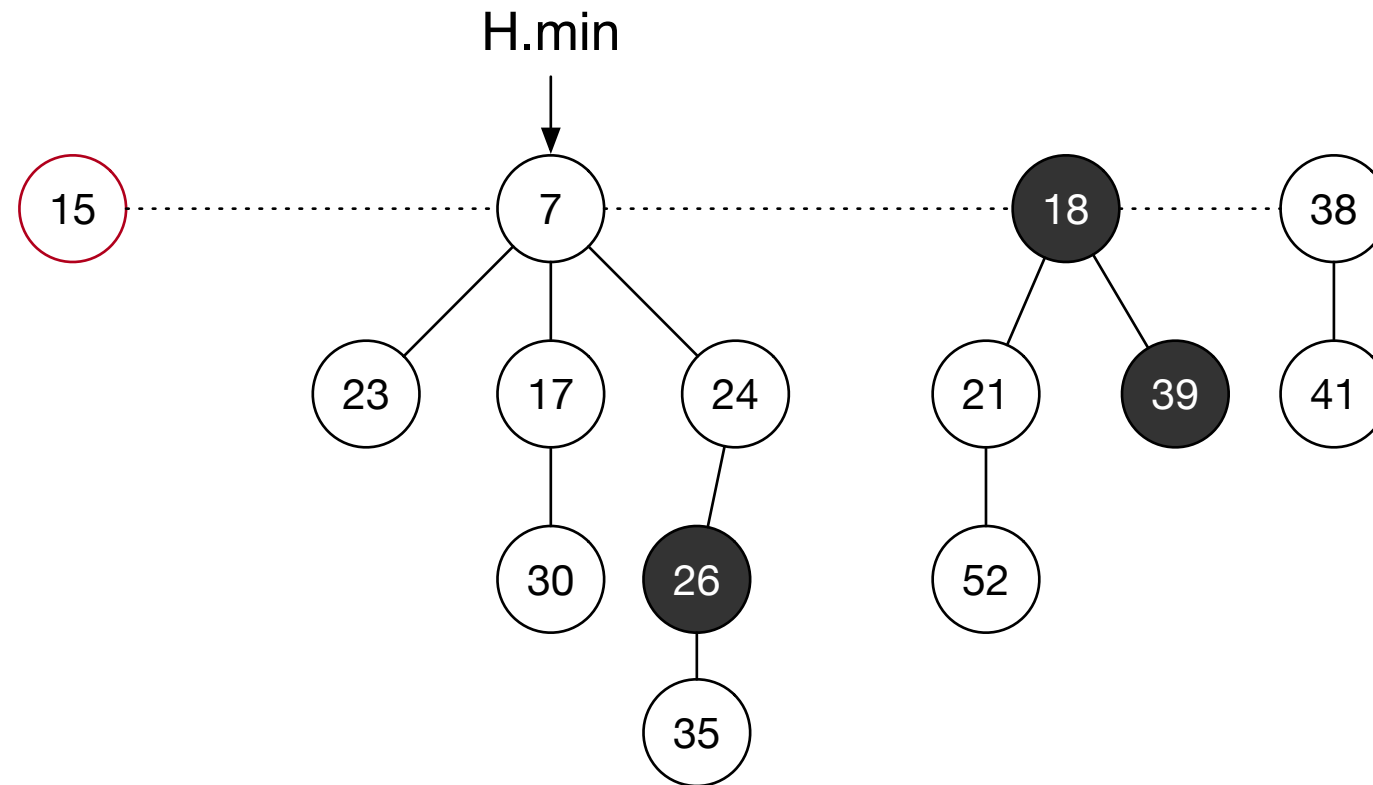


Decrease key to 15

# Fibonacci Heaps



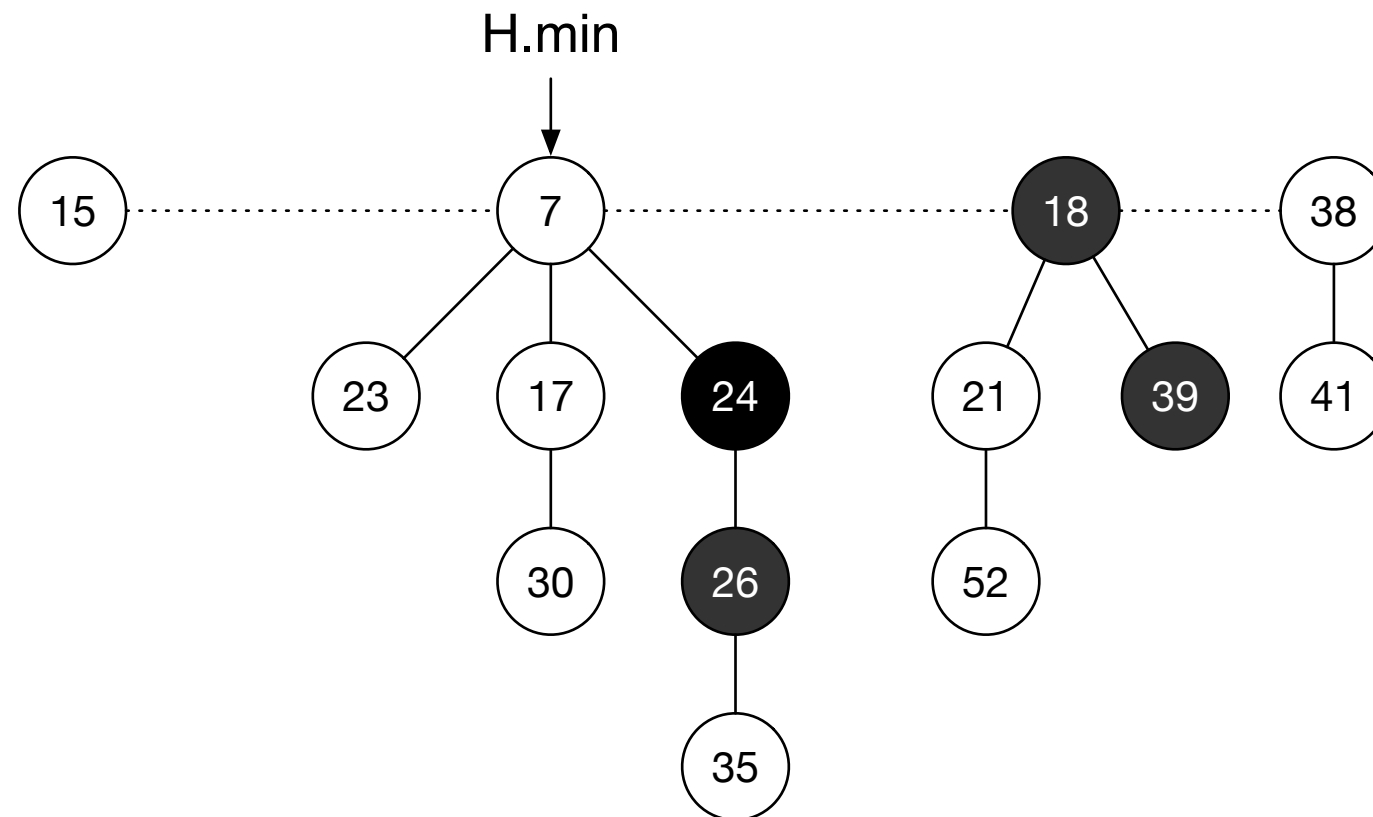
