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 more advanced data structures.

• Fibonacci Heaps (Tuesday)

• A famous and theoretically excellent priority queue.

Review: 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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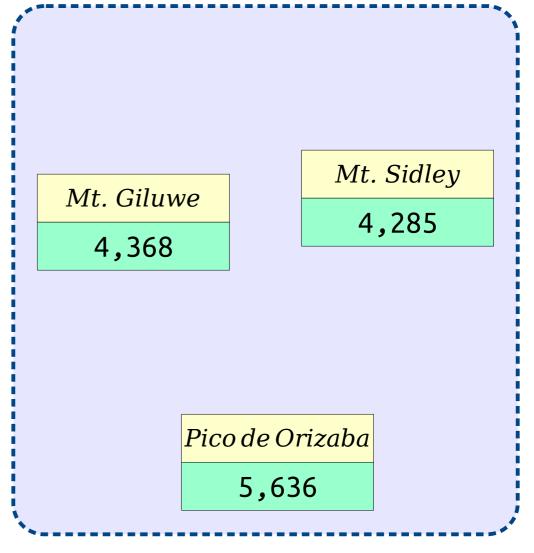
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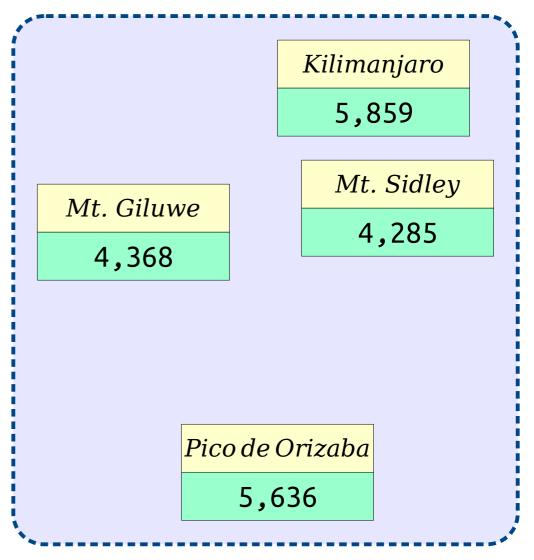
Pico de Orizaba

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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 welltuned value of d, and otherwise only the sequence heap is known to specifically outperform this family.
 - (Is this information incorrect as of 2023? 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 Δk 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-toall on the next problem set.

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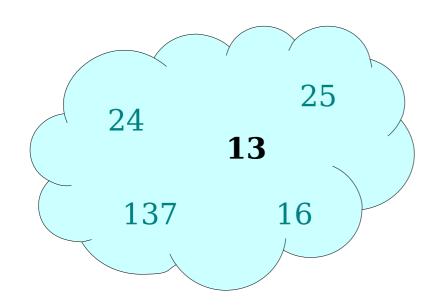
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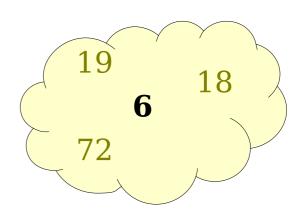
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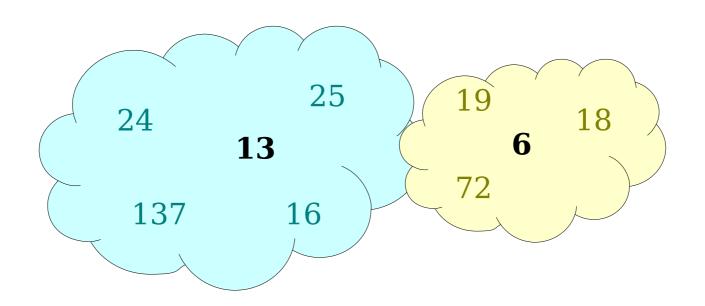
You'll design a priority queue supporting *meld* and *add-to-all* on the next problem set.

- 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 .

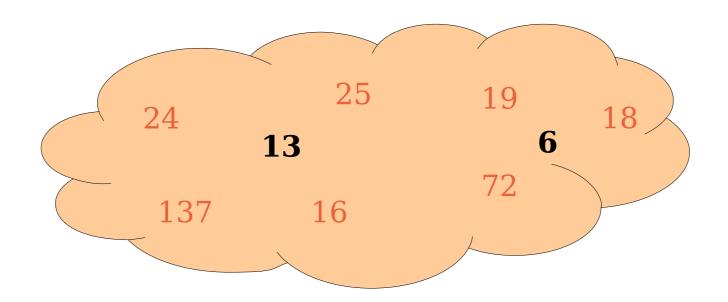




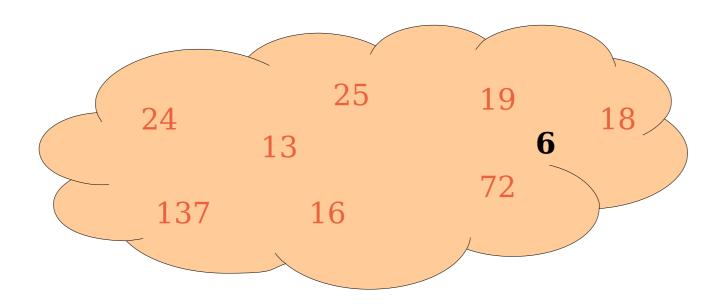
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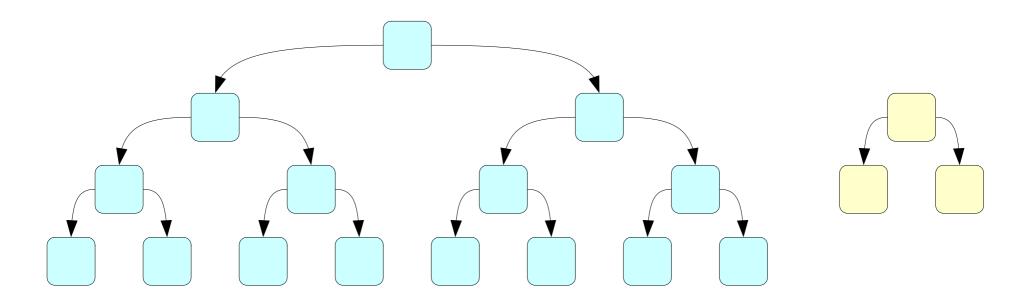


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Efficiently Meldable Queues

- 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?

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

Intuition:

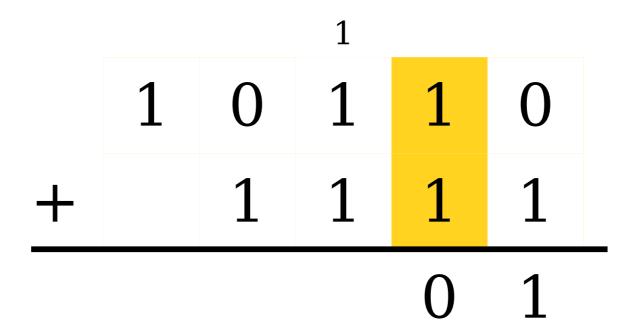
Writing out n in any "reasonable" base requires $\Theta(\log n)$ digits.

