

# Binomial Heaps

# Where We're Going

- ***Binomial Heaps (Today)***
  - A simple, flexible, and versatile priority queue.
- ***Lazy Binomial Heaps (Today)***
  - A powerful building block for designing advanced data structures.
- ***Fibonacci Heaps (Tuesday)***
  - A heavyweight and theoretically excellent priority queue.

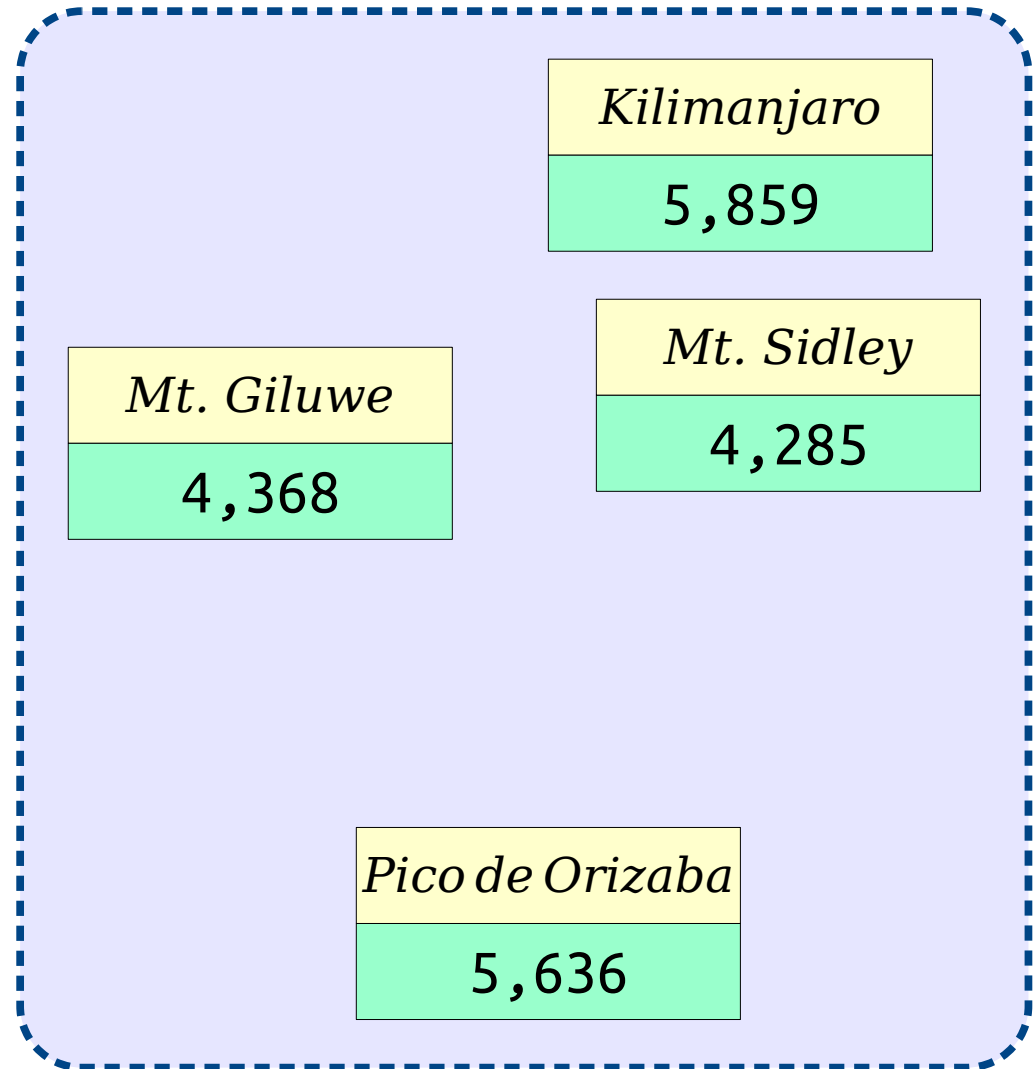
# ***Review:*** Priority Queues

# Priority Queues

- A **priority queue** is a data structure that supports these operations:
  - $pq.enqueue(v, k)$ , which enqueues element  $v$  with key  $k$ ;
  - $pq.find-min()$ , which returns the element with the least key; and
  - $pq.extract-min()$ , which removes and returns the element with the least key.
- They're useful as building blocks in a *bunch* of algorithms.

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# Binary Heaps

- Priority queues are frequently implemented as **binary heaps**.
  - **enqueue** and **extract-min** run in time  $O(\log n)$ ; **find-min** runs in time  $O(1)$ .
- These heaps are surprisingly fast in practice. It's tough to beat their performance!
  - $d$ -ary heaps can outperform binary heaps for a well-tuned value of  $d$ , and otherwise only the **sequence heap** is known to specifically outperform this family.
  - (Is this information incorrect as of 2021? Let me know and I'll update it.)
- In that case, why do we need other heaps?

# Priority Queues in Practice

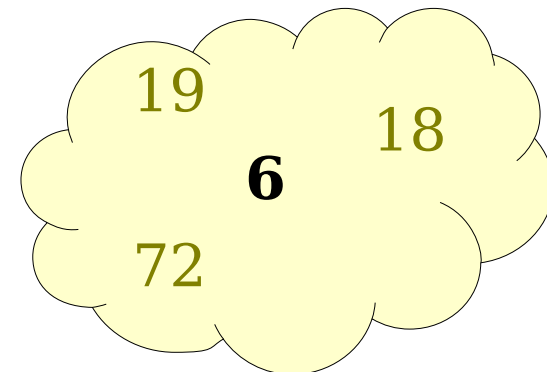
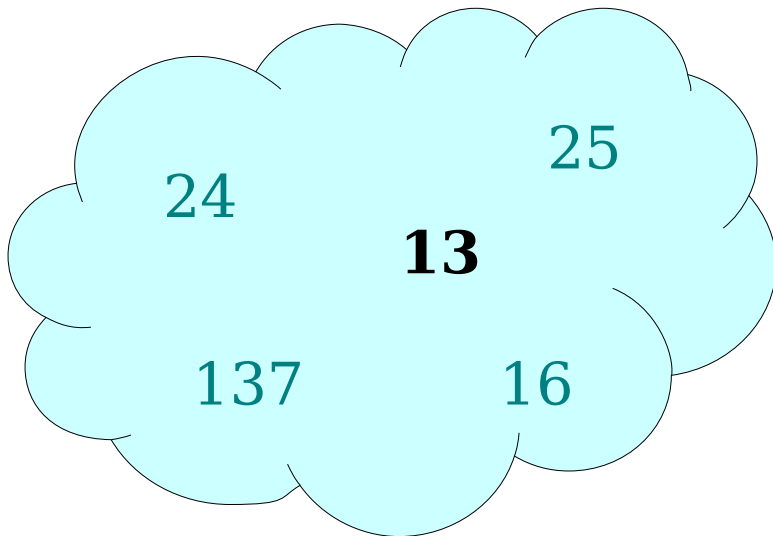
- Many graph algorithms directly rely on priority queues supporting extra operations:
  - ***meld***( $pq_1, pq_2$ ): Destroy  $pq_1$  and  $pq_2$  and combine their elements into a single priority queue. (*MSTs via Cheriton-Tarjan*)
  - $pq$ .***decrease-key***( $v, k'$ ): Given a pointer to element  $v$  already in the queue, lower its key to have new value  $k'$ . (*Shortest paths via Dijkstra, global min-cut via Stoer-Wagner*)
  - $pq$ .***add-to-all***( $\Delta k$ ): Add  $\Delta k$  to the keys of each element in the priority queue, typically used with ***meld***. (*Optimum branchings via Chu-Edmonds-Liu*)
- In lecture, we'll cover binomial heaps to efficiently support ***meld*** and Fibonacci heaps to efficiently support ***meld*** and ***decrease-key***.
- You'll design a priority queue supporting ***meld*** and ***add-to-all*** on the next problem set.

# Meldable Priority Queues



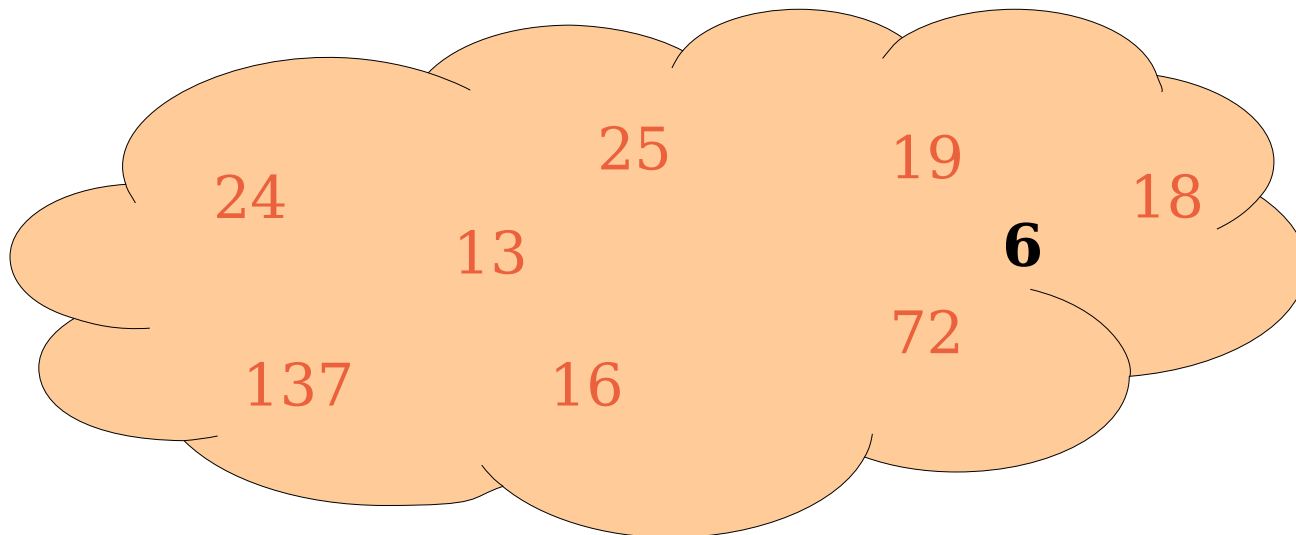
# Meldable Priority Queues

- A priority queue supporting the *meld* operation is called a *meldable priority queue*.
- *meld*( $pq_1, pq_2$ ) destructively modifies  $pq_1$  and  $pq_2$  and produces a new priority queue containing all elements of  $pq_1$  and  $pq_2$ .



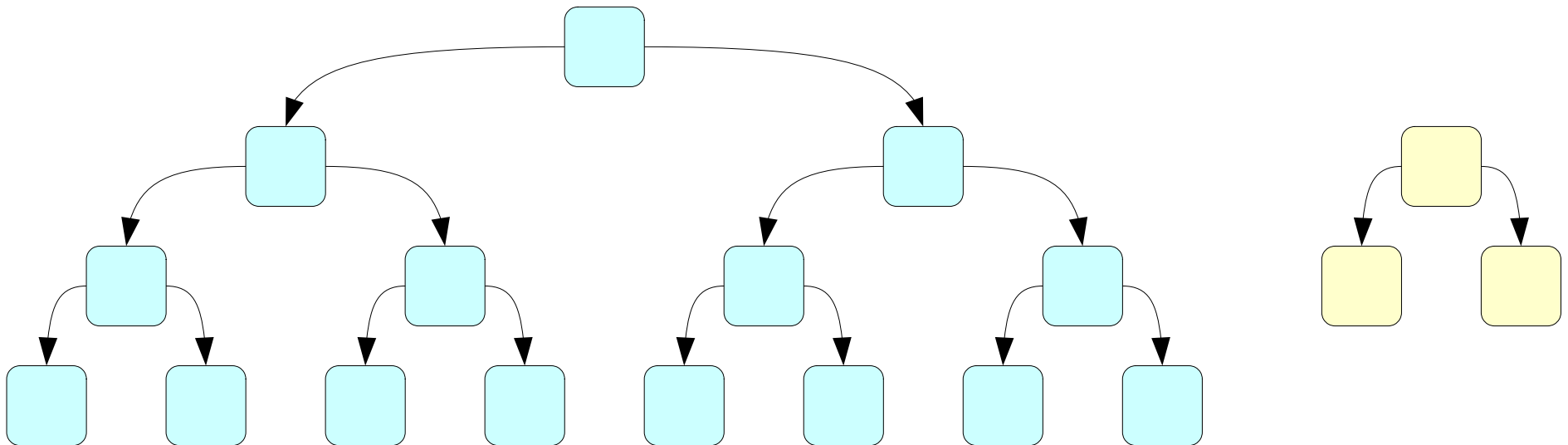
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# Efficiently Meldable Queues

- Standard binary heaps do not efficiently support *meld*.
- **Intuition**: Binary heaps are complete binary trees, and two complete binary trees cannot easily be linked to one another.