# Fibonacci Heaps



Remove node (and any descendants)



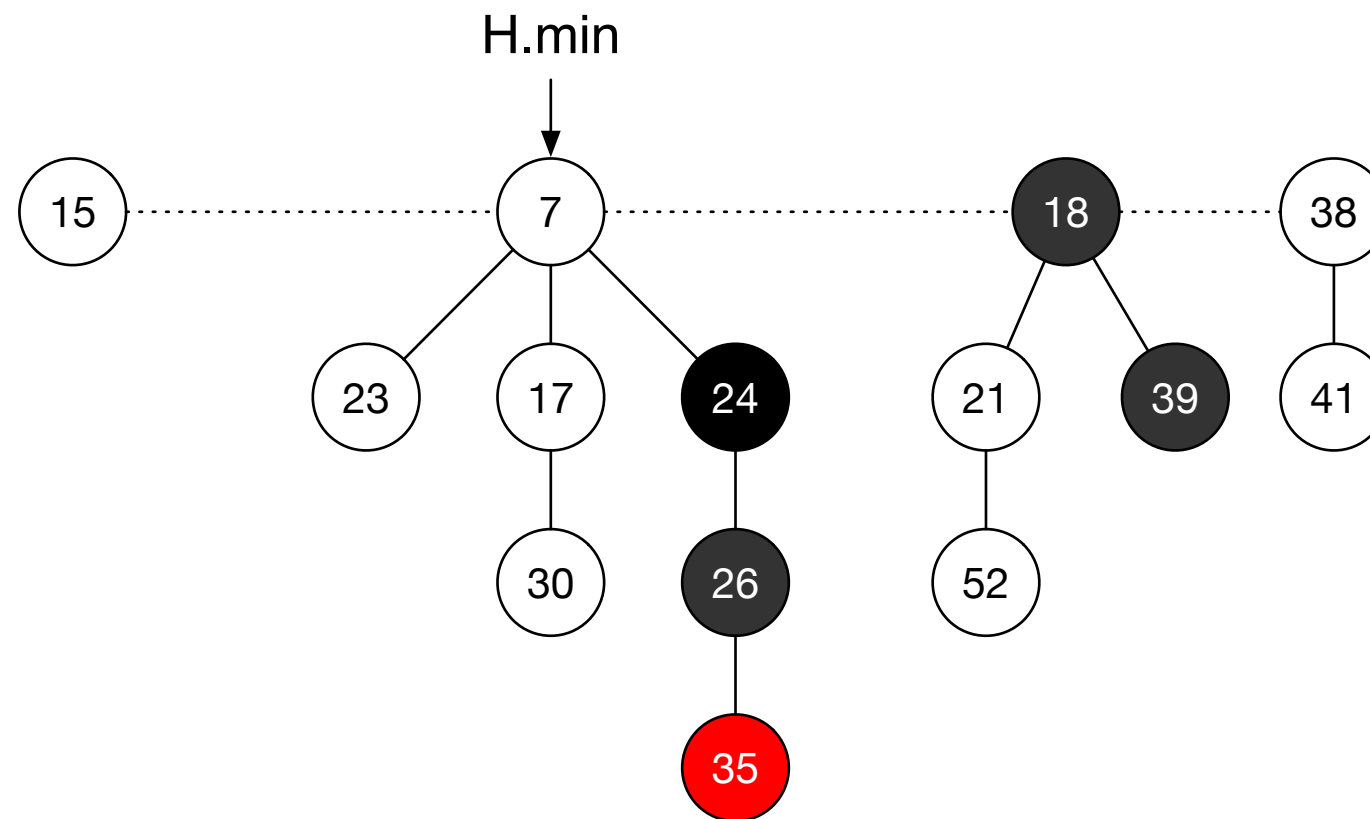
# Fibonacci Heaps



Mark parent!  
Done!