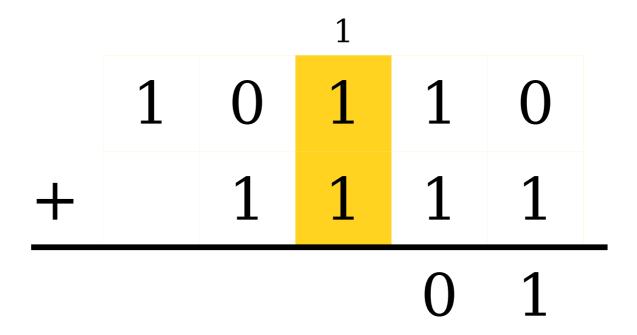
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+		1	1	1	1	

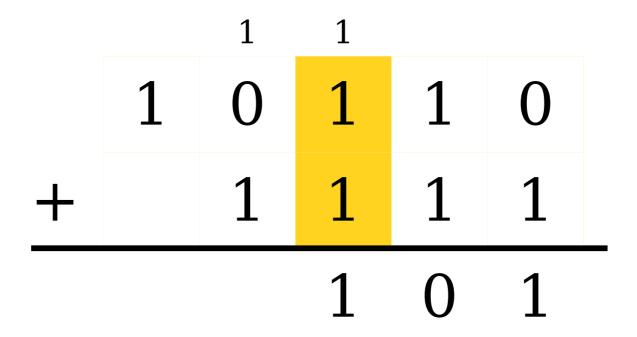
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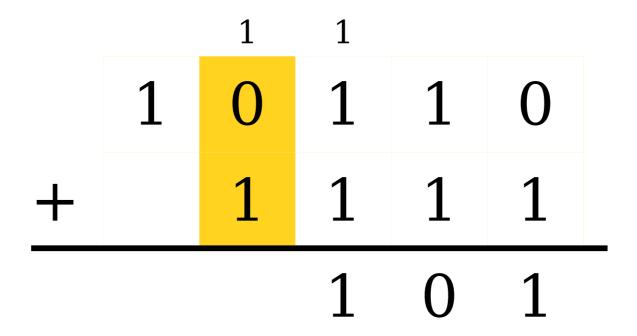
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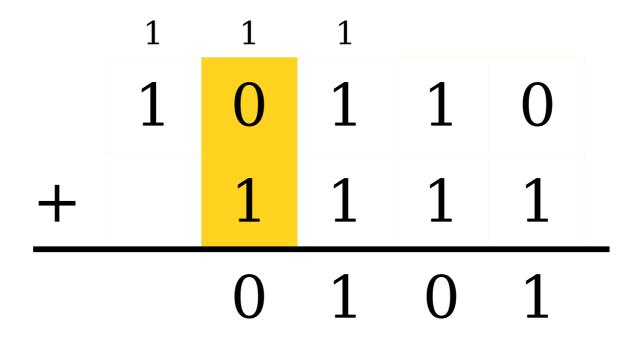
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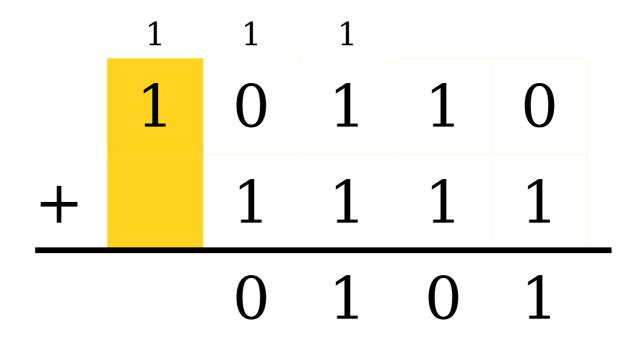


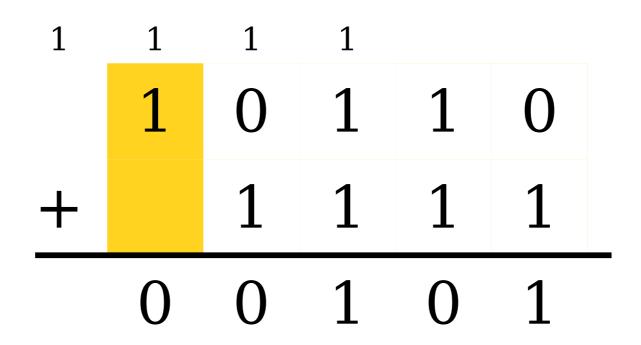












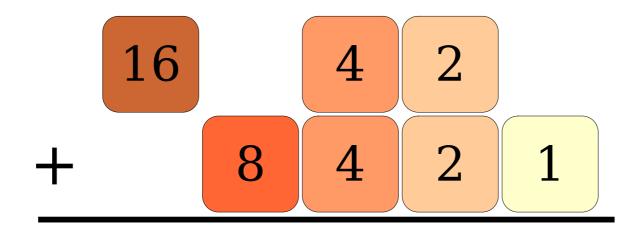
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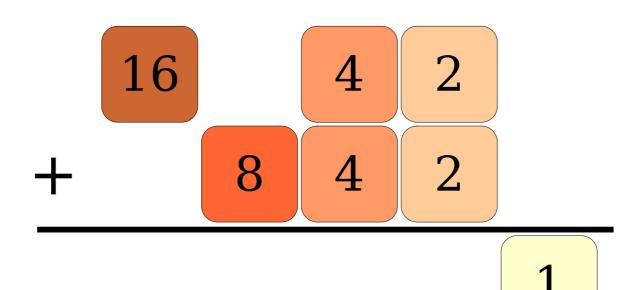
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- Adding together n and m can then be thought of as combining the packets together, eliminating duplicates