What things *can* be combined together  
efficiently?

# Adding Binary Numbers

- Given the binary representations of two numbers  $n$  and  $m$ , we can add those numbers in time  $O(\log m + \log n)$ .

***Intuition:***

Writing out  $n$  in any “reasonable” base requires  $\Theta(\log n)$  digits.

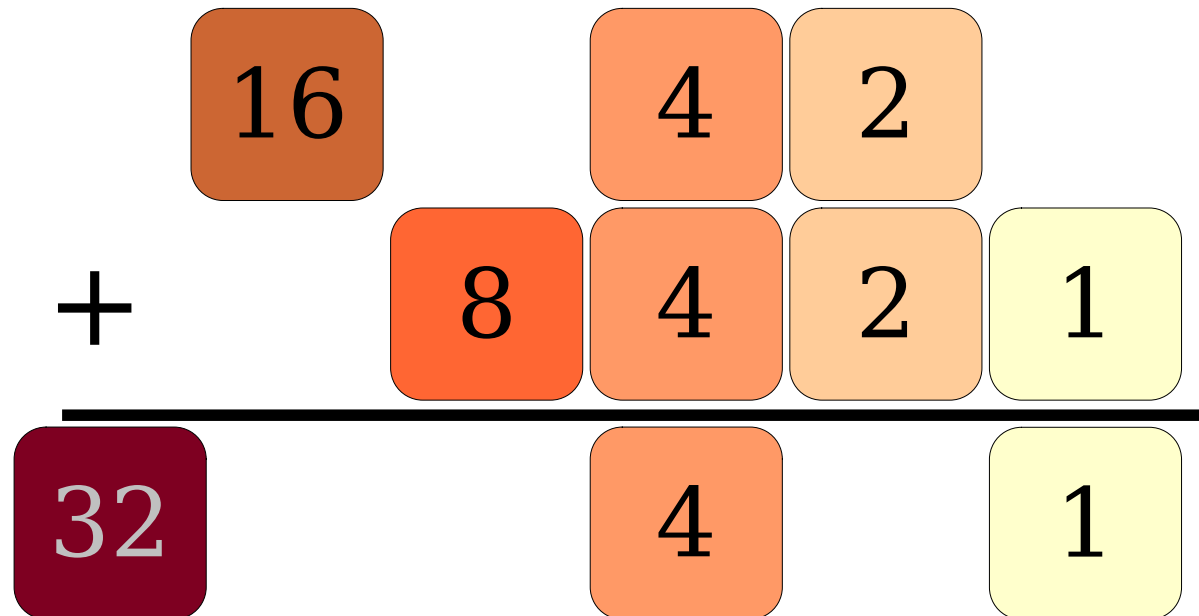
# Adding Binary Numbers

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	1		1		1		1				
		1		0		1		1		0	
+			1		1		1		1		
<hr/>											
	1		0		0		1		0		1

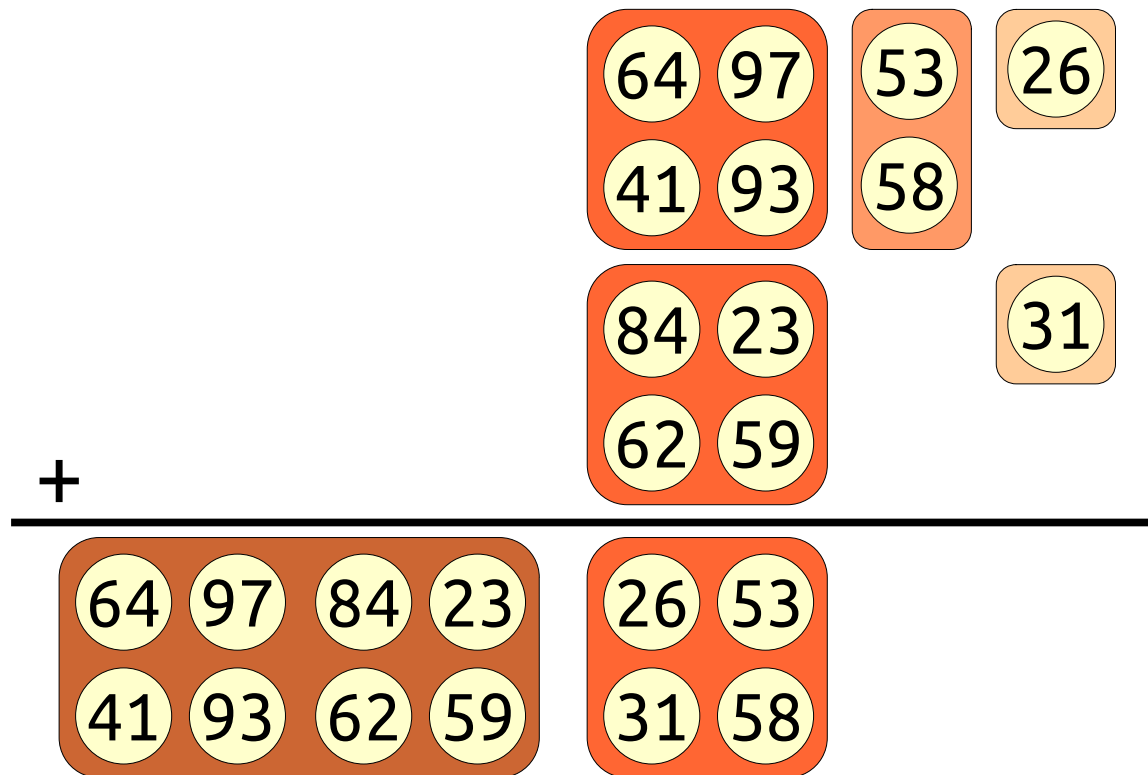
# A Different Intuition

- Represent  $n$  and  $m$  as a collection of “packets” whose sizes are powers of two.
- Adding together  $n$  and  $m$  can then be thought of as combining the packets together, eliminating duplicates



# Building a Priority Queue

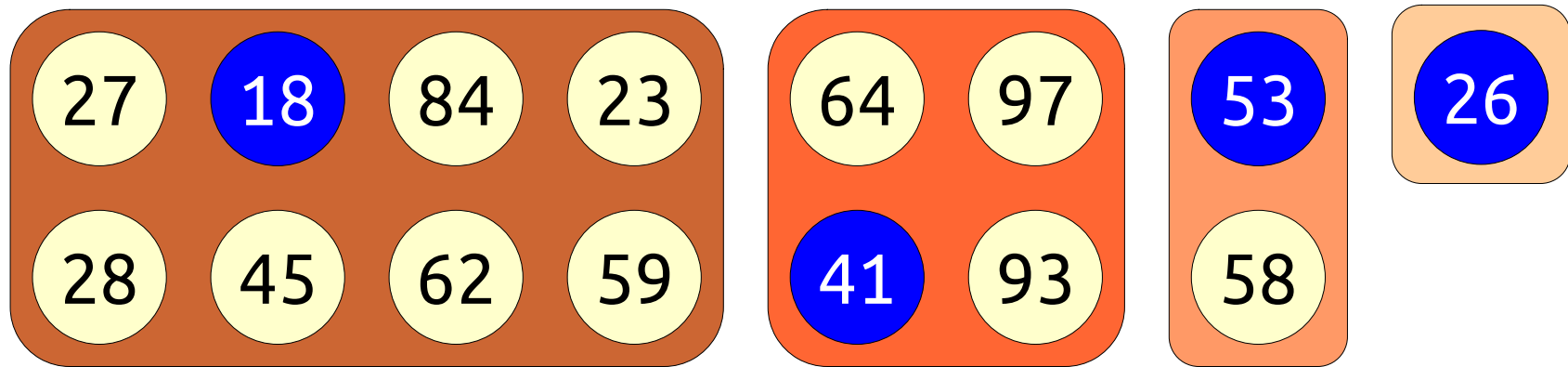
- **Idea:** Store elements in “packets” whose sizes are powers of two and *meld* by “adding” groups of packets.





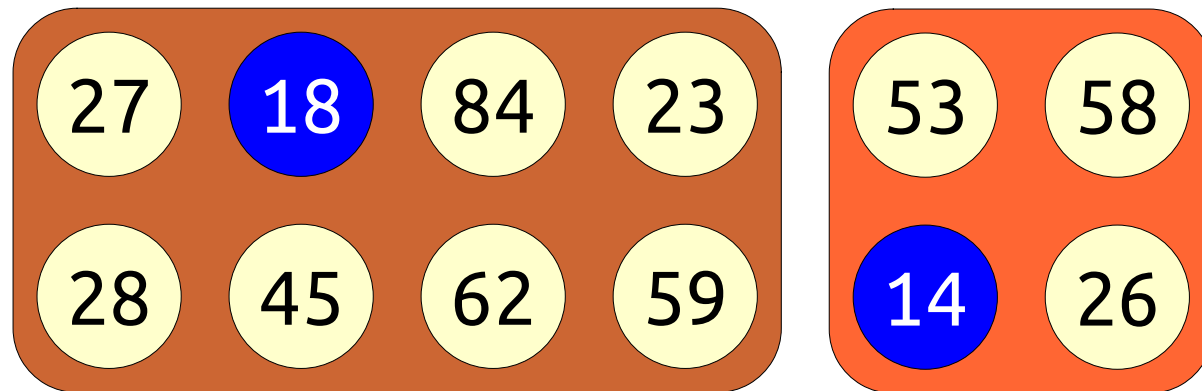
# Building a Priority Queue

- What properties must our packets have?
  - Sizes must be powers of two.
  - Can efficiently fuse packets of the same size.
  - Can efficiently find the minimum element of each packet.



# Inserting into the Queue

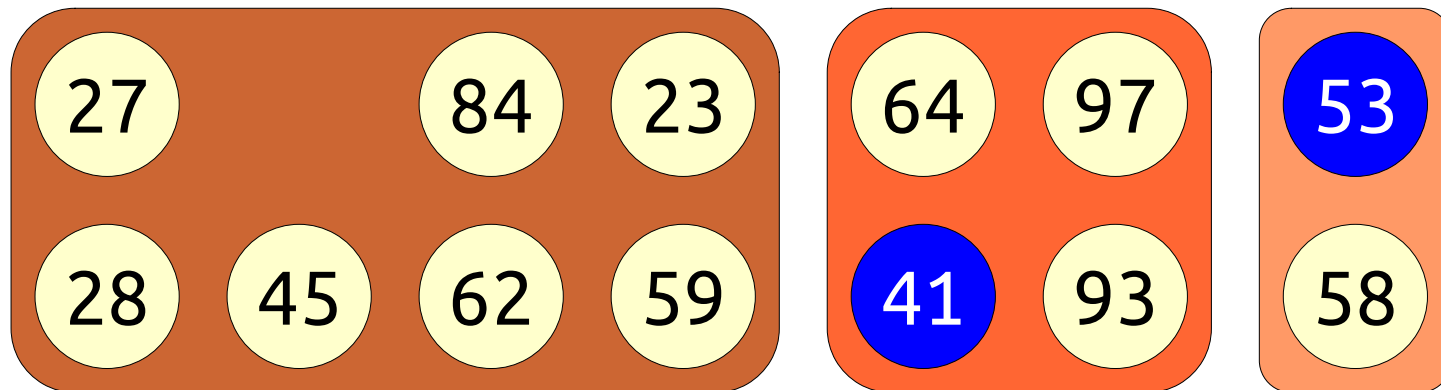
- If we can efficiently meld two priority queues, we can efficiently enqueue elements to the queue.
- **Idea:** Meld together the queue and a new queue with a single packet.



Time required:  
 $O(\log n)$  fuses.

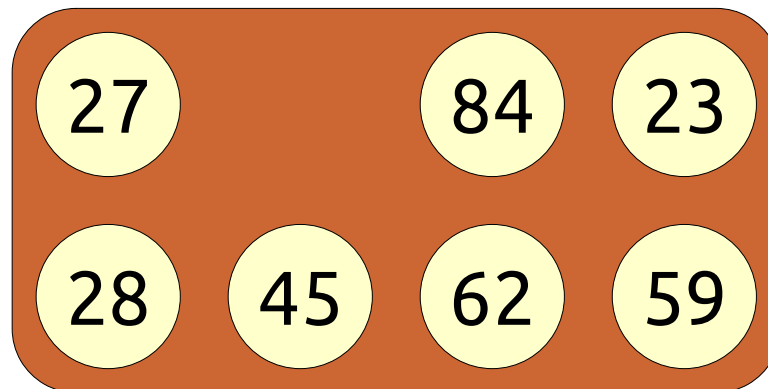
# Deleting the Minimum

- Our analogy with arithmetic breaks down when we try to remove the minimum element.
- After losing an element, the packet will not necessarily hold a number of elements that is a power of two.