# Fibonacci Heaps

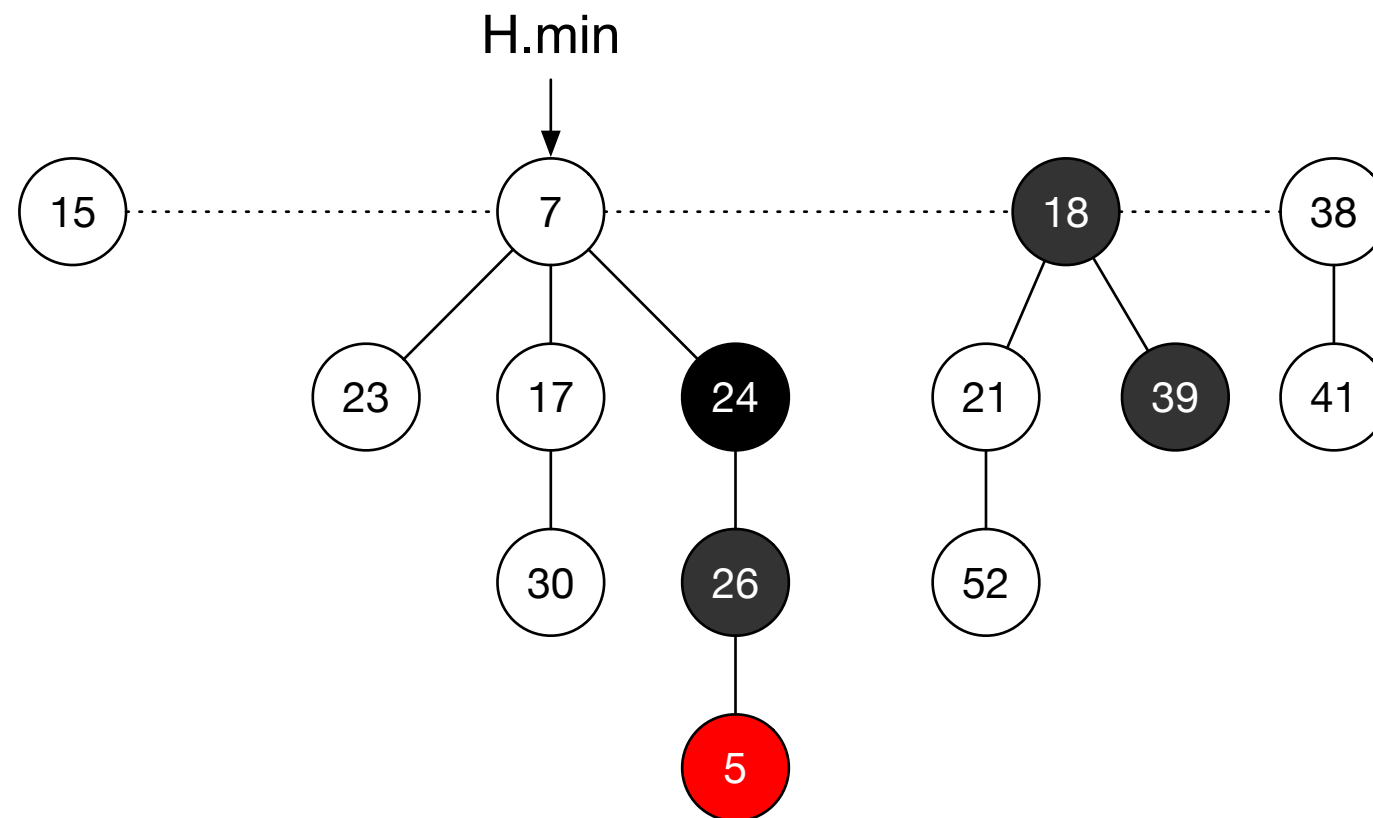
Second operation



Change key from 35 to 5

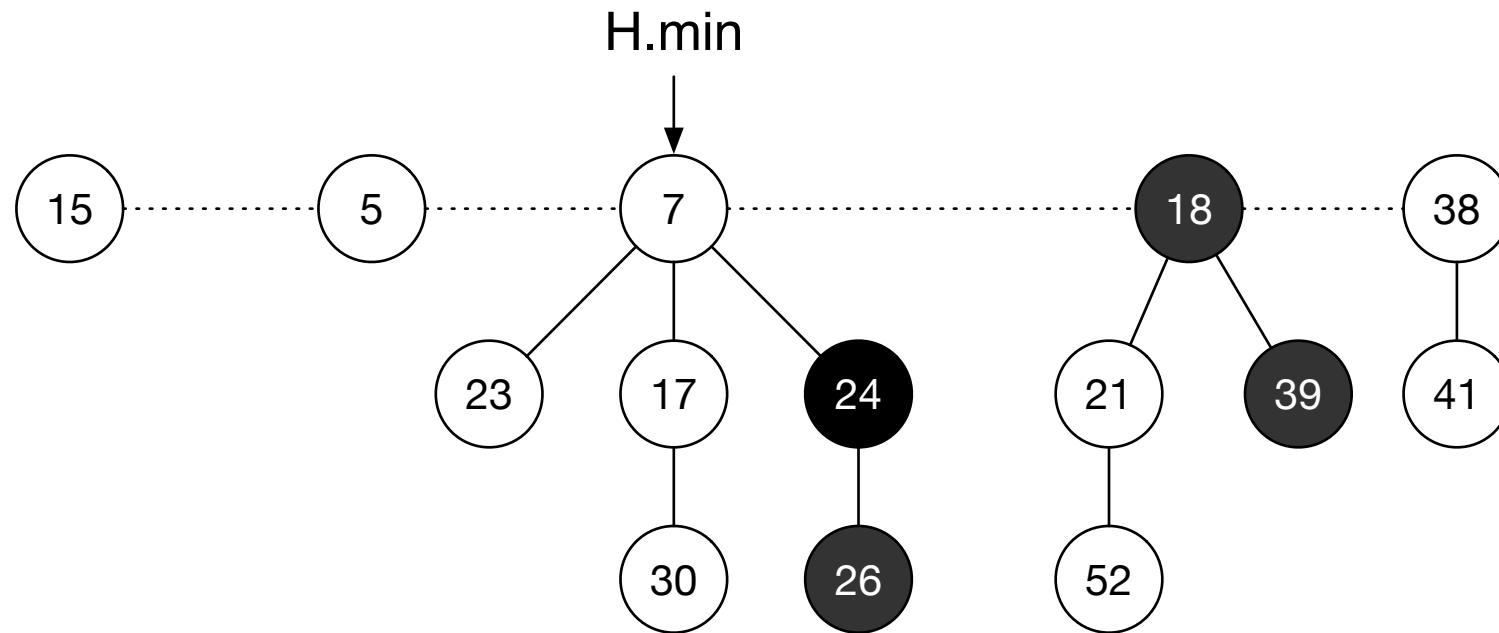