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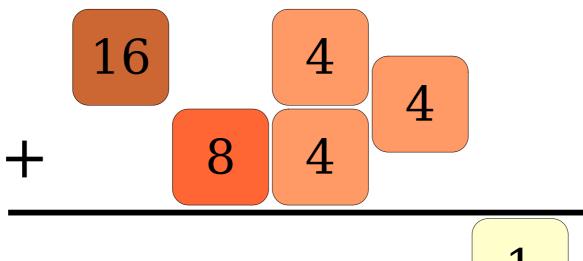
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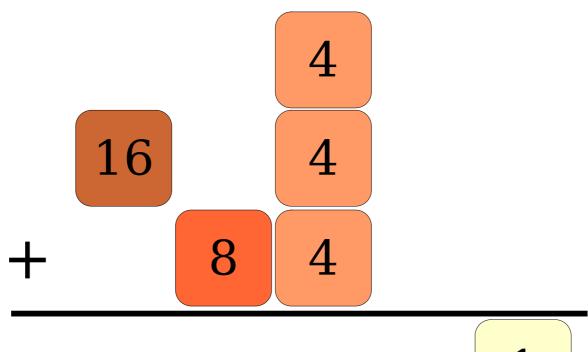
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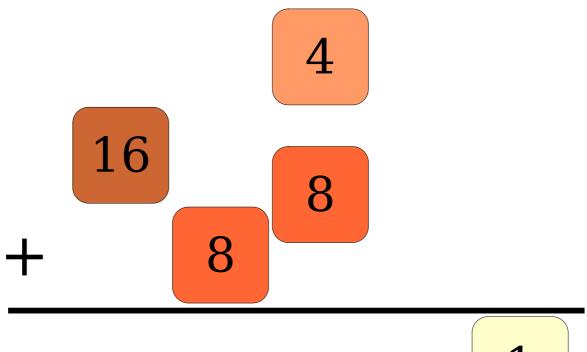
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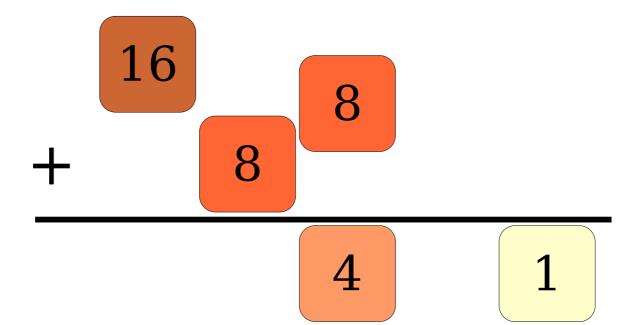
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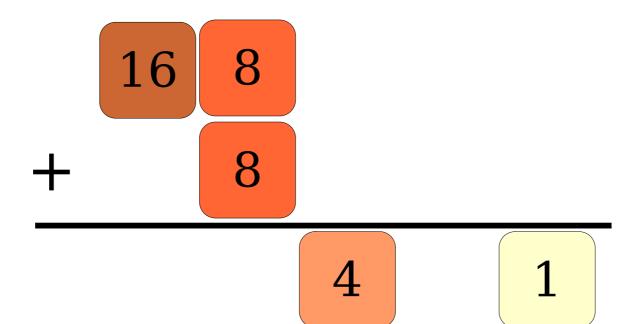
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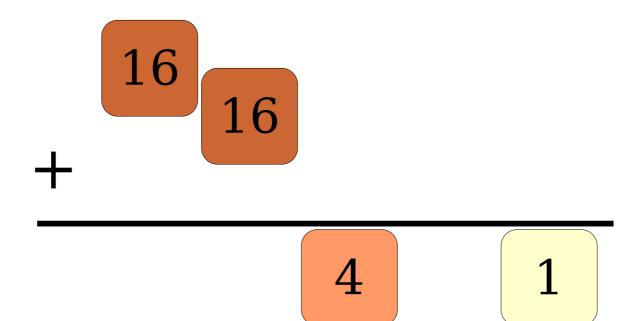
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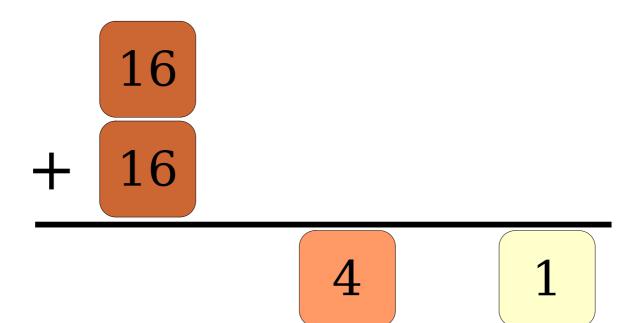
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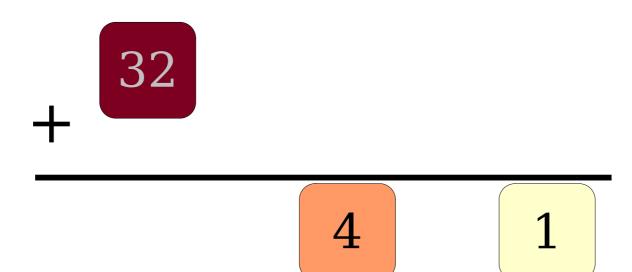
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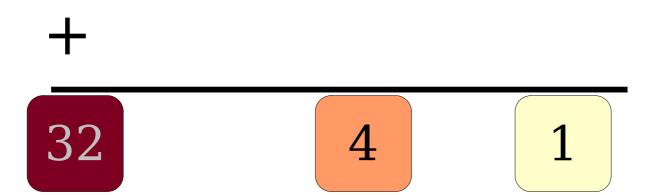
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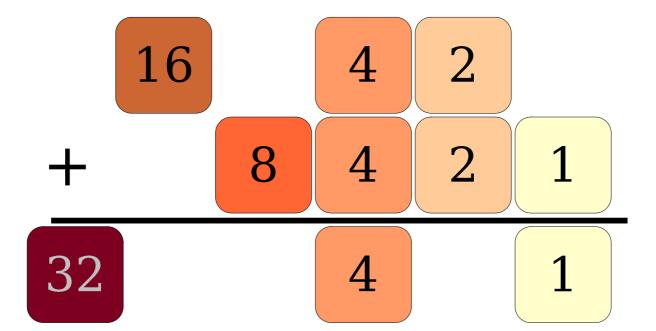
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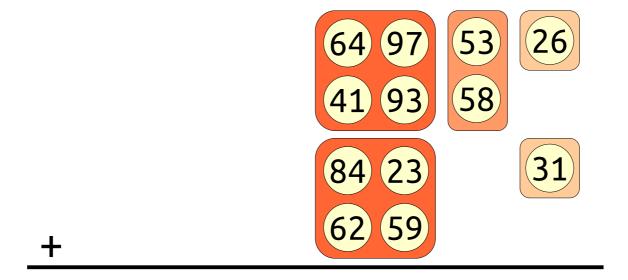


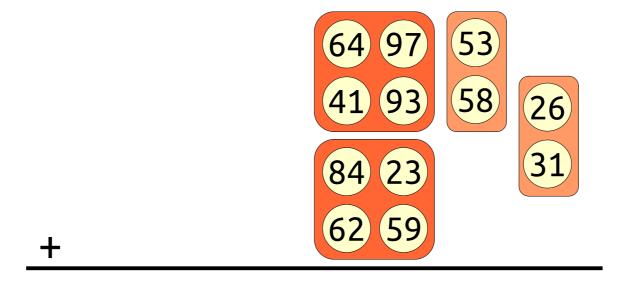
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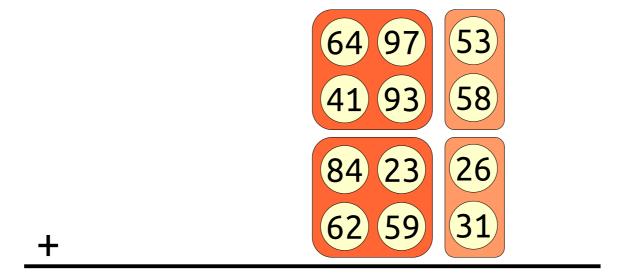


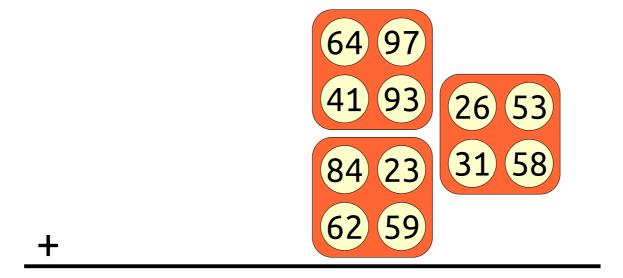
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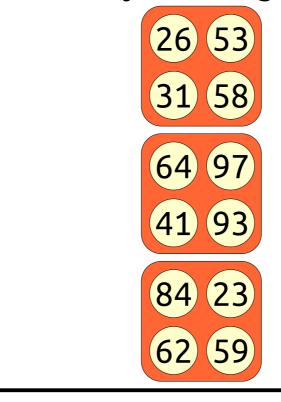












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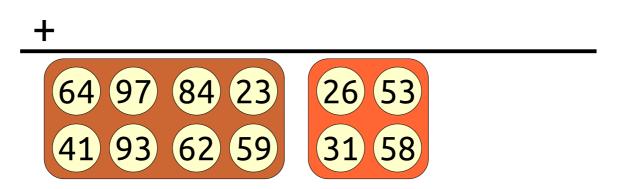