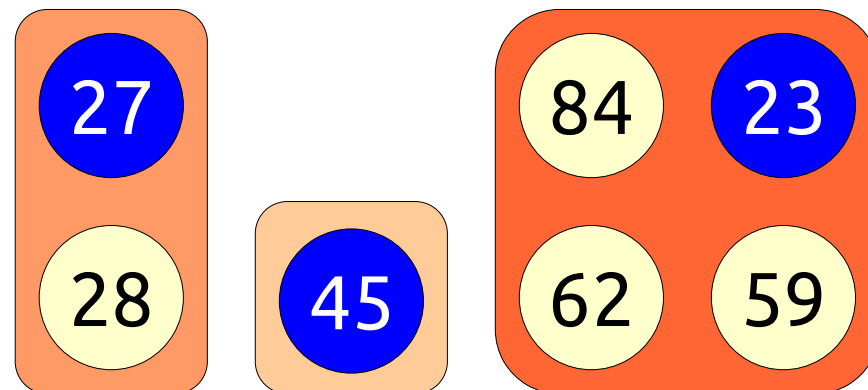
# Deleting the Minimum

- If we have a packet with  $2^k$  elements in it and remove a single element, we are left with  $2^k - 1$  remaining elements.



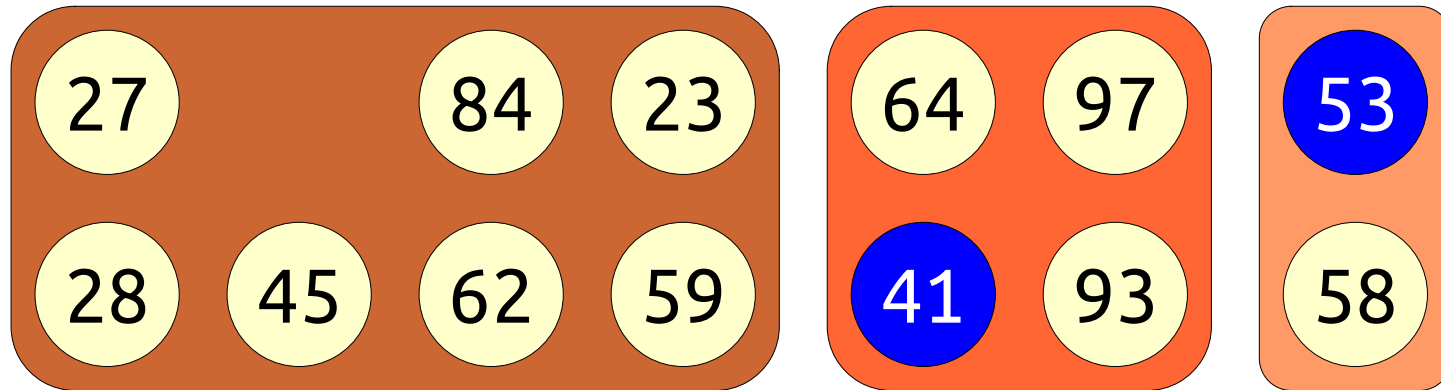
# Deleting the Minimum

- If we have a packet with  $2^k$  elements in it and remove a single element, we are left with  $2^k - 1$  remaining elements.
- **Fun fact:**  $2^k - 1 = 2^0 + 2^1 + 2^2 + \dots + 2^{k-1}$ .
- **Idea:** “Fracture” the packet into  $k$  smaller packets, then add them back in.



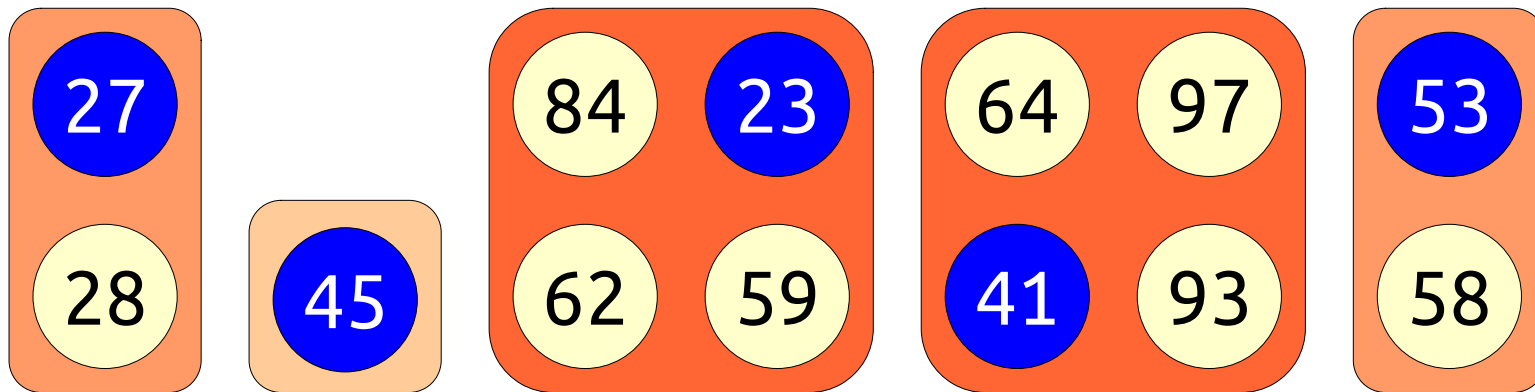
# Fracturing Packets

- We can *extract-min* by fracturing the packet containing the minimum and adding the fragments back in.



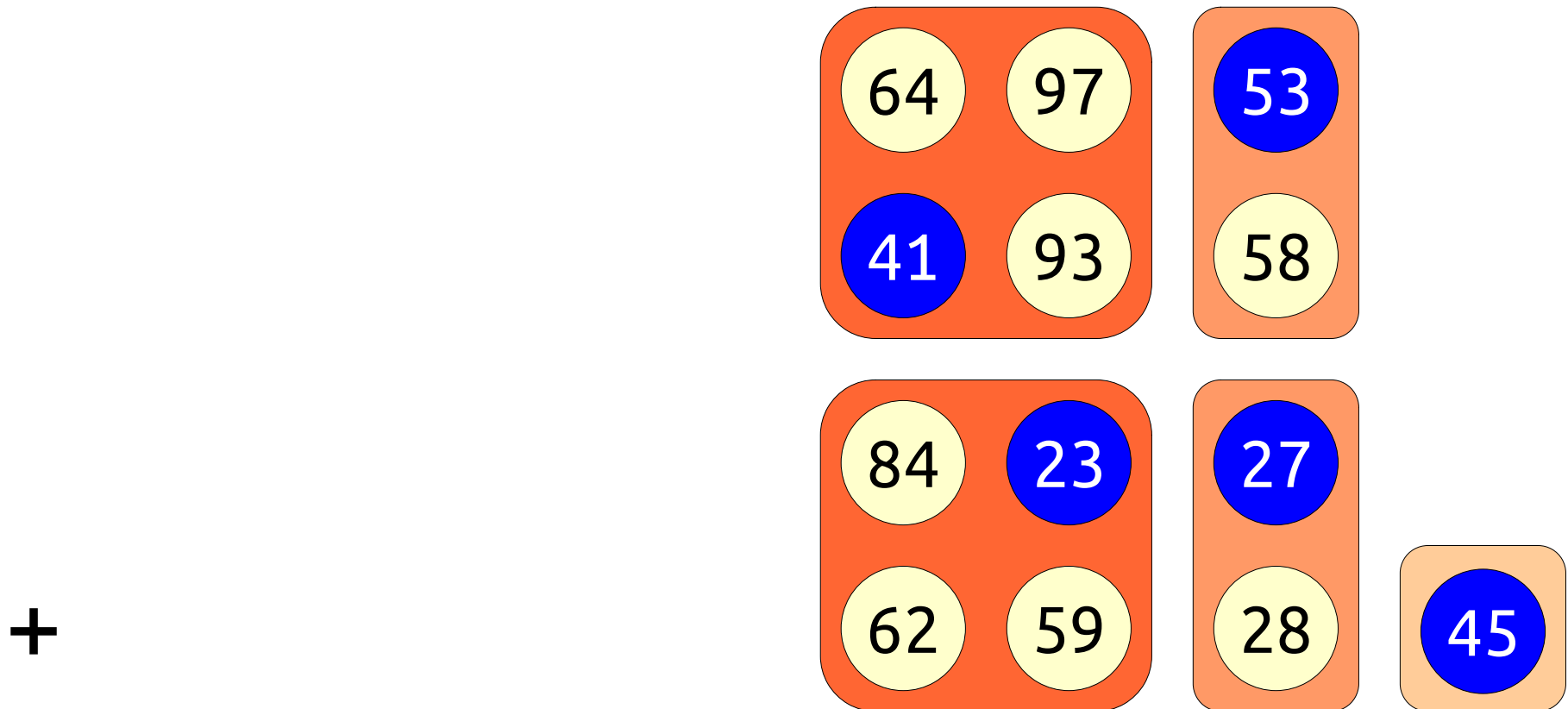
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# Fracturing Packets

- We can *extract-min* by fracturing the packet containing the minimum and adding the fragments back in.
- Runtime is  $O(\log n)$  fuses in *meld*, plus fracture cost.





# Building a Priority Queue

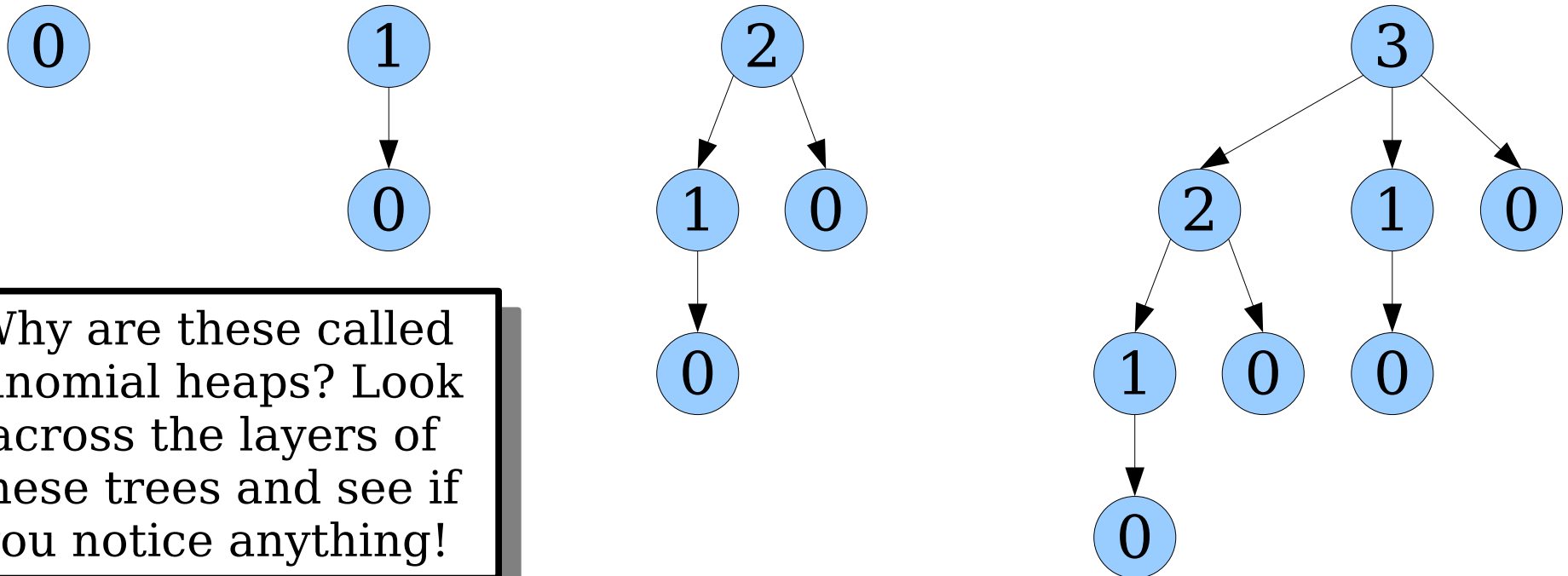
- What properties must our packets have?
  - Size is a power of two.
  - Can efficiently fuse packets of the same size.
  - Can efficiently find the minimum element of each packet.
  - Can efficiently “fracture” a packet of  $2^k$  nodes into packets of  $2^0, 2^1, 2^2, 2^3, \dots, 2^{k-1}$  nodes.
- **Question:** How can we represent our packets to support the above operations efficiently?

# Binomial Trees

- A **binomial tree of order  $k$**  is a type of tree recursively defined as follows:

*A binomial tree of order  $k$  is a single node whose children are binomial trees of order  $0, 1, 2, \dots, k - 1$ .*

- Here are the first few binomial trees:



Why are these called binomial heaps? Look across the layers of these trees and see if you notice anything!