# Fibonacci Heaps

Second operation



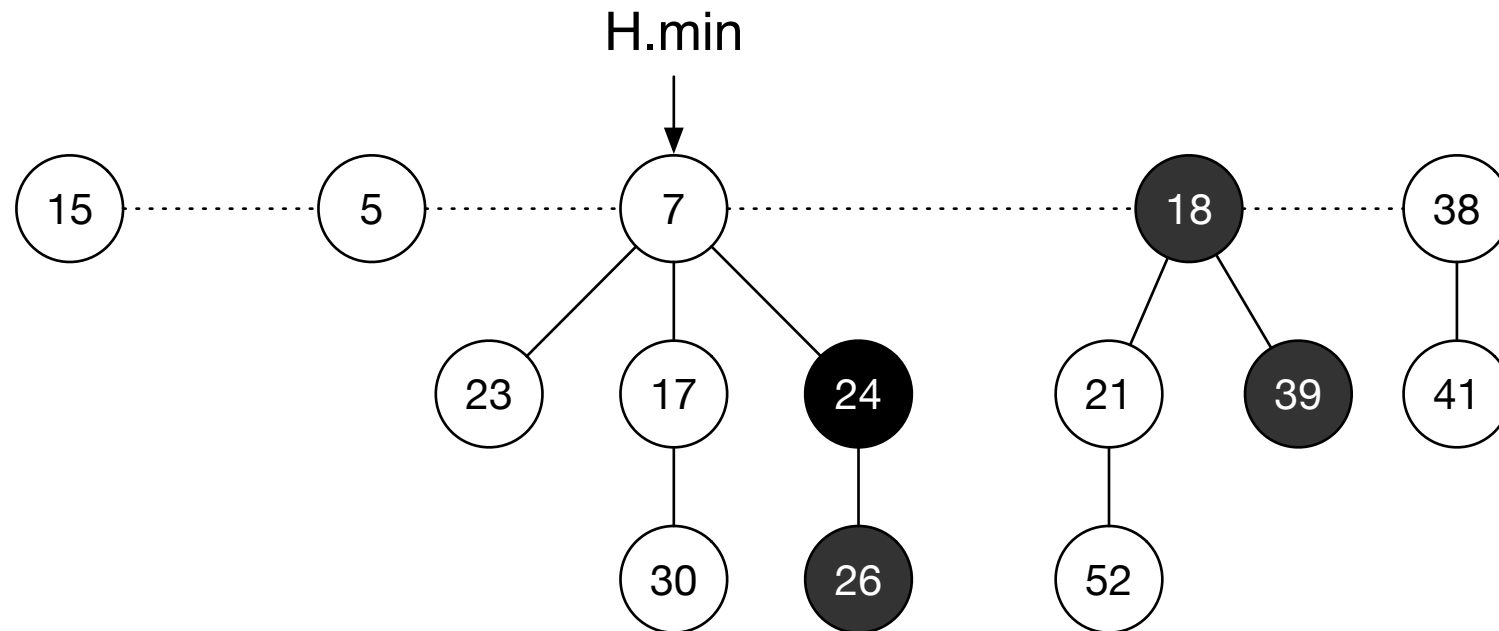
Change key from 35 to 5

# Fibonacci Heaps



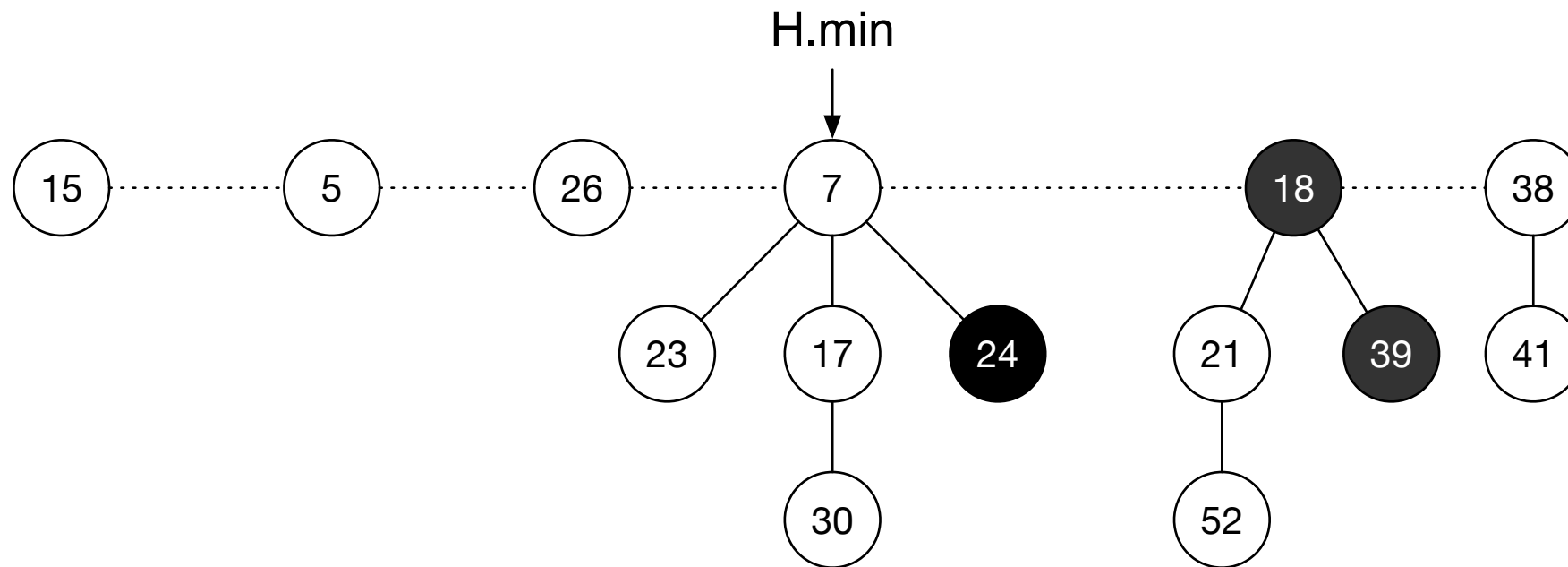
Cut node.