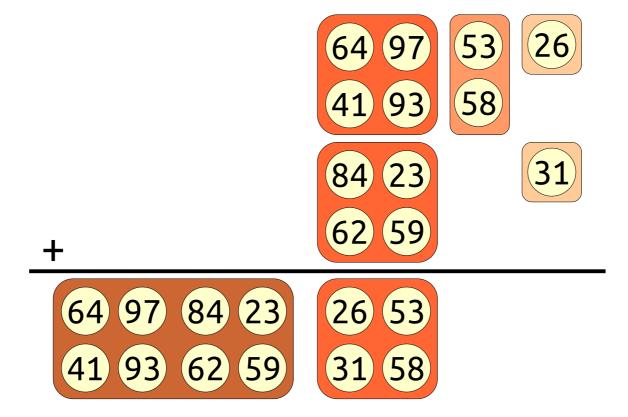
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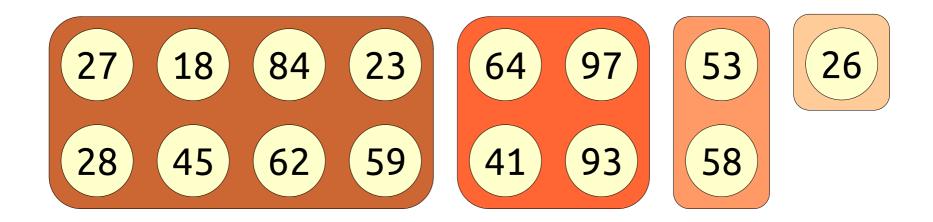




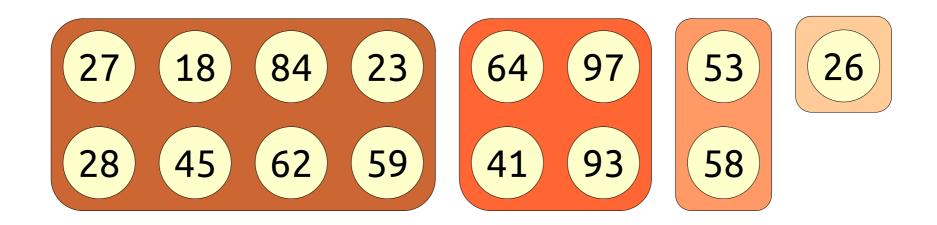


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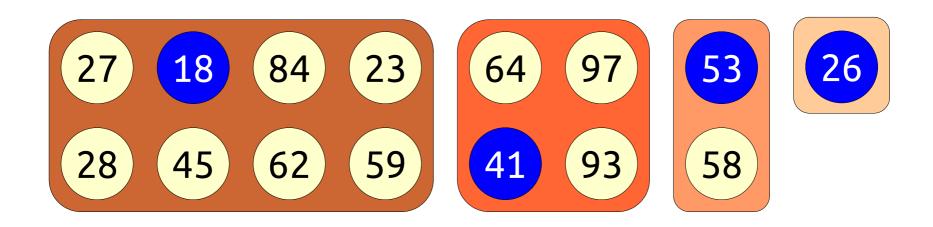
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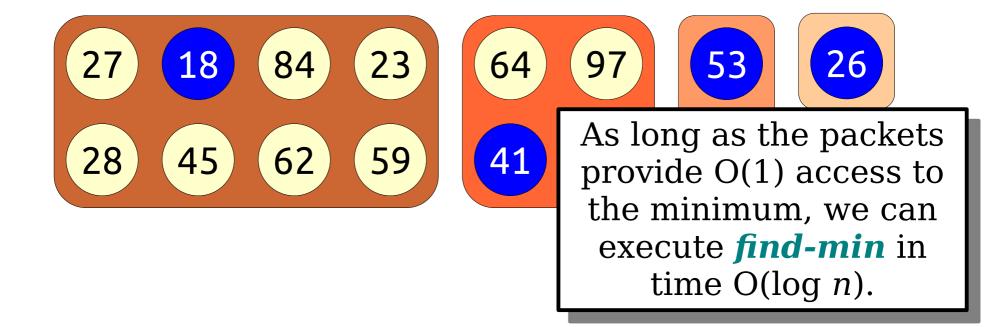
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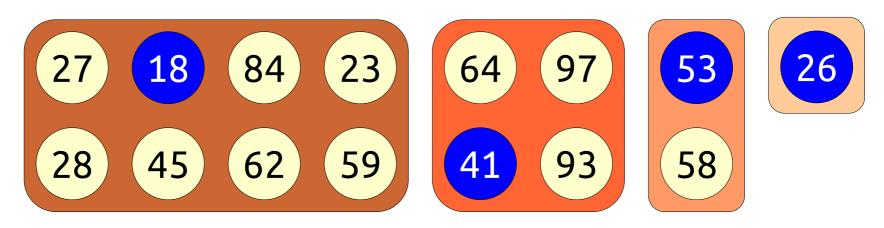
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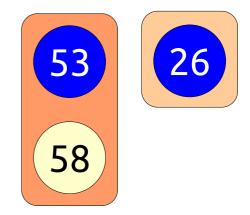
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 - Can efficiently fuse packets of the same size.
 - Can efficiently find the minimum element of each packet.



- 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.

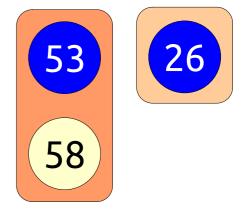
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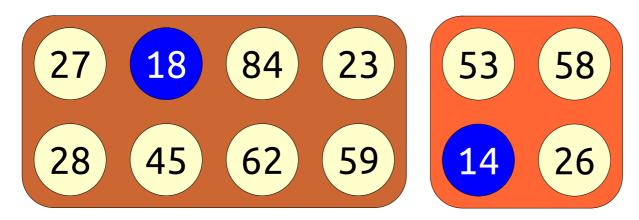


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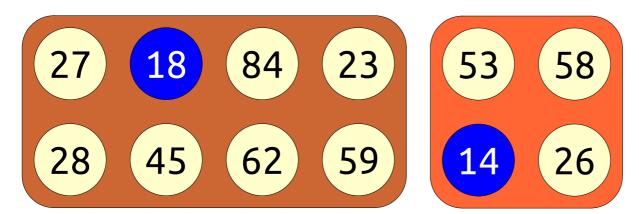




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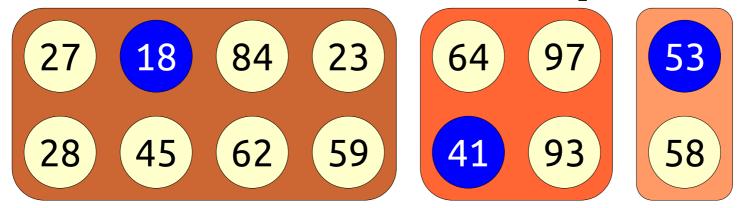
Time required: $O(\log n)$ fuses.

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.

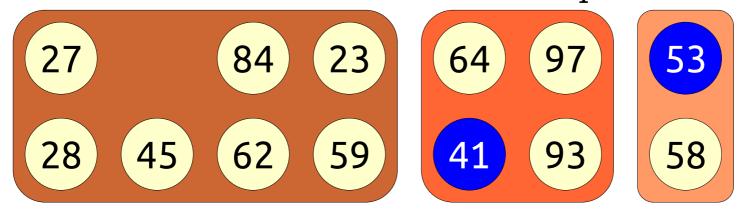
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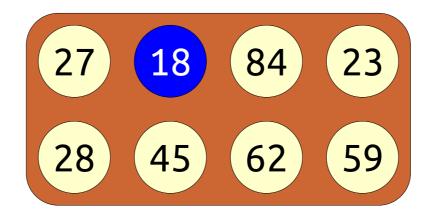