# Binomial Trees

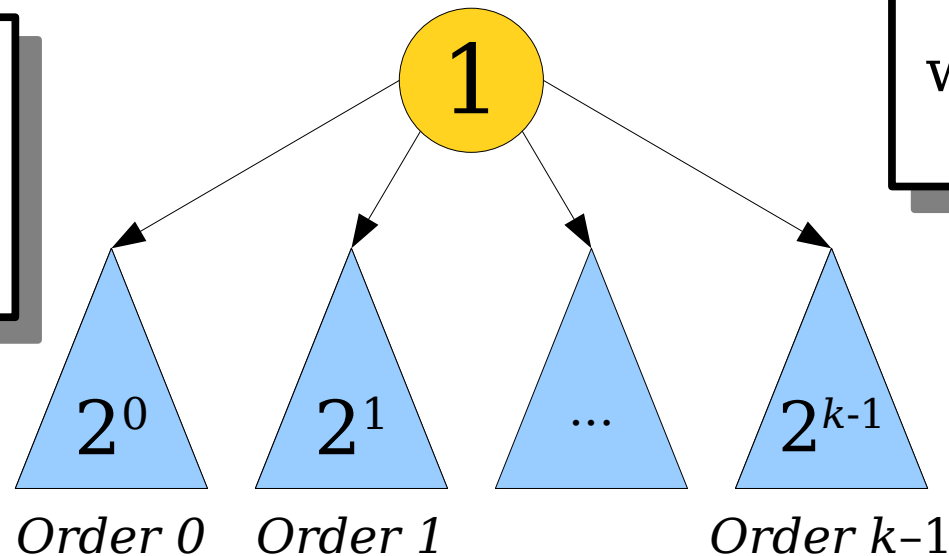
- **Theorem:** A binomial tree of order  $k$  has exactly  $2^k$  nodes.
- **Proof:** Induction on  $k$ .

Assume that binomial trees of orders  $0, 1, \dots, k - 1$  have  $2^0, 2^1, \dots, 2^{k-1}$  nodes. The number of nodes in an order- $k$  binomial tree is

$$2^0 + 2^1 + \dots + 2^{k-1} + 1 = 2^k - 1 + 1 = 2^k$$

So the claim holds for  $k$  as well. ■

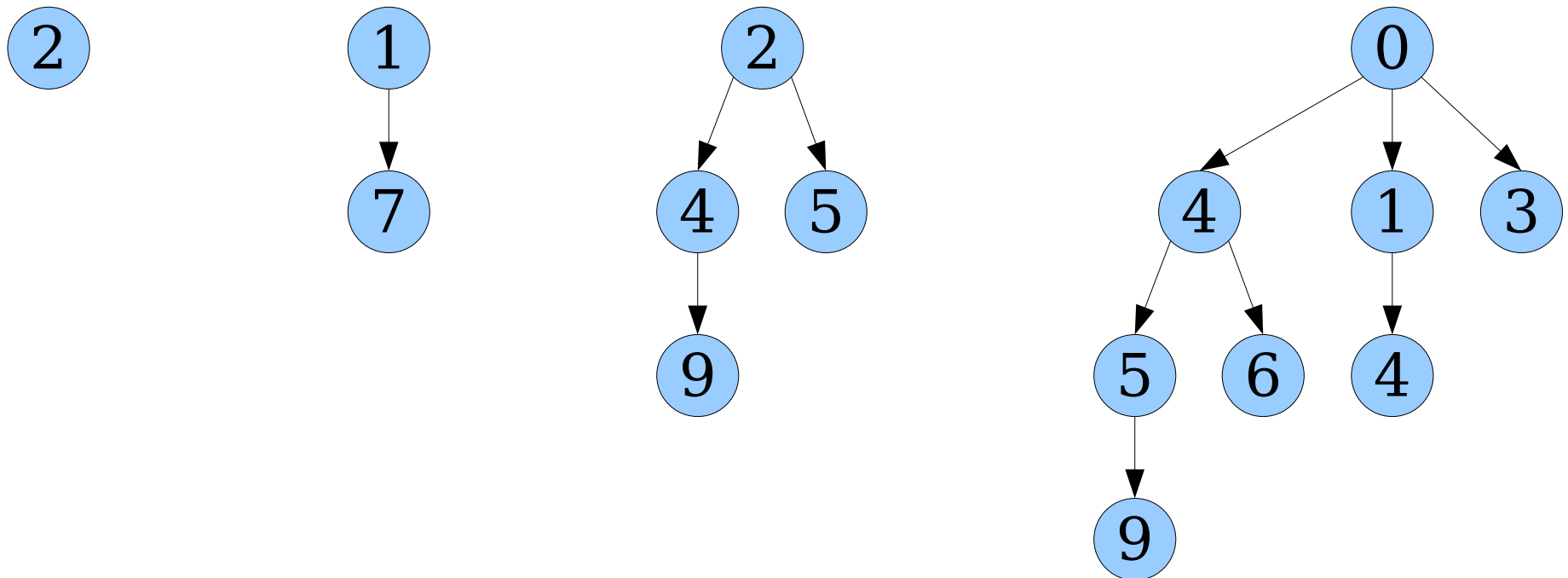
**Fun Question:**  
Why doesn't this inductive proof have a base case?



There's another way to show this. Stay tuned!

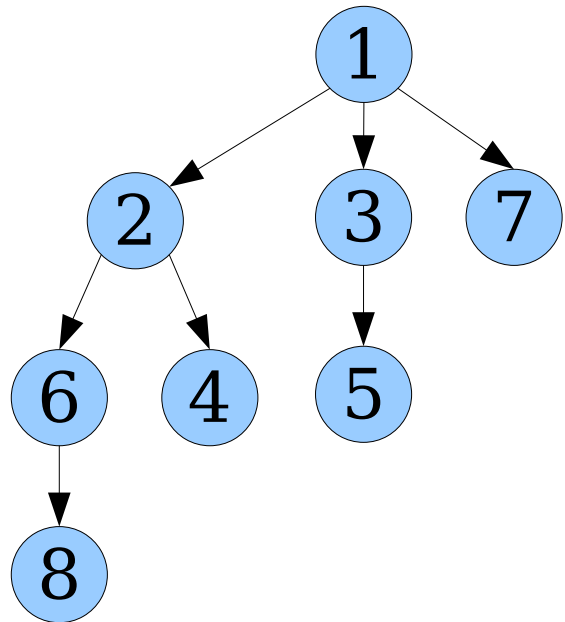
# Binomial Trees

- A **heap-ordered binomial tree** is a binomial tree whose nodes obey the heap property: all nodes are less than or equal to their descendants.
- We will use heap-ordered binomial trees to implement our “packets.”



# Binomial Trees

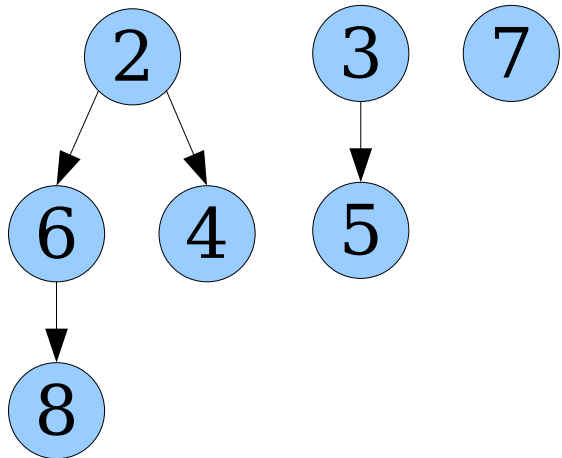
- What properties must our packets have?
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Make the binomial tree with the larger root the first child of the tree with the smaller root.

# Binomial Trees

- What properties must our packets have?
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  - Can efficiently fuse packets of the same size. ✓
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  - Can efficiently “fracture” a packet of  $2^k$  nodes into packets of  $2^0, 2^1, 2^2, 2^3, \dots, 2^{k-1}$  nodes. ✓



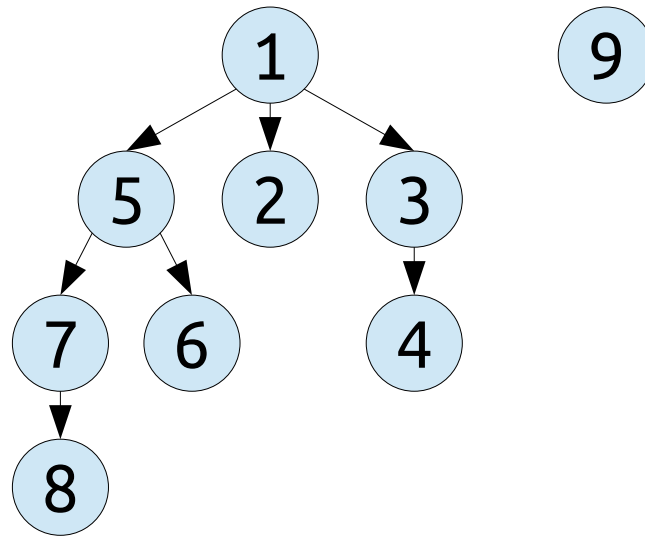
# The Binomial Heap

- A **binomial heap** is a collection of binomial trees stored in ascending order of size.
- Operations defined as follows:
  - **meld**( $pq_1, pq_2$ ): Use addition to combine all the trees.
    - Fuses  $O(\log n + \log m)$  trees. Cost:  $O(\log n + \log m)$ . Here, assume one binomial heap has  $n$  nodes, the other  $m$ .
  - $pq$ .**enqueue**( $v, k$ ): Meld  $pq$  and a singleton heap of  $(v, k)$ .
    - Total time:  $O(\log n)$ .
  - $pq$ .**find-min**(): Find the minimum of all tree roots.
    - Total time:  $O(\log n)$ .
  - $pq$ .**extract-min**(): Find the min, delete the tree root, then meld together the queue and the exposed children.
    - Total time:  $O(\log n)$ .

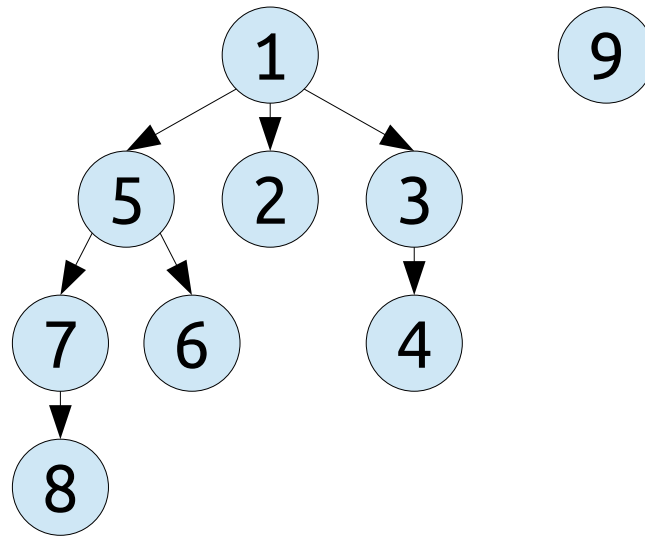
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Draw what happens if we *enqueue* the numbers 1, 2, 3, 4, 5, 6, 7, 8, and 9 into a binomial heap.

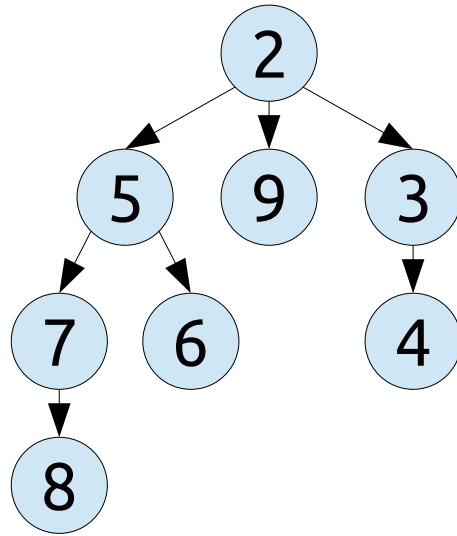




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Draw what happens after performing an *extract-min* in this binomial heap.



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Draw what happens after performing an *extract-min* in this binomial heap.

# Where We Stand

- Here's the current scorecard for the binomial heap.
- This is a fast, elegant, and clever data structure.
- **Question:** Can we do better?

## Binomial Heap

- **enqueue**:  $O(\log n)$
- **find-min**:  $O(\log n)$
- **extract-min**:  $O(\log n)$
- **meld**:  $O(\log m + \log n)$ .

# Where We Stand

- **Theorem:** No comparison-based priority queue structure can have *enqueue* and *extract-min* each take time  $o(\log n)$ .
- **Proof:** Suppose these operations each take time  $o(\log n)$ . Then we could sort  $n$  elements by perform  $n$  *enqueues* and then  $n$  *extract-mins* in time  $o(n \log n)$ . This is impossible with comparison-based algorithms. ■

## Binomial Heap

- *enqueue*:  $O(\log n)$
- *find-min*:  $O(\log n)$
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- *meld*:  $O(\log m + \log n)$ .