# Fibonacci Heaps



Mark parent.  
But parent is already marked, so:  
Cascading Cut!

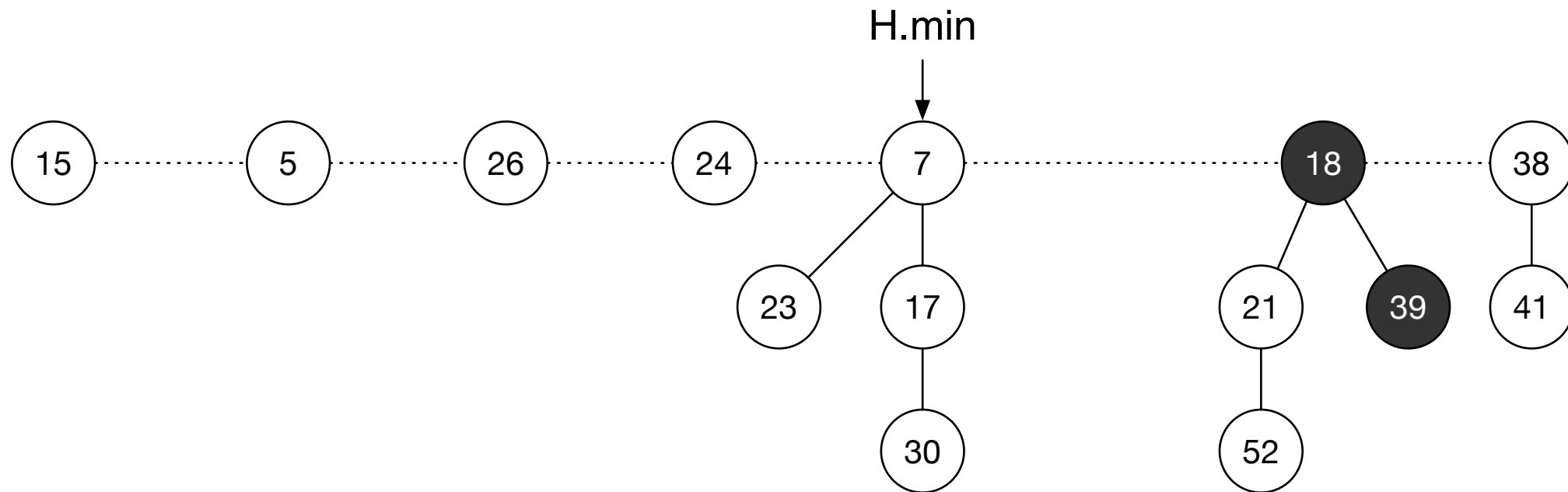
# Fibonacci Heaps



26 is removed.

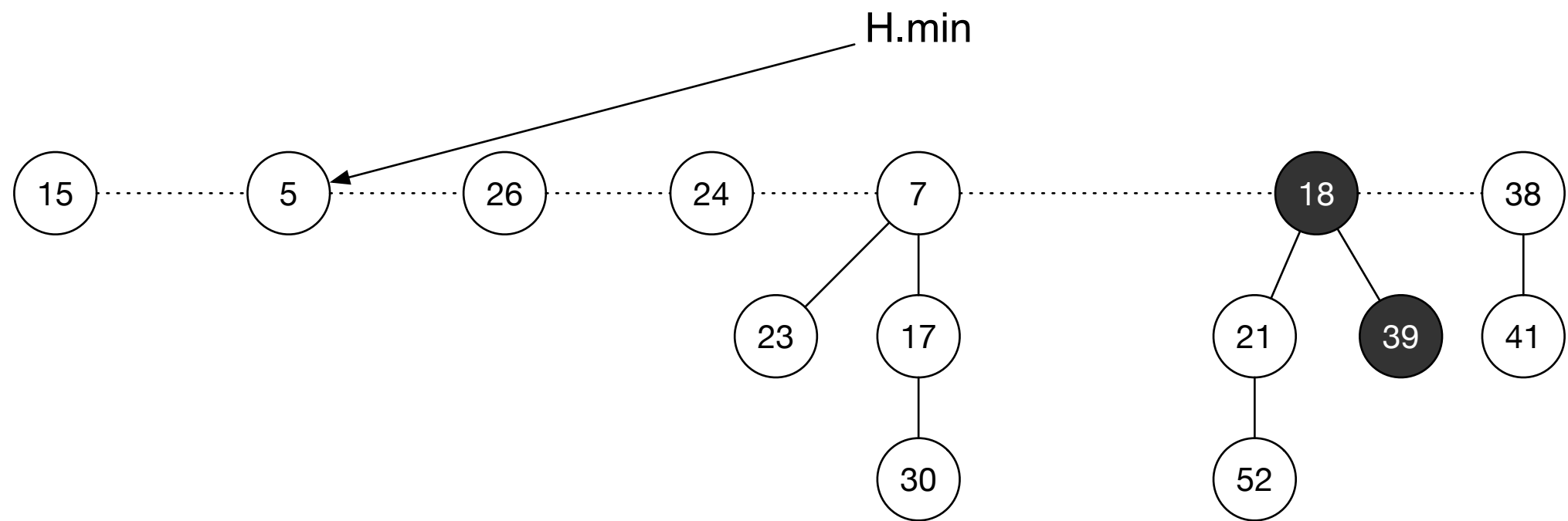
Look at 24: It is also marked

# Fibonacci Heaps



Cut 24, unmarking it.  
Cut stops here, parent is unmarked.