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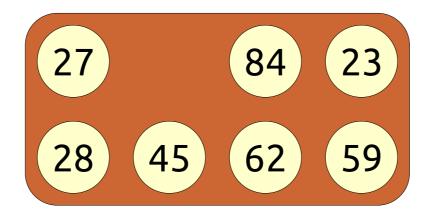
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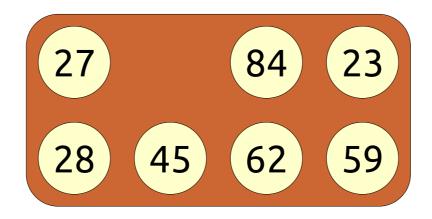
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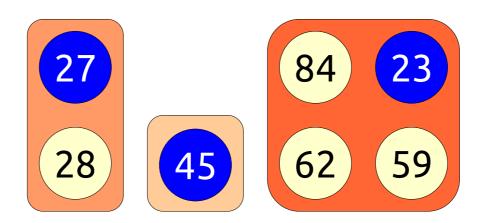
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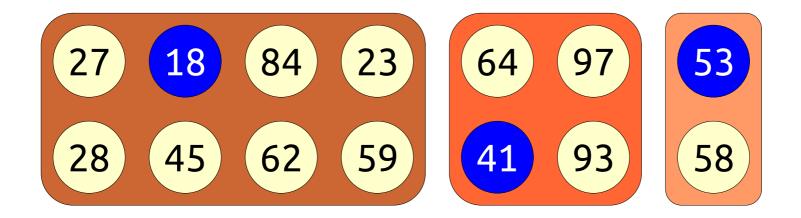


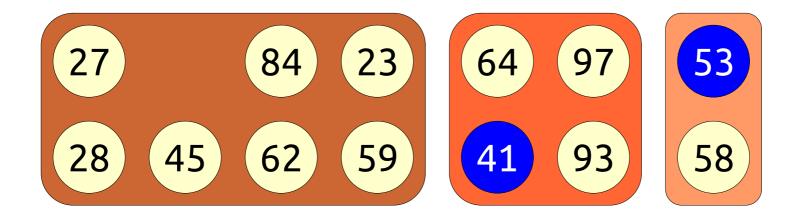
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- Fun fact: $2^k 1 = 2^0 + 2^1 + 2^2 + ... + 2^{k-1}$.
- *Idea*: "Fracture" the packet into *k* smaller packets, then add them back in.

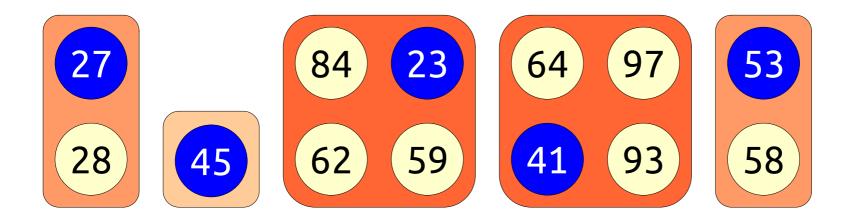


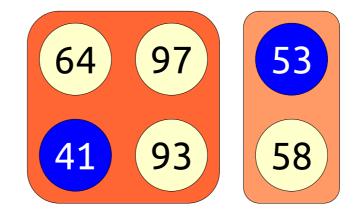
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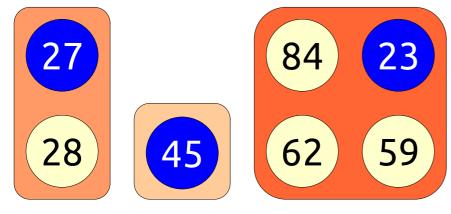


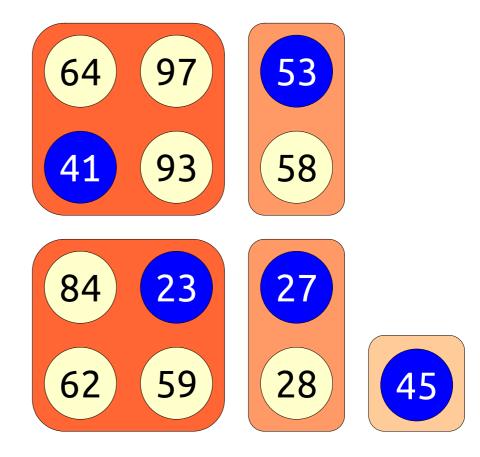


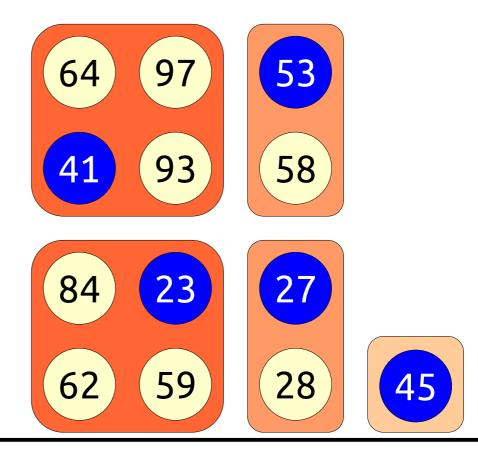




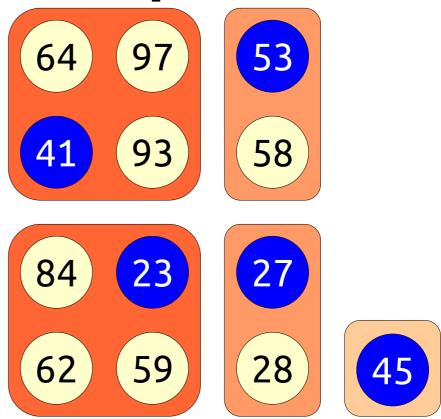








- We can *extract-min* by fracturing the packet containing the minimum and adding the fragments back in.
- Runtime is $O(\log n)$ fuses in **meld**, plus fracture cost.



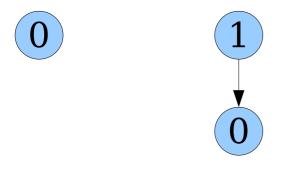
Building a Priority Queue

- 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 , ..., 2^{k-1} nodes.
- *Question:* How can we represent our packets to support the above operations efficiently?

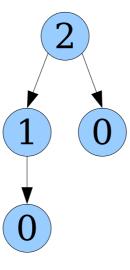
 A binomial tree of order k is a type of tree recursively defined as follows:

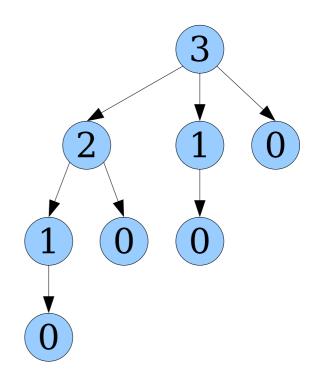
A binomial tree of order k is a single node whose children are binomial trees of order 0, 1, 2, ..., k-1.

Here are the first few binomial trees:



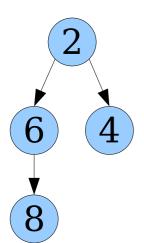
Why are these called binomial heaps? Look across the layers of these trees and see if you notice anything!



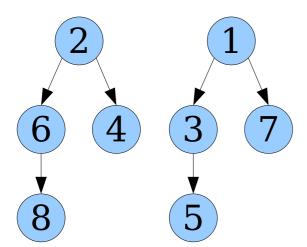


- What properties must our packets have?
 - Size must be a power of two.
 - Can efficiently fuse packets of the same size.
 - Can efficiently find the minimum element of each packet.
 - Can efficiently "fracture" a packet of 2^k nodes into packets of 2^0 , 2^1 , 2^2 , 2^3 , ..., 2^{k-1} nodes.