# Where We Stand

- We can't make both *enqueue* and *extract-min* run in time  $o(\log n)$ .
- However, we could conceivably make one of them faster.
- **Question:** Which one should we prioritize?
- Probably *enqueue*, since we aren't guaranteed to have to remove all added items.
- **Goal:** Make *enqueue* take time  $O(1)$ .

## Binomial Heap

- *enqueue*:  $O(\log n)$
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# Where We Stand

- The *enqueue* operation is implemented in terms of *meld*.
- If we want *enqueue* to run in time  $O(1)$ , we'll need *meld* to take time  $O(1)$ .
- How could we accomplish this?

## Binomial Heap

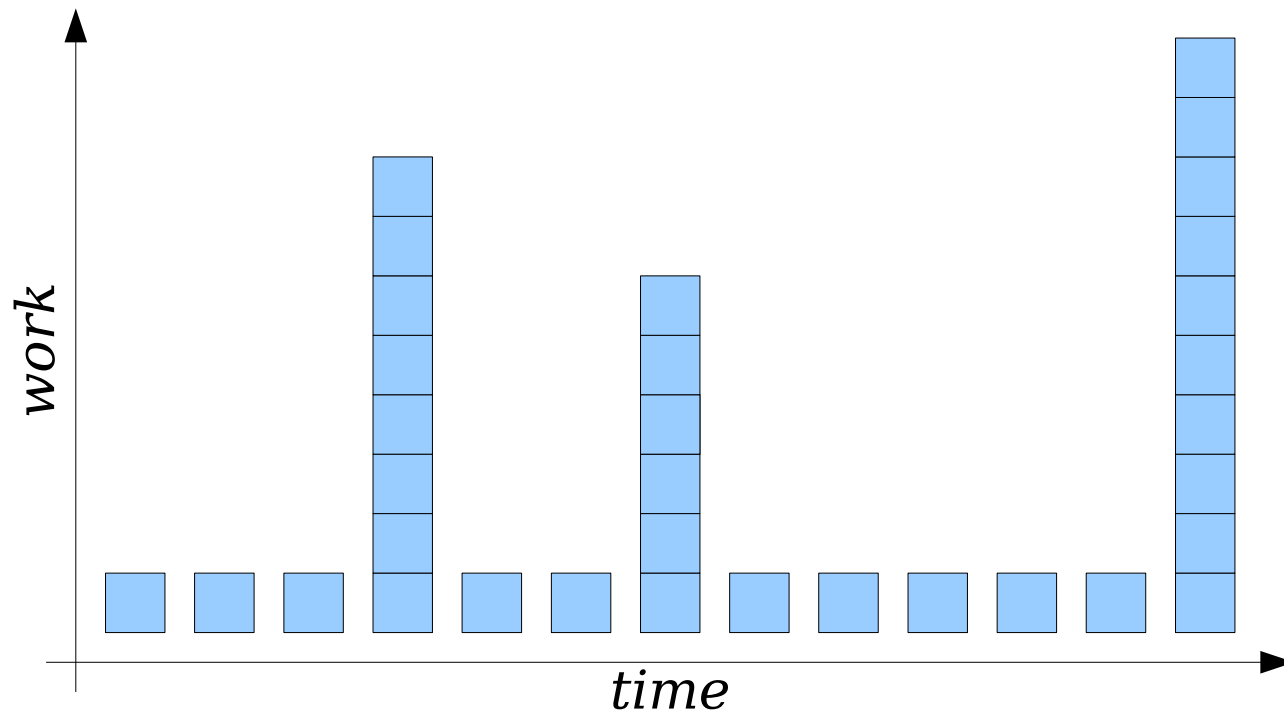
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- *meld*:  $O(\log m + \log n)$ .

# Thinking With Amortization



# Refresher: Amortization

- In an amortized efficient data structure, some operations can take much longer than others, provided that previous operations didn't take too long to finish.
- Think dishwashers: you may have to do a big cleanup at some point, but that's because you did basically no work to wash all the dishes you placed in the dishwasher.

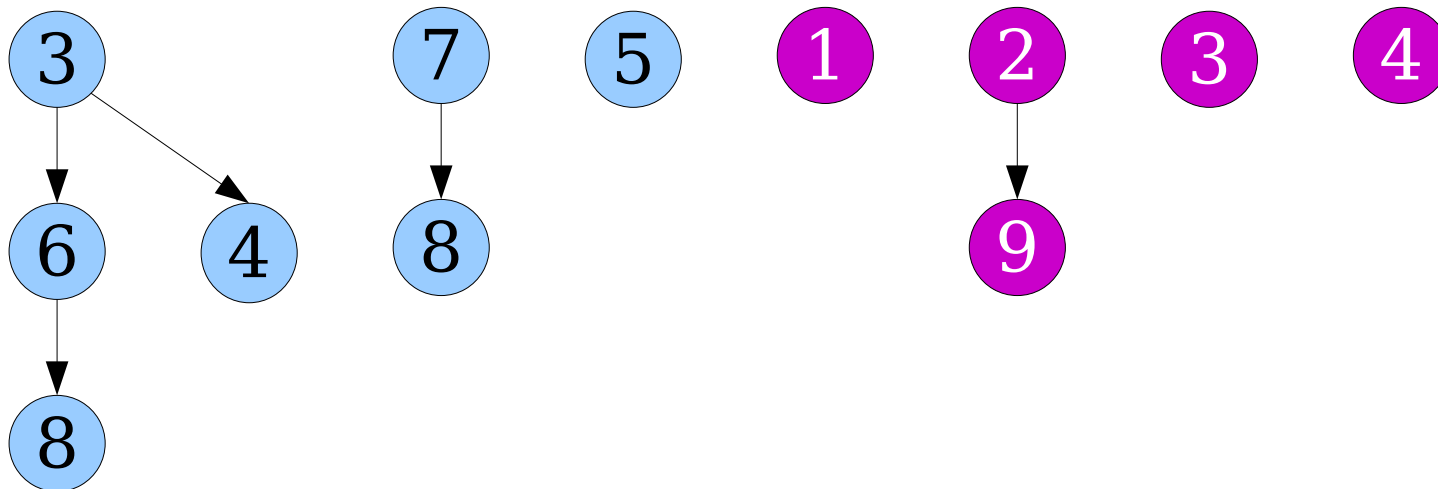


# Lazy Melding

- Consider the following lazy *melding* approach:

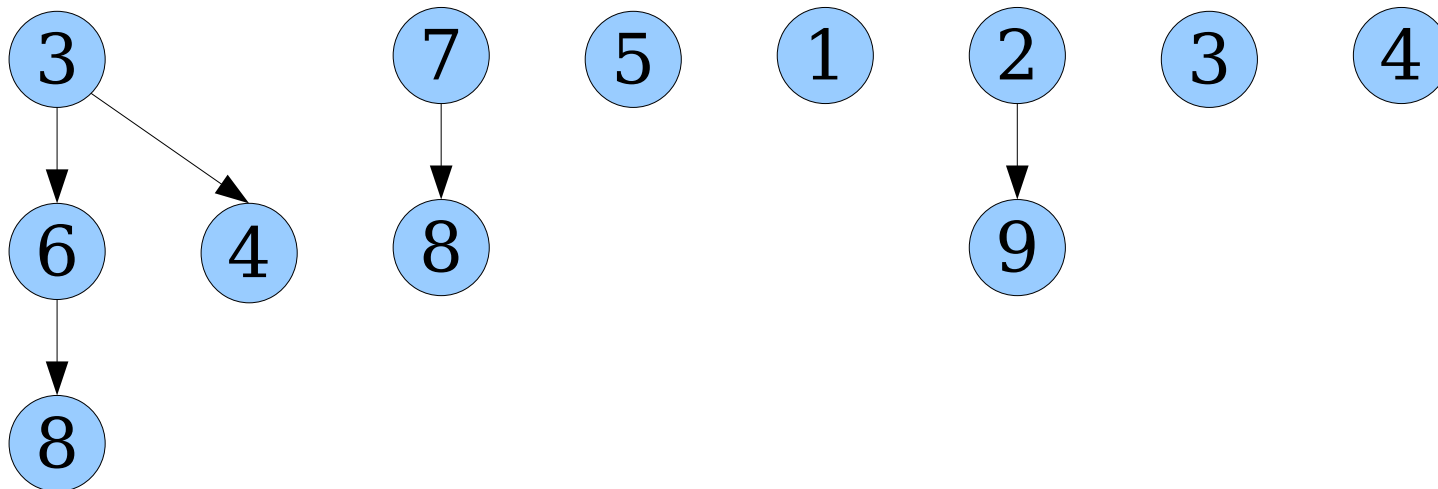
*To meld together two binomial heaps, just combine the two sets of trees together.*

- Intuition:** Why do any work to organize keys if we're not going to do an *extract-min*? We'll worry about cleanup then.



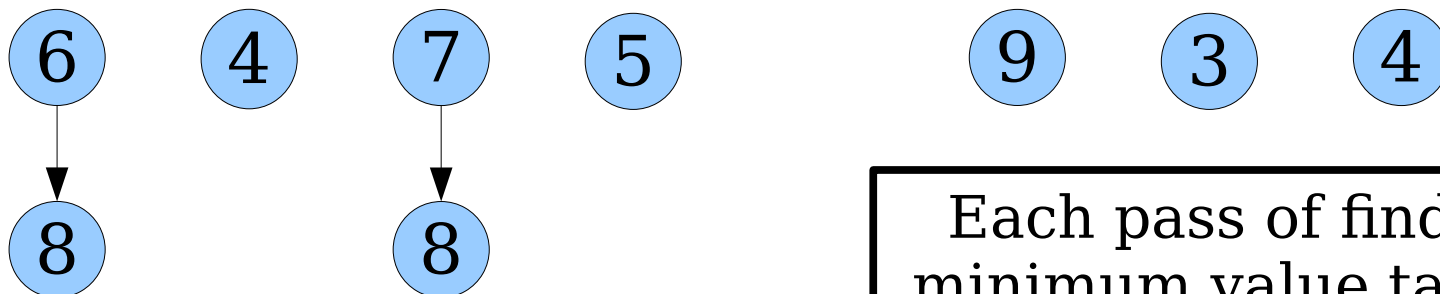
# Lazy Melding

- If we store our list of trees as circularly, doubly-linked lists, we can concatenate tree lists in time  $O(1)$ .
  - Cost of a *meld*:  **$O(1)$** .
  - Cost of an *enqueue*:  **$O(1)$** .
- If it sounds too good to be true, it probably is.



# Lazy Melding

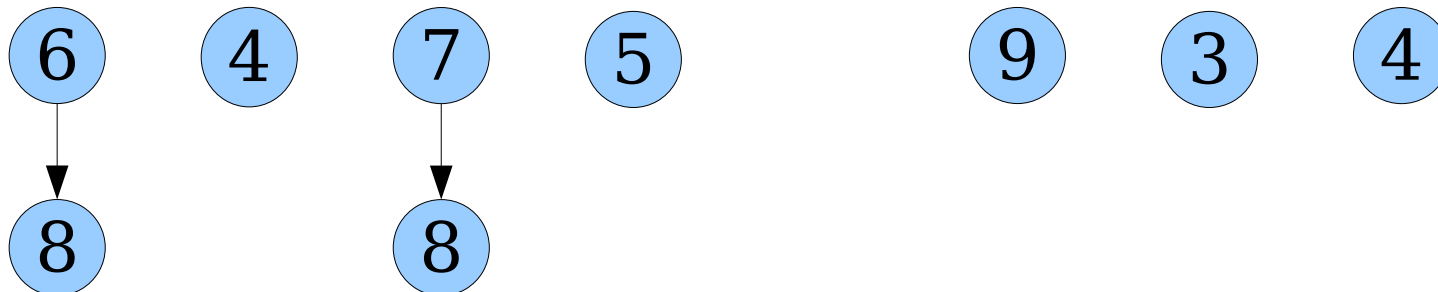
- Imagine that we implement *extract-min* the same way as before:
  - Find the packet with the minimum.
  - “Fracture” that packet to expose smaller packets.
  - Meld those packets back in with the master list.
- What happens if we do this with lazy melding?



Each pass of finding the minimum value takes time  $\Theta(n)$  in the worst case. We've lost our nice runtime guarantees!