# Fibonacci Heaps



Reset minimum



# Fibonacci Heaps

- Change in potential:
  - Cut creates a new tree and potentially clears a mark
  - Each Cascading-Cut cuts a marked node and clears the mark bit (with exception of the last one)
- H now has  $c-1$  trees produced by cascading cuts and the tree at  $x$ :  $t(H)+c$  trees
- At most  $m(H)-c+2$  marked nodes
- $\leq t(H) + c + 2(m(H) - c + 2) - t(H) - 2m(H) = 4 - c$

# Fibonacci Heaps

- Amortized cost of decrease key:
  - $O(c) + 4 - c = O(1)$

# Fibonacci Heaps

- Deleting a node

- ```
def delete(H, x):  
    decrease_key(H, x, -infty)  
    extract_min(H)
```

# Fibonacci Heaps

- Left to investigate:
  - Upper bound  $D(n)$  on the degrees is  $O(\log(n))$

- Lemma:  $F_{k+2} = 1 + \sum_{i=0}^k F_i$

# Fibonacci Heaps

- Lemma:  $F_{k+2} \geq \phi^k$  with  $\phi = \frac{1 + \sqrt{5}}{2}$

# Fibonacci Heaps

- Lemma: Let  $x$  be any node in a Fibonacci heap with  $x.\text{degree} = k$ . Let  $y_1, y_2, \dots, y_k$  be the children of  $x$  in the order in which they were linked to  $x$  (by consolidate).  
Then:  
 $y_1.\text{degree} \geq 0$ ,  $y_i.\text{degree} \geq i - 2$  for all  $i = 2, 3, \dots, k$

# Fibonacci Heaps

- Proof:
  - Clearly,  $y_1$  . degree  $\geq 0$
  - Assume a general  $i \geq 2$ .
    - When CONSOLIDATE links  $y_i$  to  $x$ , then all of  $y_1, y_2, \dots, y_{i-1}$  was linked to  $x$
    - This means  $x$  . degree  $\geq i$
    - Because  $y_i$  is linked to  $x$  only if  $x$  . degree =  $y_i$  . degree
      - $y_i$  . degree  $\geq i - 1$

# Fibonacci Heaps

- But in the mean-time, the degree of  $y_i$  might have decreased
- But Cascading-Cut would cut  $y_i$  from  $x$  if  $y_i$  has lost more than two children
- Therefore  $y_i$  . degree  $\geq i - 2$



# Fibonacci Heaps

- Lemma: Let  $x$  be any node in a Fibonacci heap and let  $k = x$ . degree. Then  $\text{size}(x) \geq F_{k+2} \geq \Phi^k$ .
- Let  $s_k$  denote the minimum possible size of any node of degree  $k$  in a Fibonacci heap.
- $s_0 = 1, \quad s_1 = 2$
- Adding children does not decrease the node's size, the value of  $s_k$  increases monotonically with  $k$ .

# Fibonacci Heaps

- Notice  $s_k \leq \text{size}(x)$ , so giving a lower bound on  $s_k$  is sufficient
- Take a node  $z$  with
  - $z$ . degree =  $k$        $\text{size}(z) = s_k$
- Let  $y_1, y_2, \dots, y_k$  denote the children of  $z$  in the order in which they were linked to  $z$
- To bound  $s_k$ , we have one for  $z$  and one for  $y_1$  and the rest

# Fibonacci Heaps

- This gives
  - $\text{size}(x) \geq s_k$
  - $\geq 2 + \sum_{i=2}^k s_{y_i \cdot \text{degree}}$
  - $\geq 2 + \sum_{i=2}^k s_{i-2}$  because  $y_i \cdot \text{degree} \geq i - 2$  by Lemma and monotonicity of  $s_k$

# Fibonacci Heaps

- We prove by induction that  $s_i \geq F_{i+2}$

- Induction step:

- $s_k \geq 2 + \sum_{i=2}^k s_{i-2}$

- $\geq 2 + \sum_{i=2}^k F_i$

- $= 1 + \sum_{i=0}^k F_i$

- $= F_{k+2} \geq \Phi^k$

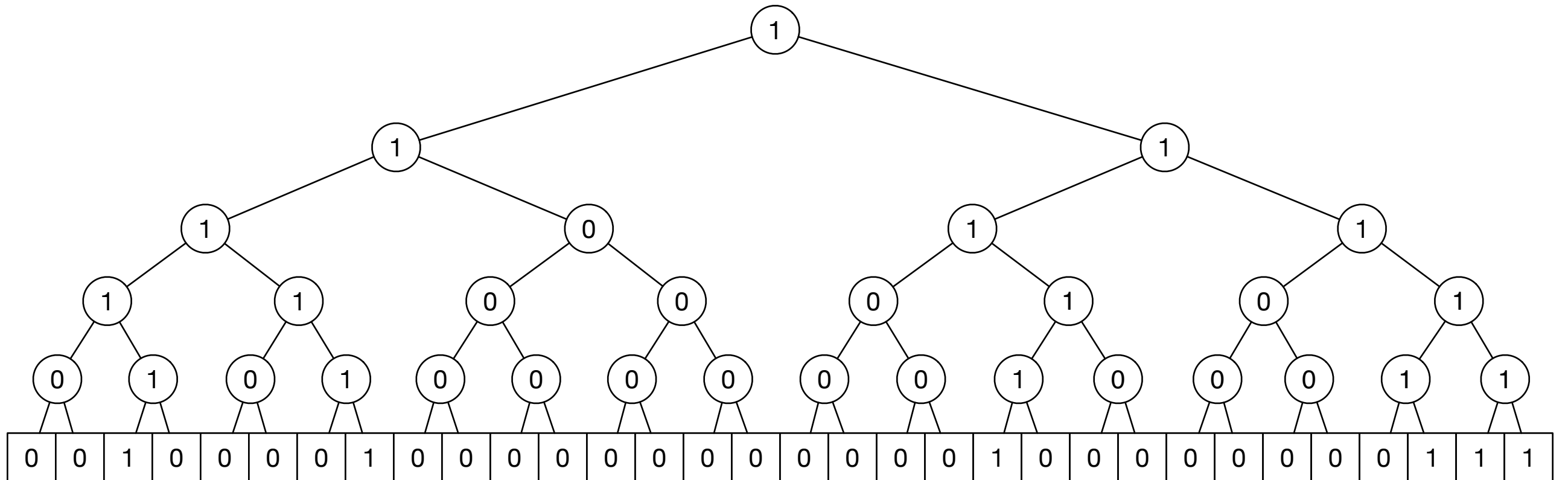
# Fibonacci Heaps

- Let  $x$  be the node with maximum degree into an  $n$ -node Fibonacci Heap
- Degree of  $x$  is  $k$
- By lemma,  $n \geq \text{size}(x) \geq \Phi^k$
- Thus:  $\log_{\Phi}(n) \geq k$
- Thus: maximum degree is  $O(\log n)$

# van Emde Boas Trees

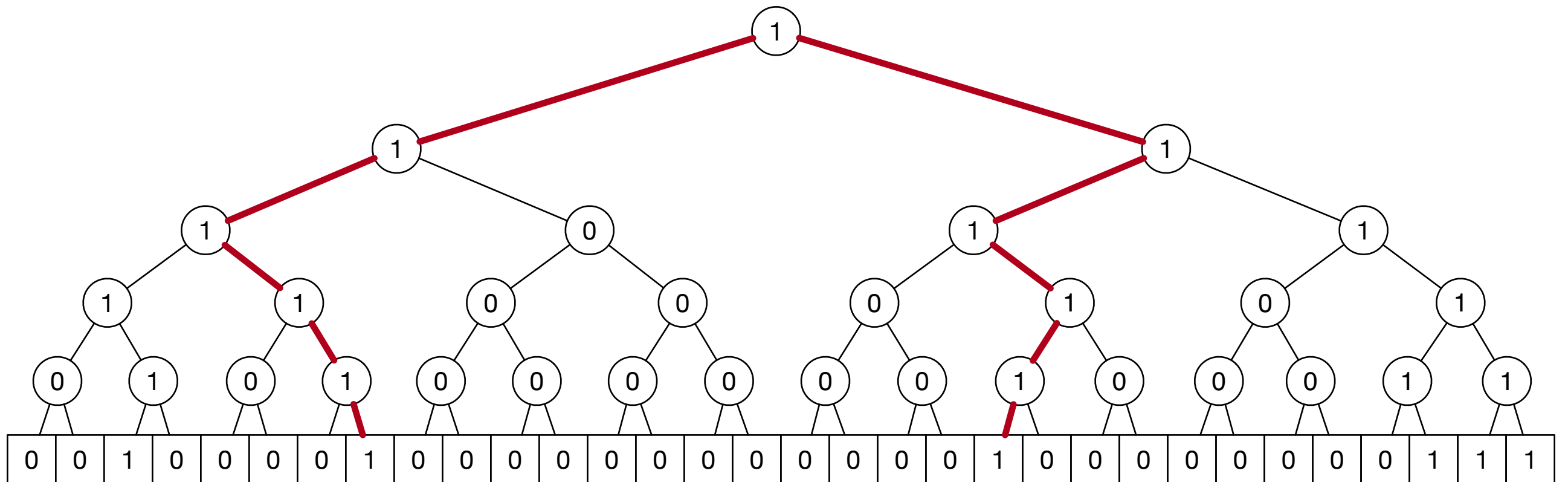
- Operations of priority queues have at least one logarithmic operation
- If everything would be less, then we could sort in time  $n \cdot \log n$
- But we can sort an array of integers in range  $1 \dots n$  in time  $O(n)$