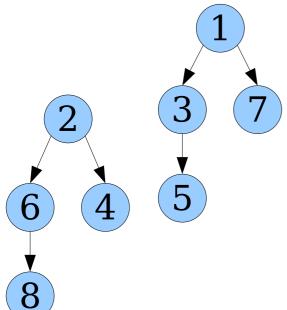
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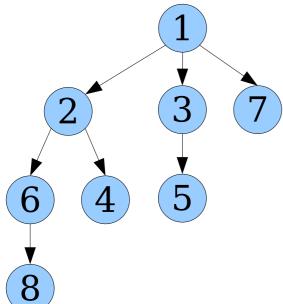
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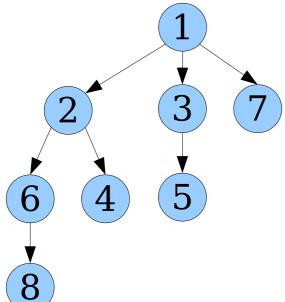
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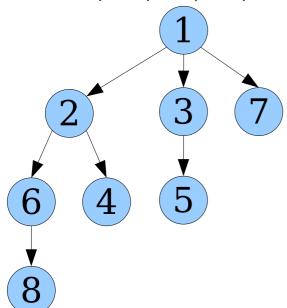


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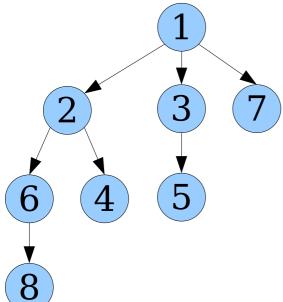


Make the binomial tree with the larger root the first child of the tree with the smaller root.

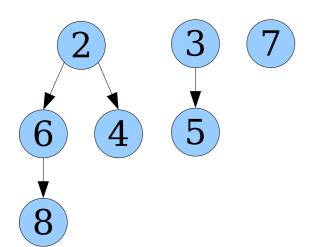
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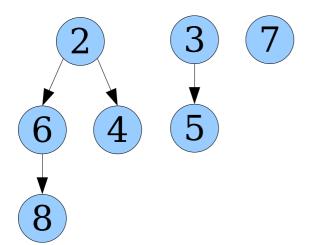
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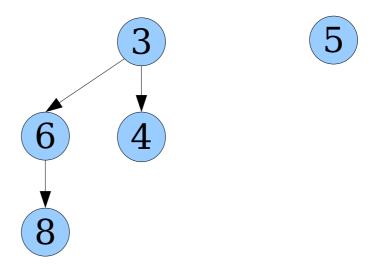


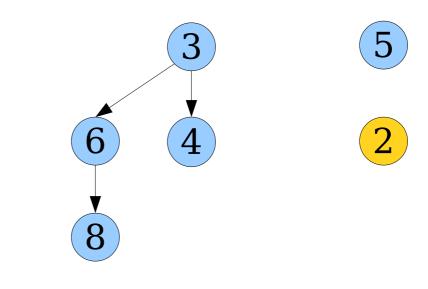
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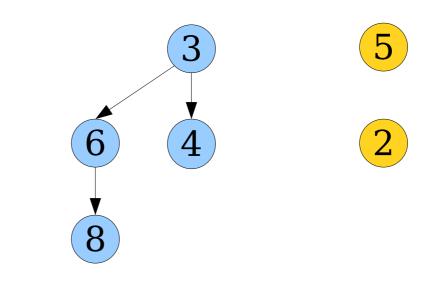


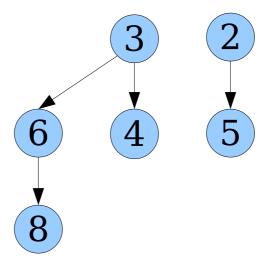
The Binomial Heap

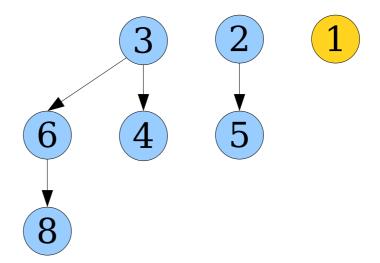
- A *binomial heap* is a collection of binomial trees stored in ascending order of size.
- Operations defined as follows:
 - $meld(pq_1, pq_2)$: Use addition to combine all the trees.
 - Fuses $O(\log n + \log m)$ trees. Cost: $O(\log n + \log m)$. Here, assume one binomial heap has n nodes, the other 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: $O(\log n)$.