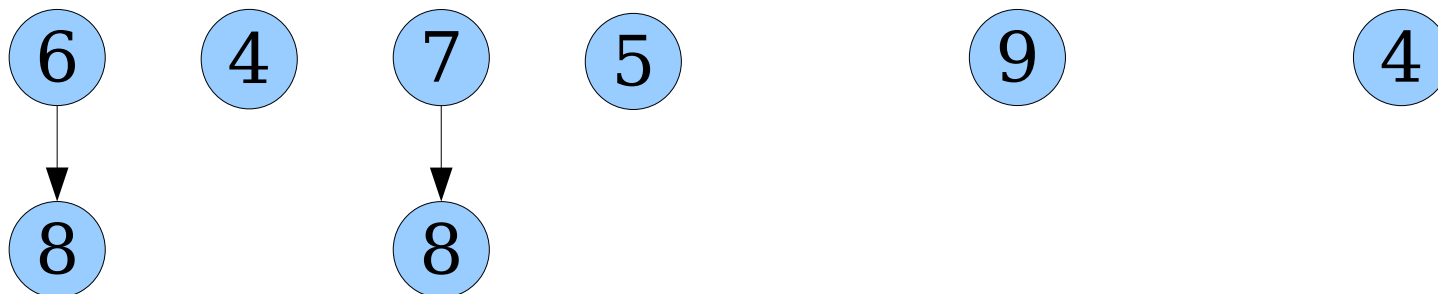
# Washing the Dishes

- Every *meld* (and *enqueue*) creates some “dirty dishes” (small trees) that we need to clean up later.
- If we never clean them up, then our *extract-min* will be too slow to be usable.
- **Idea:** Change *extract-min* to “wash the dishes” and make things look nice and pretty again.
- **Question:** What does “wash the dishes” mean here?



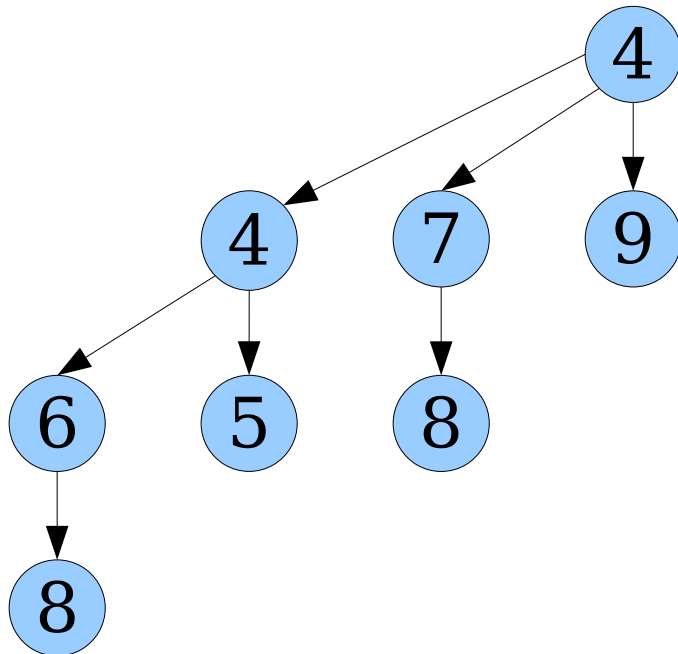
# Washing the Dishes

- With our eager *meld* (and *enqueue*) strategy, our priority queue never had more than one tree of each order.
- This kept the number of trees low, which is why each operation was so fast.
- **Idea:** After doing an *extract-min*, do a *coalesce step* to ensure there's at most one tree of each order. This gets us to where we would be if we had been doing cleanup as we go.



# Washing the Dishes

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At this point, the mess is cleaned up, and we're left with what we would have had if we had been cleaning up as we go.

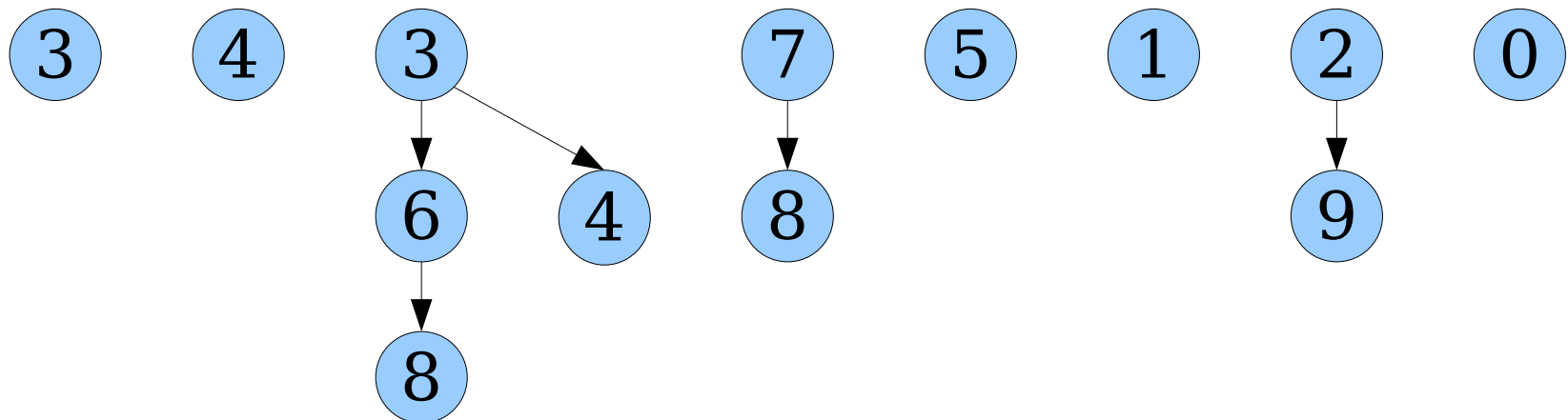
# Where We're Going

- A **lazy binomial heap** is a binomial heap, modified as follows:
  - The **meld** operation is lazy. It just combines the two groups of trees together.
  - After doing an **extract-min**, we do a **coalesce** to combine together trees until there's at most one tree of each order.
- Intuitively, we'd expect this to amortize away nicely, since the "mess" left by **meld** gets cleaned up later on by a future **extract-min**.
- Questions left to answer:
  - How do we efficiently implement the **coalesce** operation?
  - How efficient is this approach, in an amortized sense?



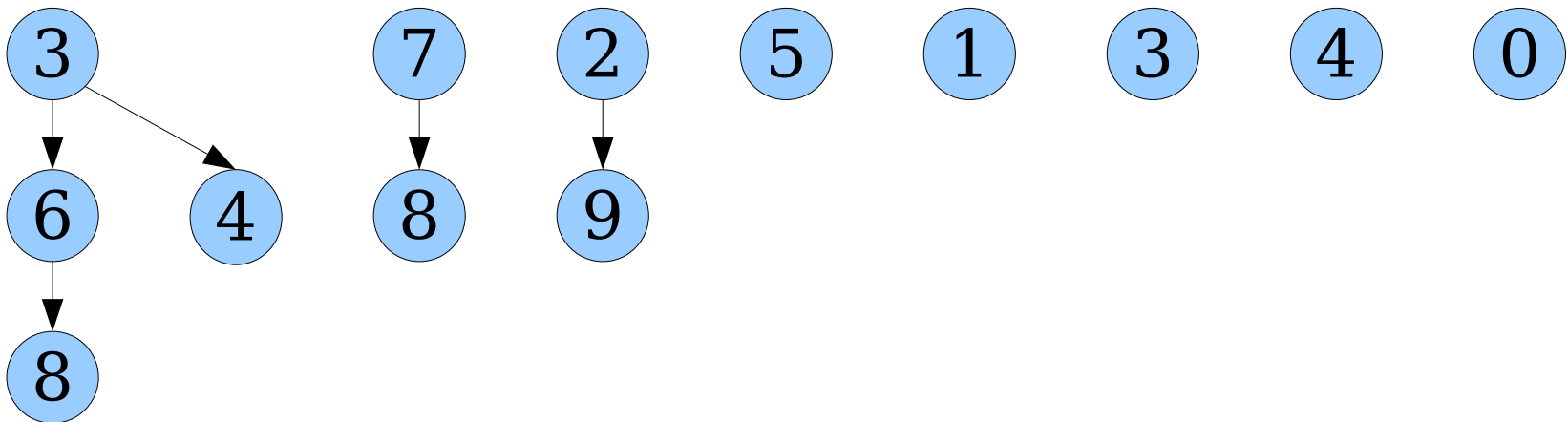
# Coalescing Trees

- The *coalesce* step repeatedly combines trees together until there's at most one tree of each order.
- How do we implement this so that it runs quickly?



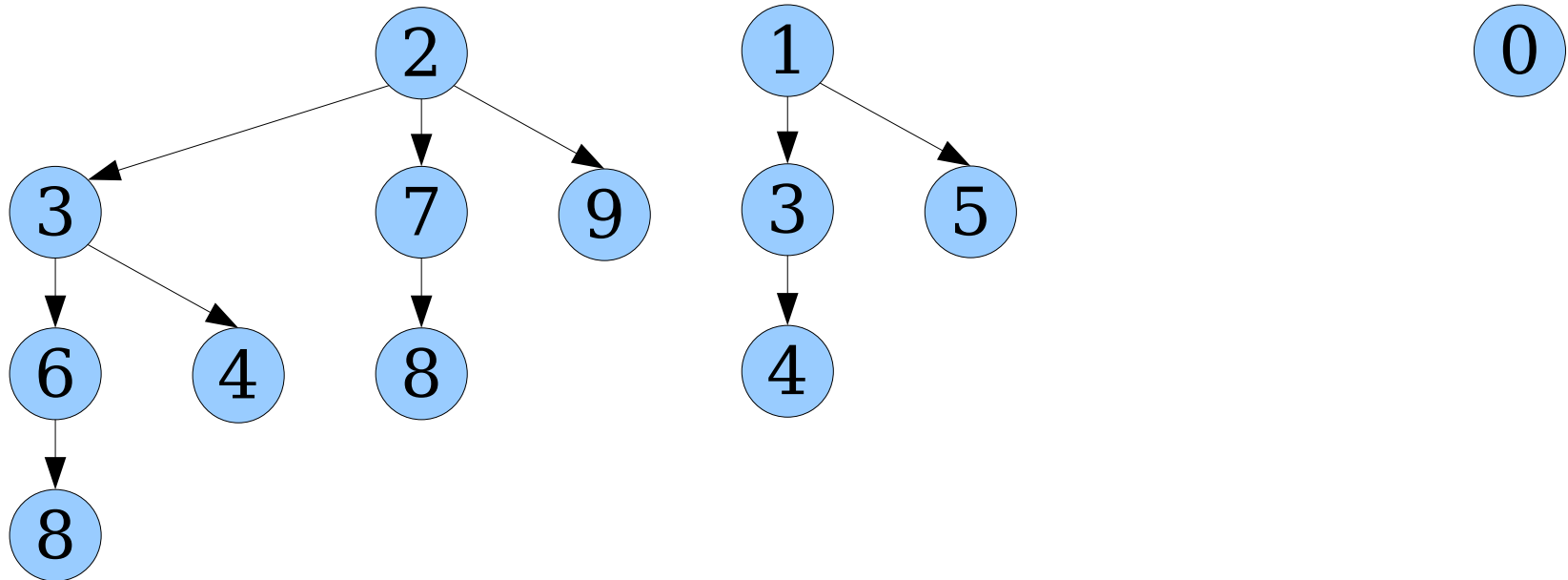
# Coalescing Trees

- **Observation:** This would be a *lot* easier to do if all the trees were sorted by size.



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# Coalescing Trees

- **Observation:** This would be a *lot* easier to do if all the trees were sorted by size.
- We can sort our group of  $t$  trees by size in time  $O(t \log t)$  using a standard sorting algorithm.
- **Better idea:** All the sizes are small integers. Use counting sort!

# Coalescing Trees

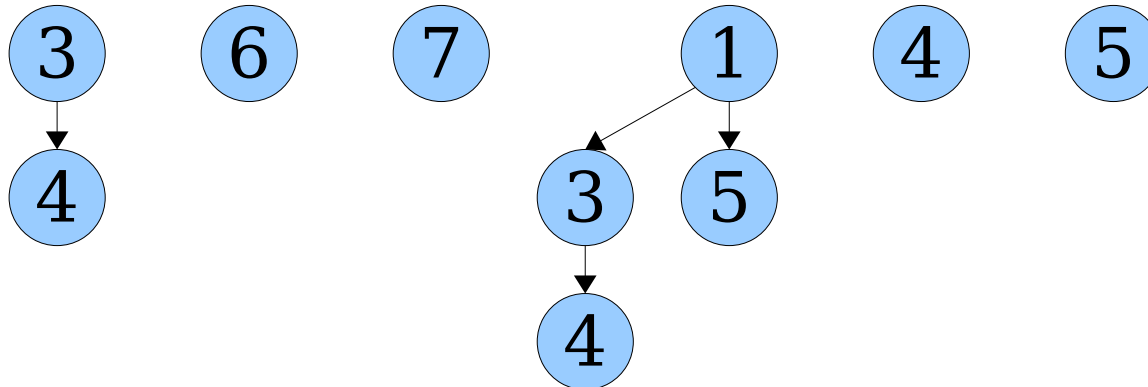
- Here is a fast implementation of *coalesce*:
  - Distribute the trees into an array of buckets big enough to hold trees of orders 0, 1, 2, ...,  $\lceil \log_2 (n + 1) \rceil$ .
  - Start at bucket 0. While there's two or more trees in the bucket, fuse them and place the result one bucket higher.