# van Emde Boas Trees



Binary tree on top of a bitvector

# van Emde Boas Trees



Finding the NextIndex / PreviousIndex  
is logarithmic





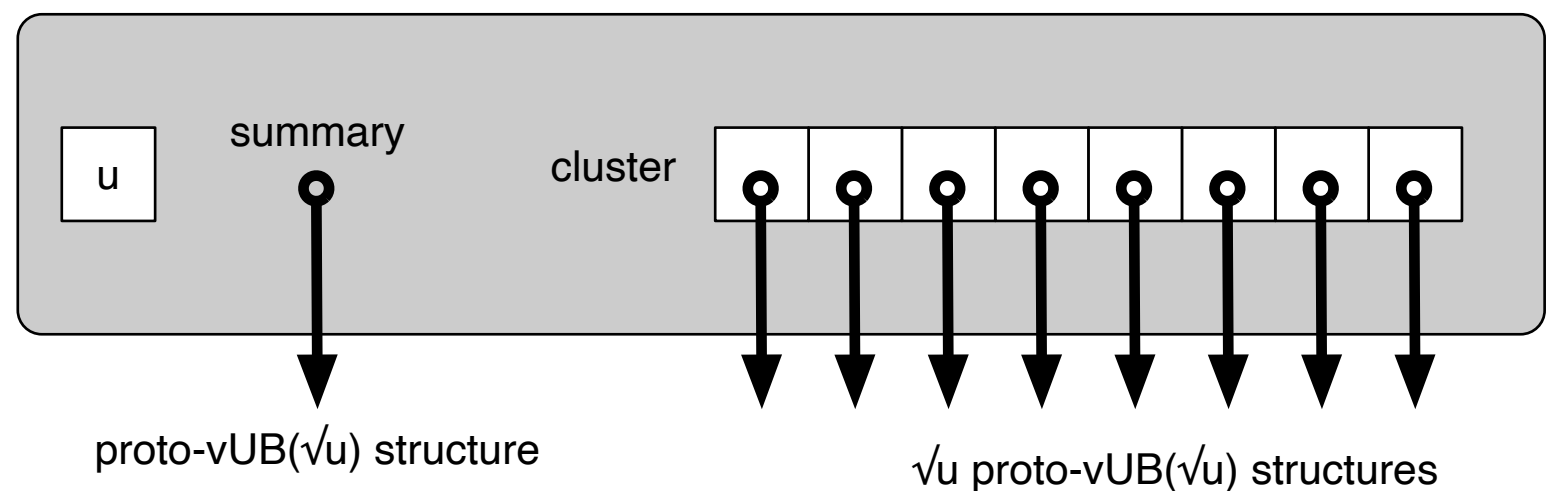


# van Emde Boas Trees

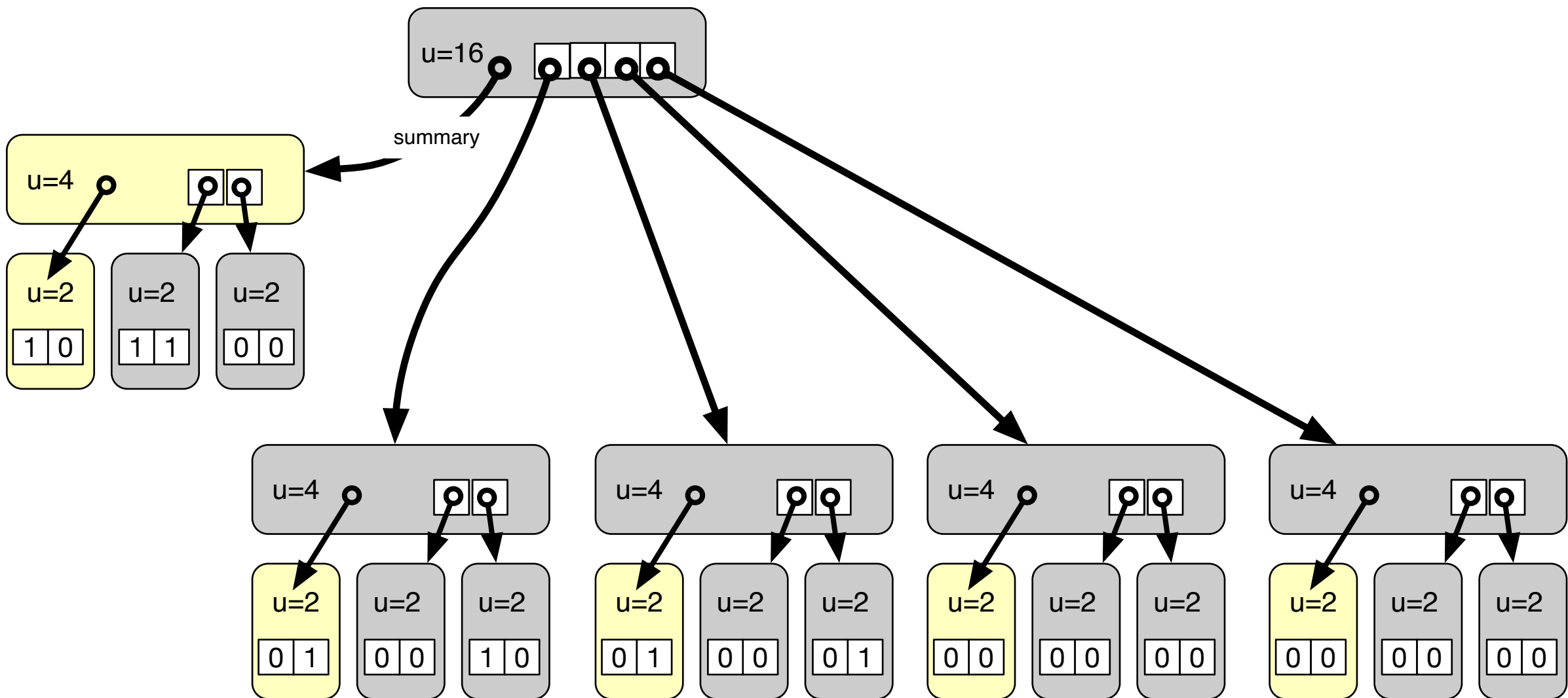
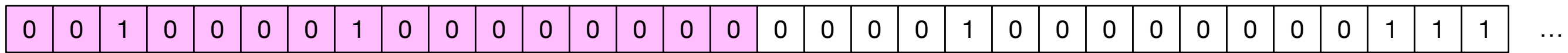
- Assume that the number  $u$  of elements in the universe is  $u = 2^{2^k}$
- To build a recursive structure:
  - Recurrence  $T(u) = T(\sqrt{u}) + O(1)$ 
    - Setting  $m = \log_2(u)$ ,  $S(m) = T(2^m)$
    - Recurrence is  $S(m) = S(m/2) + O(1)$
    - MT  $\implies S(m) = O(\log_2(m))$
    - $T(u) = T(2^m) = S(m) = O(\log_2(m)) = O(\log_2(\log_2(u)))$

# van Emde Boas Trees

- Proto-vEM-structures for  $u$ 
  - $u = 2$ : array of two bits
  - $u = 2^{2^k}$  with  $k \geq 1$ :
    - summary pointer towards parent
    - cluster towards children



# van Emde Boas Trees



# van Emde Boas Trees

- Operations on proto-vEB-Trees

- For a given  $x \in U$ :

- $\text{high}(x) = \lfloor \frac{x}{\sqrt{u}} \rfloor$  number of cluster

- $\text{low}(x) = x \pmod{\sqrt{u}}$  position of  $x$  in cluster

- $\text{index}(x, y) = x\sqrt{u} + y$  rebuilds index from number of cluster and position in cluster

# van Emde Boas Trees

- Operations on proto-vEB-Trees
  - Membership

```
def pvEB_member(V, x):  
    if V.u == 2:  
        return V.A[x]  
    else:  
        return pvEB_member(V.cluster[high(x)].low(x))
```

- Runtime has recurrence  $T(u) = T(\sqrt{u}) + O(1)$

# van Emde Boas Trees

- Operations on proto-vEB-Trees
  -



# van Emde Boas Trees

- Operations on proto-vEB-Trees

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