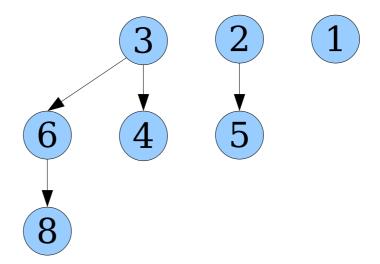


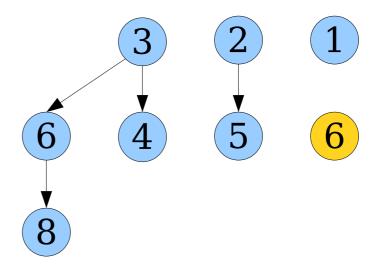


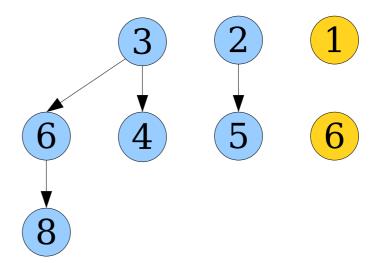


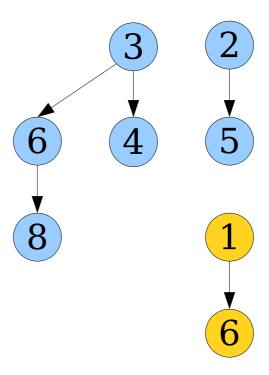


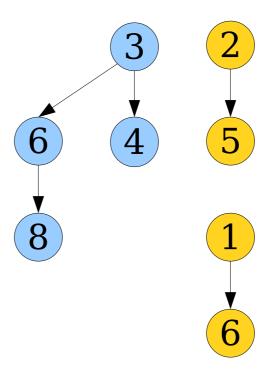


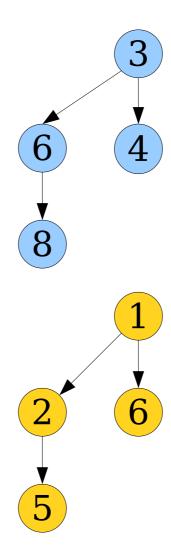


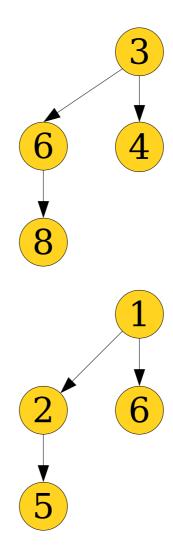


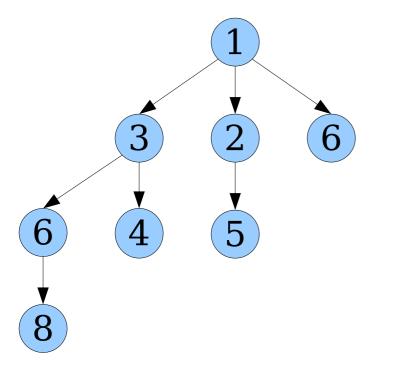


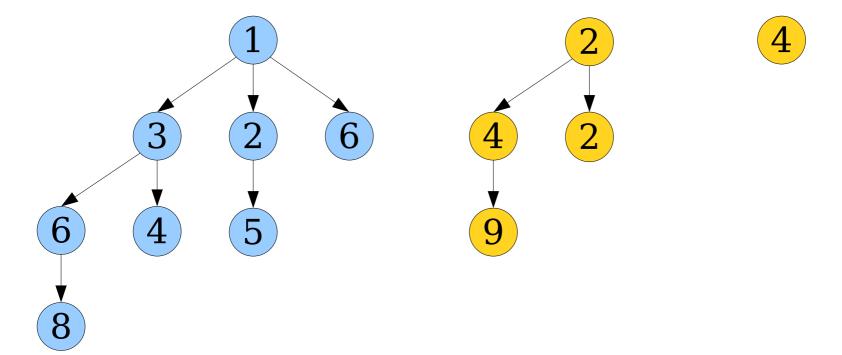


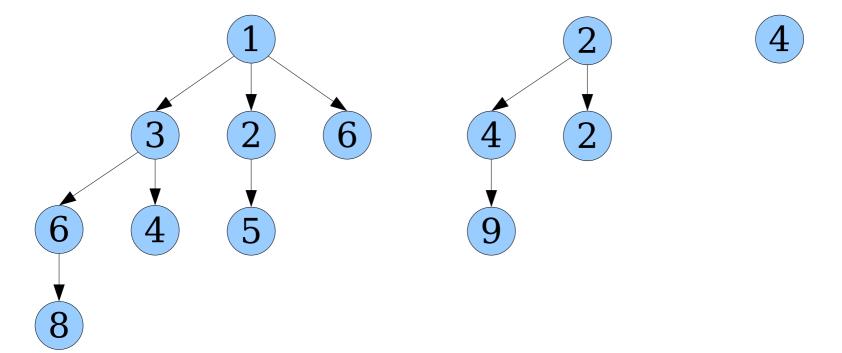


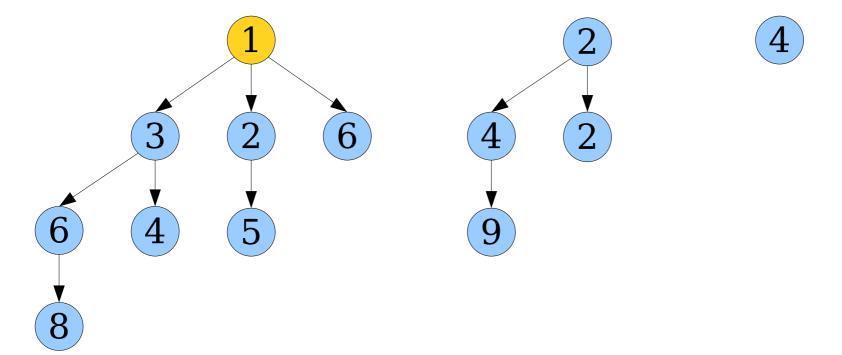


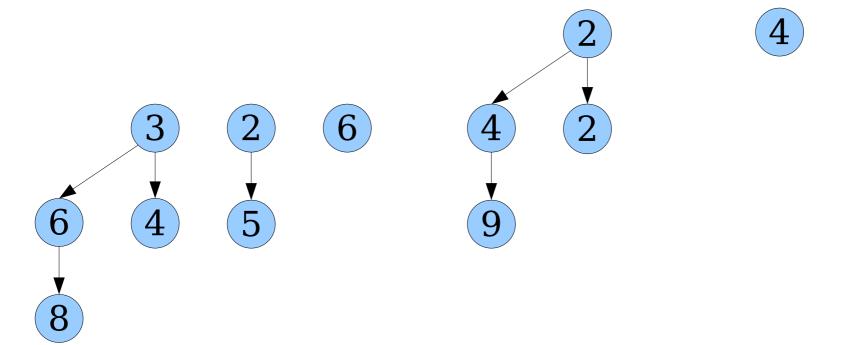


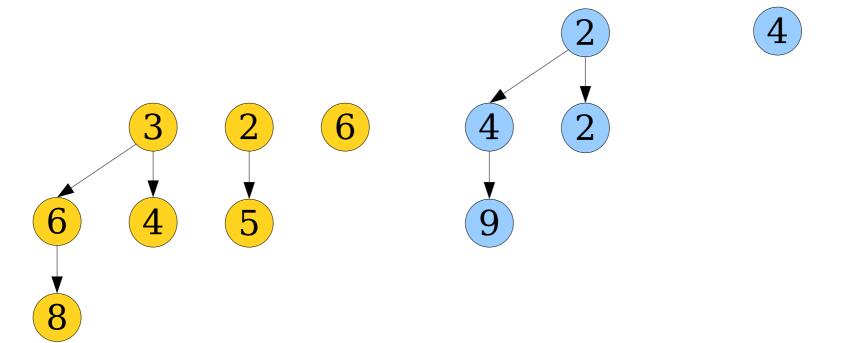


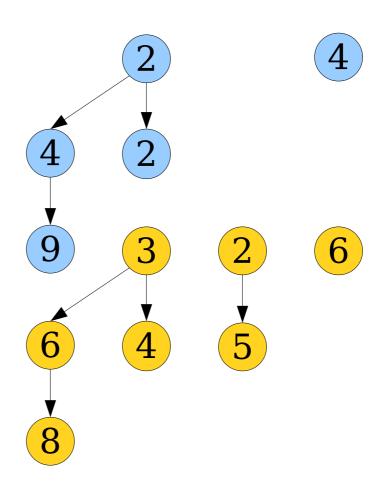


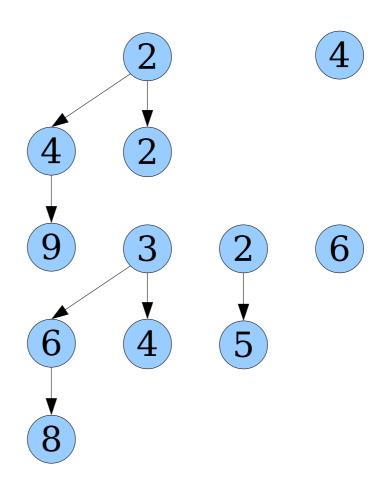


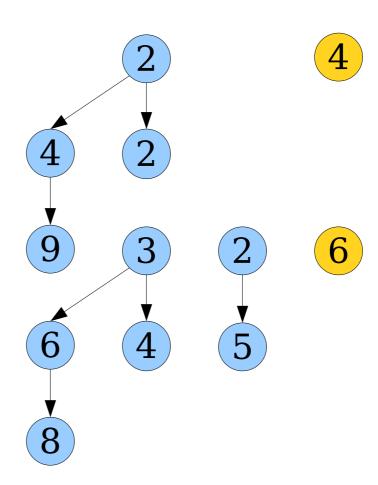


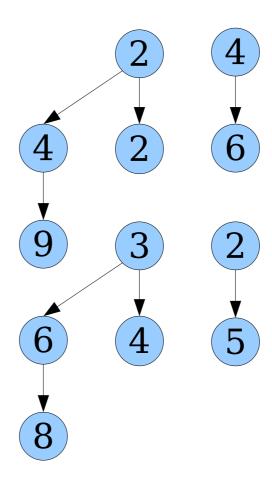


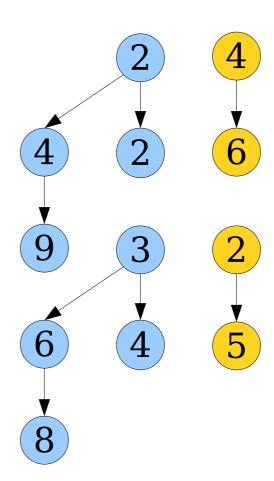


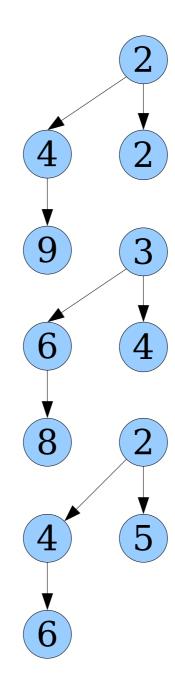


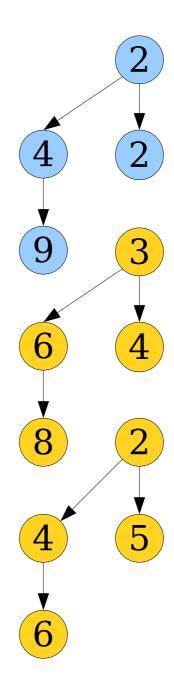


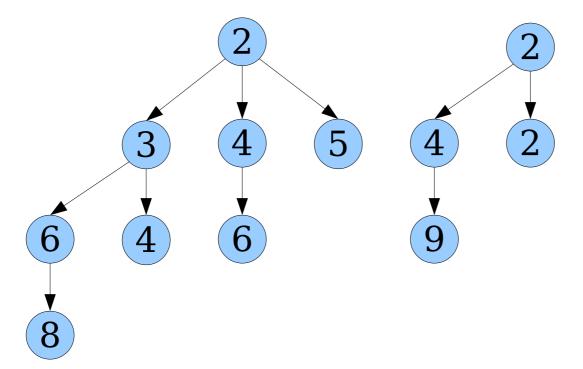