*Order 3*

*Order 2*

*Order 1*

*Order 0*



# Coalescing Trees

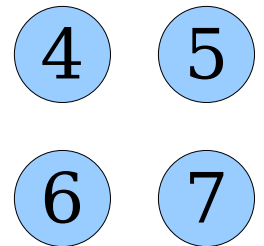
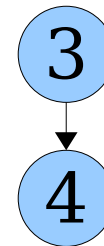
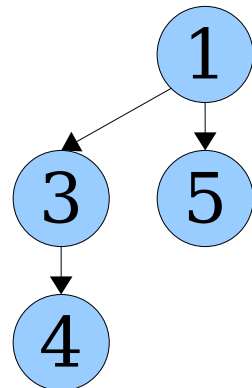
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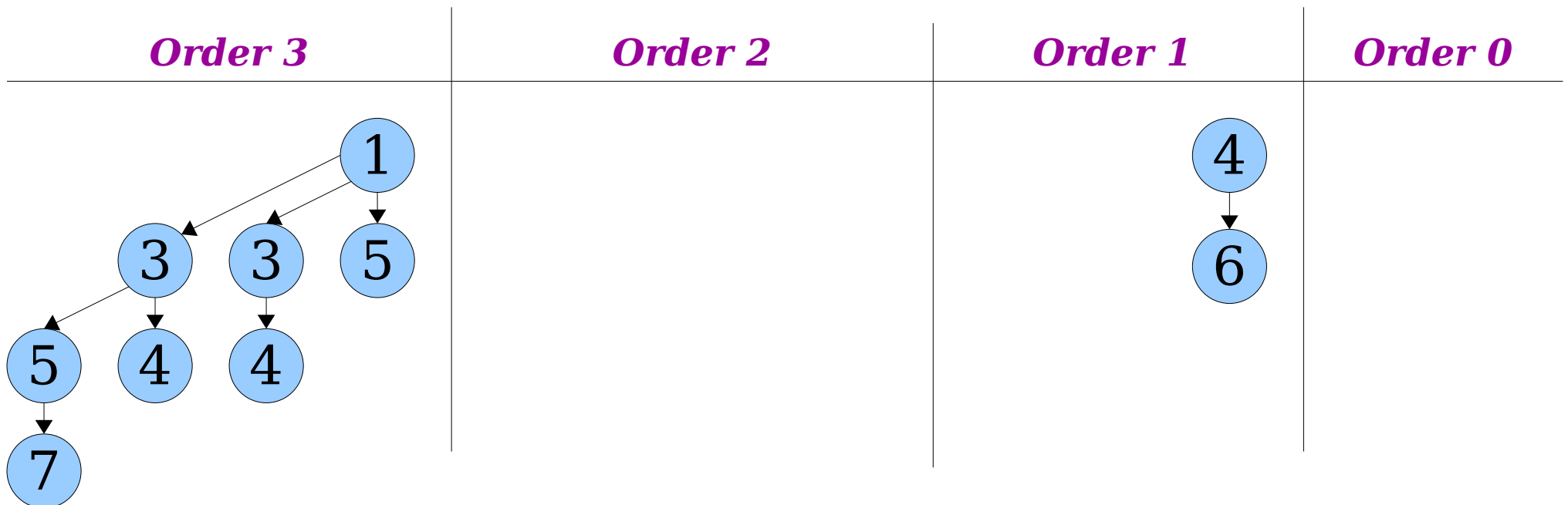
*Order 1*

*Order 0*



# Coalescing Trees

- Here is a fast implementation of *coalesce*:
  - Distribute the trees into an array of buckets big enough to hold trees of orders 0, 1, 2, ...,  $\lceil \log_2 (n + 1) \rceil$ .
  - Start at bucket 0. While there's two or more trees in the bucket, fuse them and place the result one bucket higher.



# Analyzing Coalesce

- **Claim:** Coalescing a group of  $t$  trees takes time  $O(t + \log n)$ .
  - Time to create the array of buckets:  $O(\log n)$ .
  - Time to distribute trees into buckets:  $O(t)$ .
  - Time to fuse trees:  $O(t + \log n)$ 
    - Number of fuses is  $O(t)$ , since each fuse decreases the number of trees by one. Cost per fuse is  $O(1)$ .
    - Need to iterate across  $O(\log n)$  buckets.
- Total work done:  **$O(t + \log n)$** .
- In the worst case, this is  $O(n)$ .

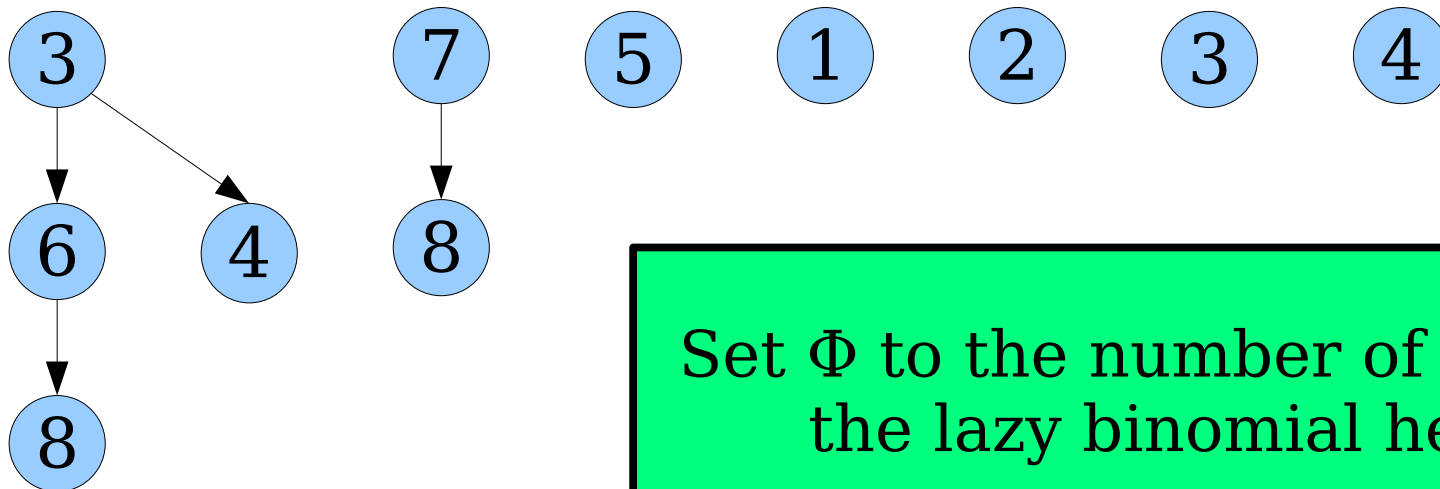


# The Story So Far

- A binomial heap with lazy melding has these worst-case time bounds:
  - ***enqueue***:  $O(1)$
  - ***meld***:  $O(1)$
  - ***find-min***:  $O(1)$
  - ***extract-min***:  $O(n)$ .
- But these are *worst-case* time bounds. Intuitively, things should nicely amortize away.
  - The number of trees grows slowly (one per ***enqueue***).
  - The number of trees drops quickly (at most one tree per order) after an ***extract-min***).

# An Amortized Analysis

- This is a great spot to use an amortized analysis by defining a potential function  $\Phi$ .
- In each case, the idea is to clearly mark what “messes” we need to clean up.
- In our case, each tree is a “mess,” since our future *coalesce* operation has to clean it up.



Set  $\Phi$  to the number of trees in the lazy binomial heap.

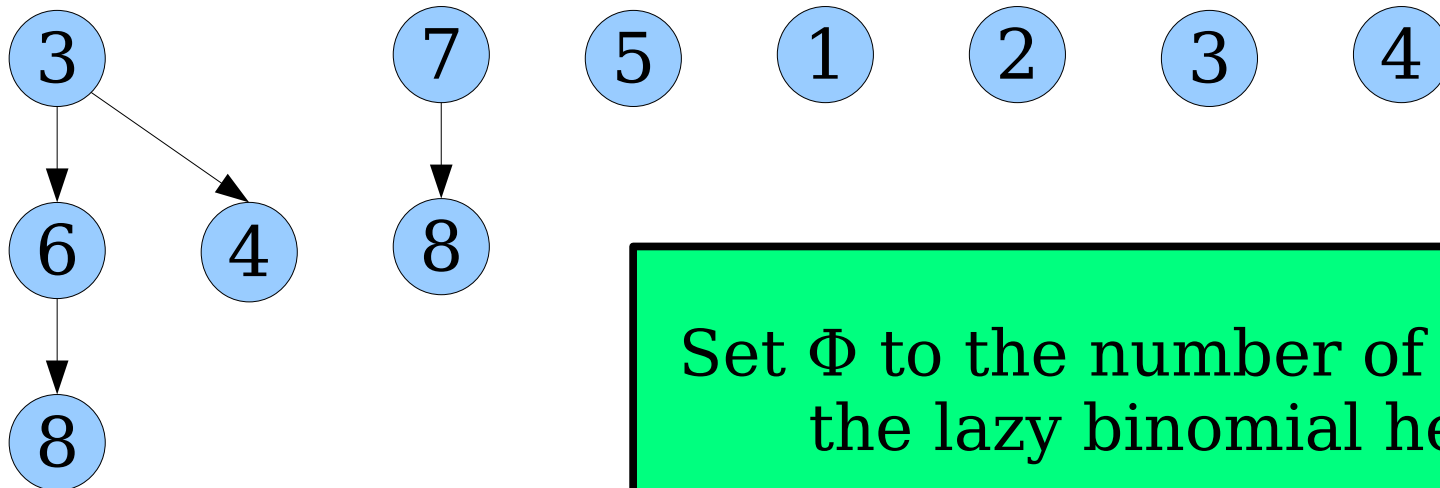
# An Amortized Analysis

- **Recall:** We assign amortized costs as

$$\text{amortized-cost} = \text{real-cost} + O(1) \cdot \Delta\Phi,$$

where  $\Delta\Phi = \Phi_{\text{after}} - \Phi_{\text{before}}$ .

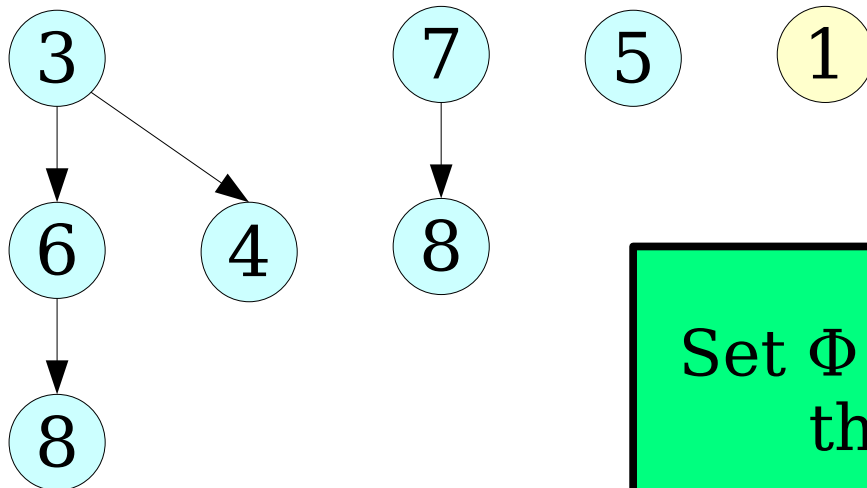
- Increasing  $\Phi$  (adding more trees) artificially boosts costs.
- Decreasing  $\Phi$  (removing trees) artificially lowers costs.
- Let's work out the amortized costs of each operation on a lazy binomial heap.



Set  $\Phi$  to the number of trees in the lazy binomial heap.

# Analyzing an Insertion

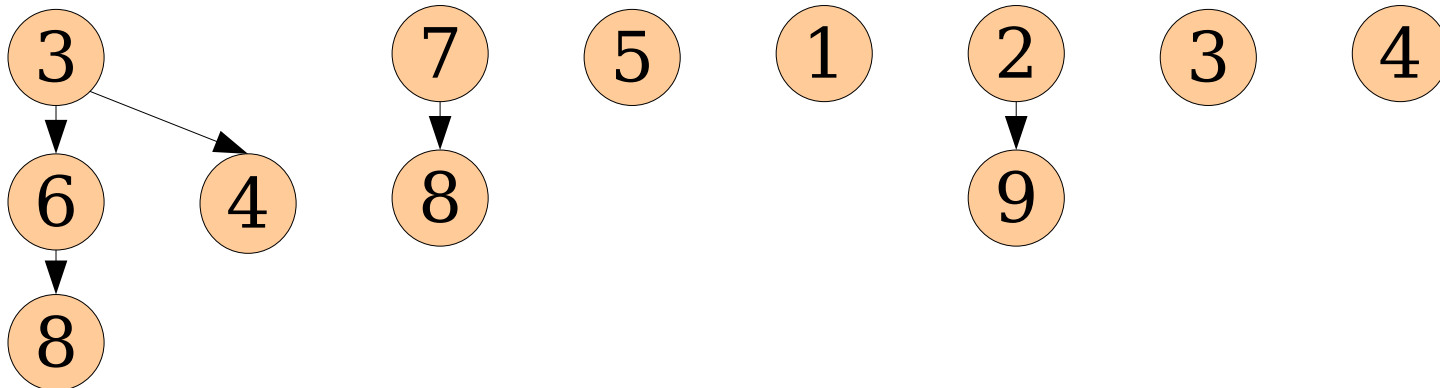
- To *enqueue* a key, we add a new binomial tree to the forest.
- Actual time:  $O(1)$ .  $\Delta\Phi$ :  $+1$
- Amortized cost:  **$O(1)$** .



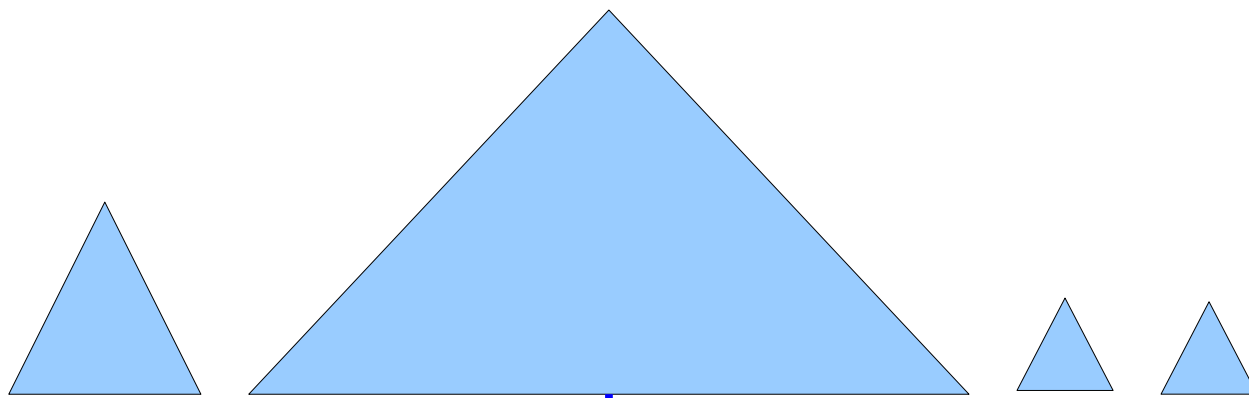
Set  $\Phi$  to the number of trees in the lazy binomial heap.

# Analyzing a Meld

- Suppose that we *meld* two lazy binomial heaps  $B_1$  and  $B_2$ . Actual cost:  $O(1)$ .
- We have the same number of trees before and after we do this, so  $\Delta\Phi = 0$ .
- Amortized cost:  **$O(1)$** .

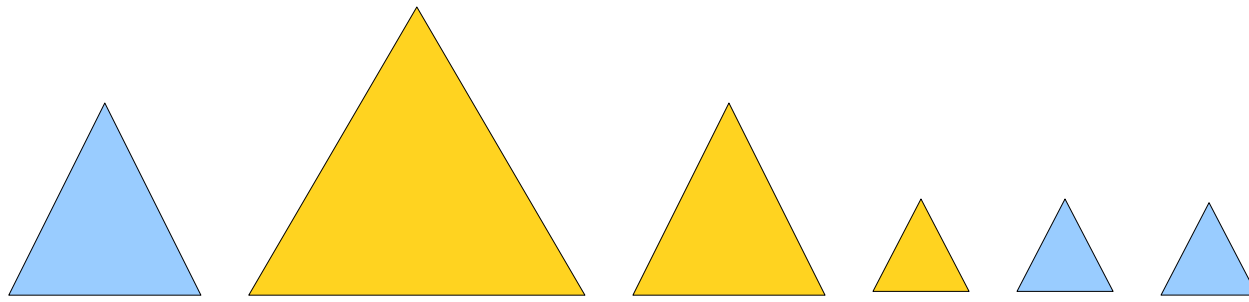


Analyzing *extract-min*



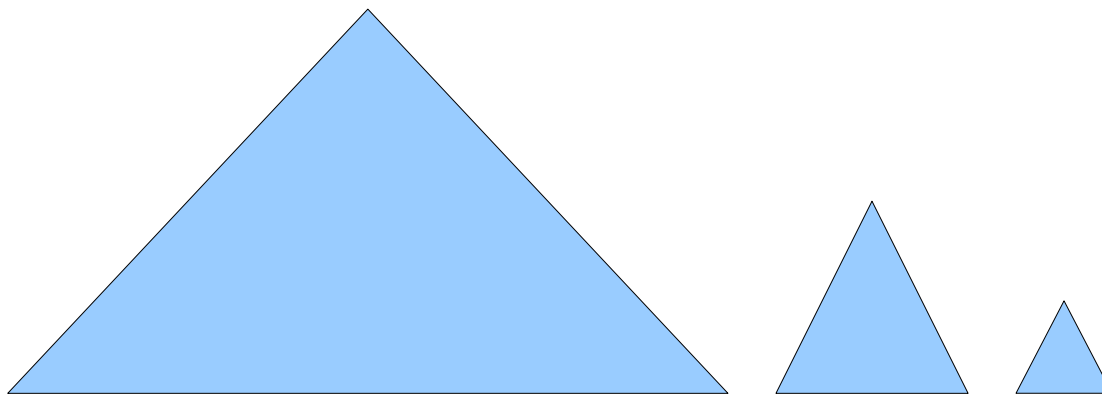
*Find tree with minimum key.*

Work:  $O(t)$   
 $\Phi = t$



*Remove min.  
 Add children to list of trees.*

Work:  $O(\log n)$

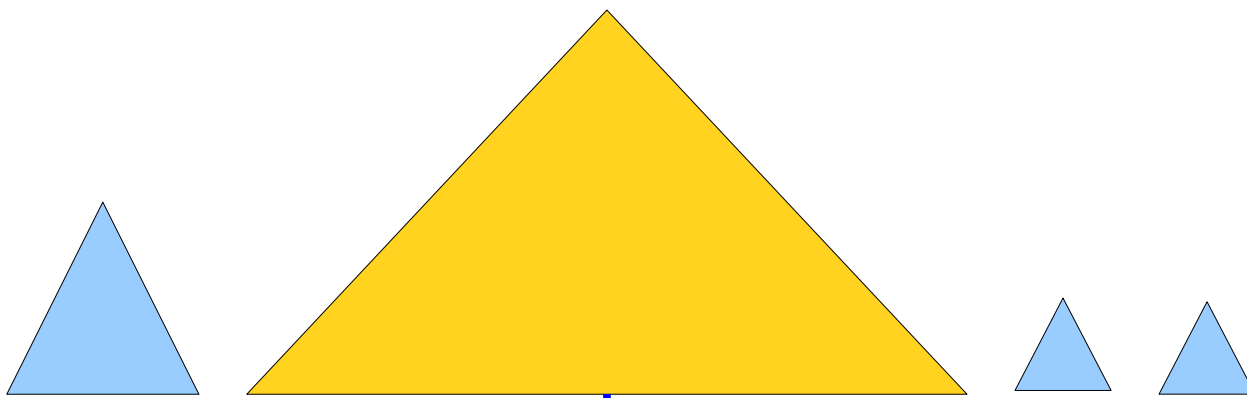


*Run the coalesce algorithm.*

Work:  $O(t + \log n)$   
 $\Phi = O(\log n)$

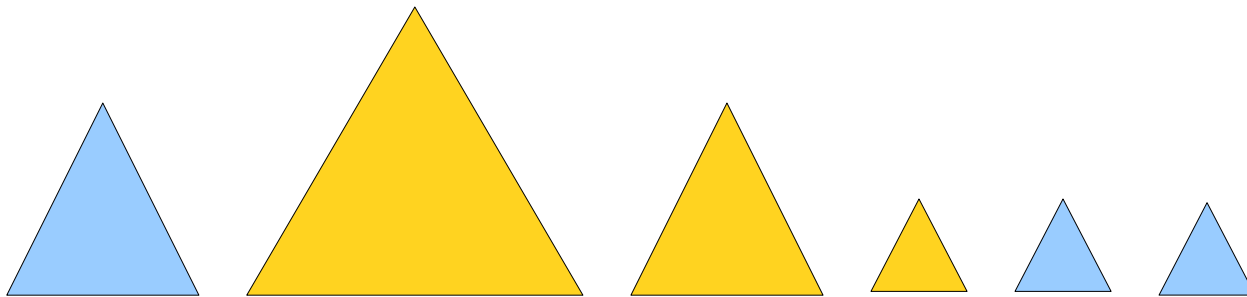
Work:  $O(t + \log n)$

$\Delta\Phi: O(-t + \log n)$



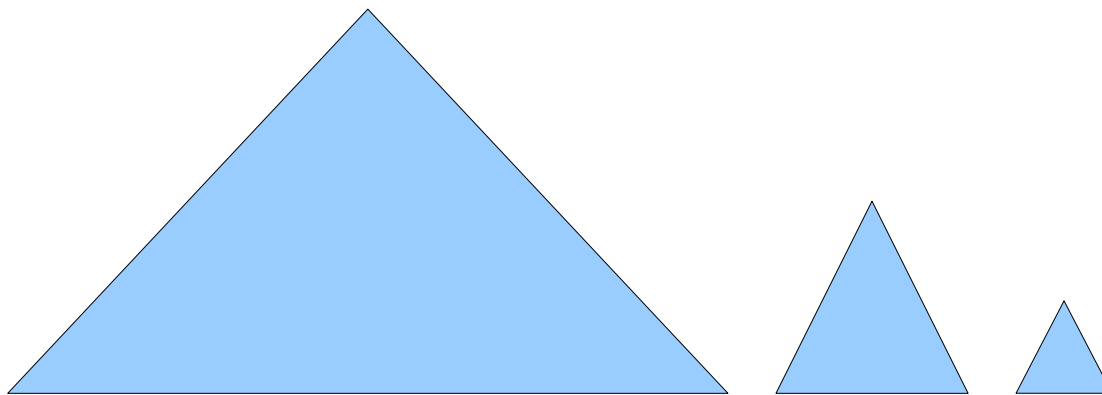
*Find tree with minimum key.*

Work:  $O(t)$   
 $\Phi = t$



*Remove min.  
 Add children to list of trees.*

Work:  $O(\log n)$



*Run the coalesce algorithm.*

Work:  $O(t + \log n)$   
 $\Phi = O(\log n)$

Amortized cost:  **$O(\log n)$** .



# Analyzing Extract-Min

- Suppose we perform an *extract-min* on a binomial heap with  $t$  trees in it.
- Initially, we expose the children of the minimum element. This increases the number of trees to  $t + O(\log n)$ .
- The runtime for coalescing these trees is  $O(t + \log n)$ .
- When we're done merging, there will be  $O(\log n)$  trees remaining, so  $\Delta\Phi = -t + O(\log n)$ .
- Amortized cost is

$$\begin{aligned} & O(t + \log n) + O(1) \cdot (-t + O(\log n)) \\ &= O(t) - O(1) \cdot t + O(1) \cdot O(\log n) \\ &= O(\log n). \end{aligned}$$

# The Final Scorecard

- Here's the final scorecard for our lazy binomial heap.
- These are *great* runtimes! We can't improve upon this except by making ***extract-min*** worst-case efficient.
  - This is possible! Check out ***bootstrapped skew binomial heaps*** or ***strict Fibonacci heaps*** for details!

## Lazy Binomial Heap

- ***Insert***:  $O(1)$
- ***Find-Min***:  $O(1)$
- ***Extract-Min***:  $O(\log n)^*$
- ***Meld***:  $O(1)$

\* *amortized*

# Major Ideas from Today

- Isometries are a *great* way to design data structures.
  - Here, binomial heaps come from binary arithmetic.
- Designing for amortized efficiency is about building up messes slowly and rapidly cleaning them up.
  - Each individual *enqueue* isn't too bad, and a single *extract-min* fixes all the prior problems.

# Next Time

- ***The Need for decrease-key***
  - A powerful and versatile operation on priority queues.
- ***Fibonacci Heaps***
  - A variation on lazy binomial heaps with efficient decrease-key.
- ***Implementing Fibonacci Heaps***
  - ... is harder than it looks!