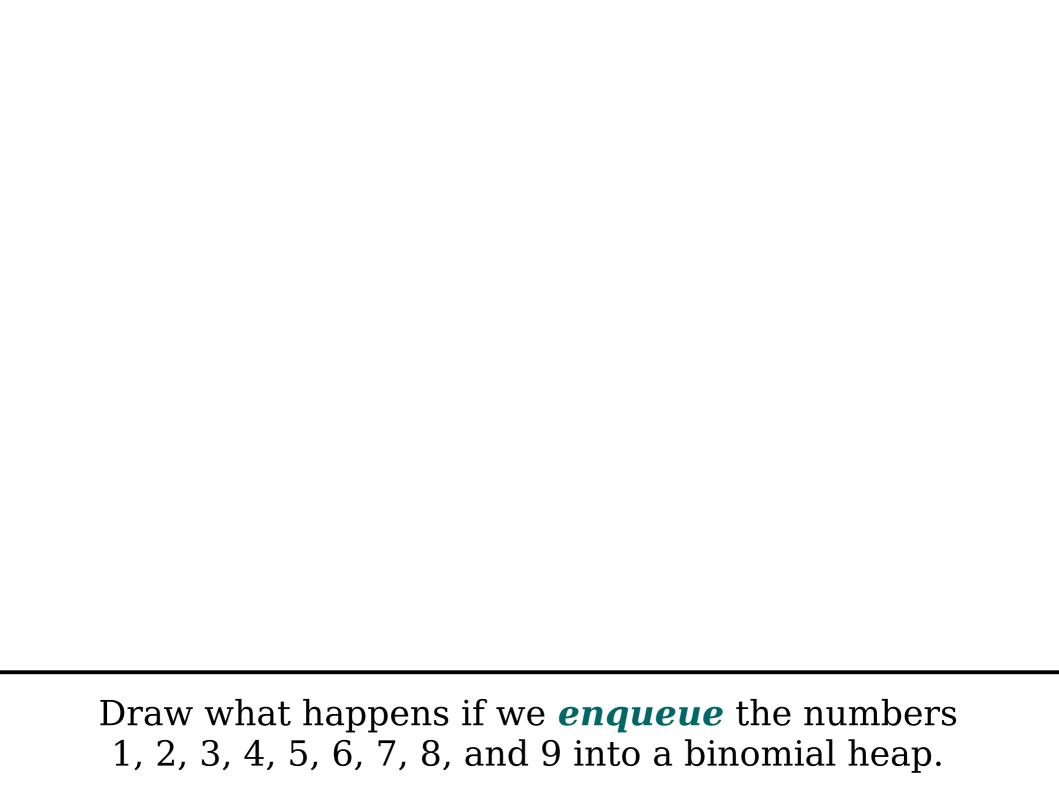


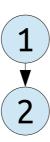


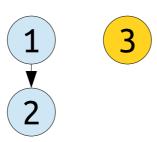


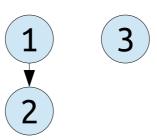




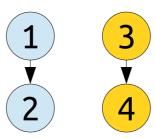




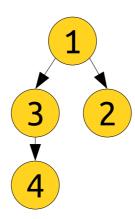


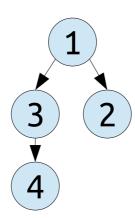


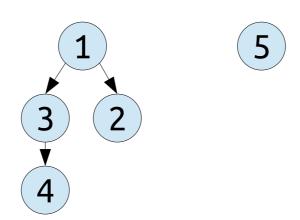
1 3 v 2 4

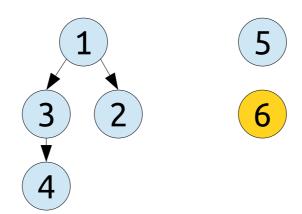


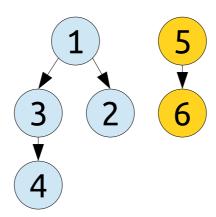


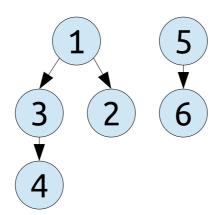


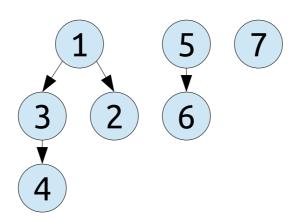


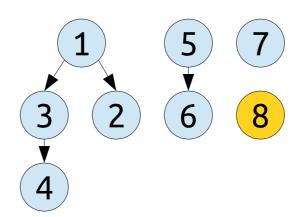


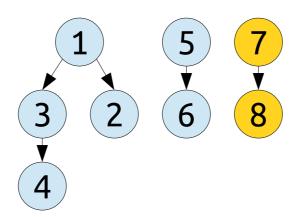


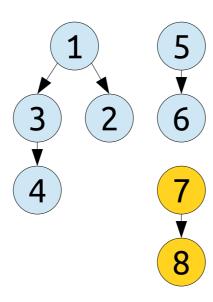


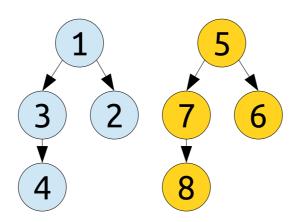


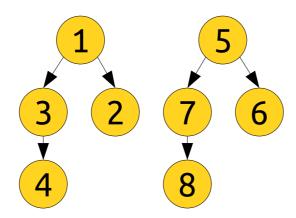


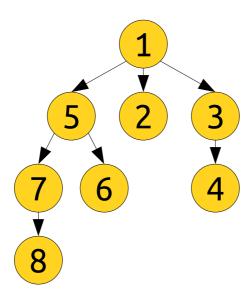


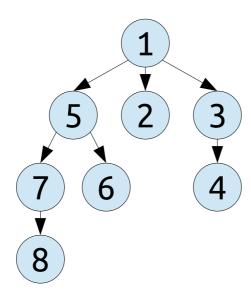




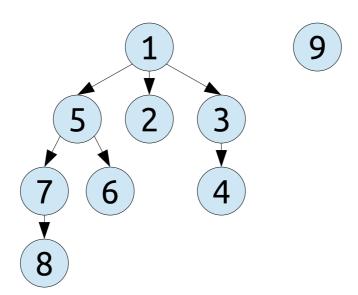




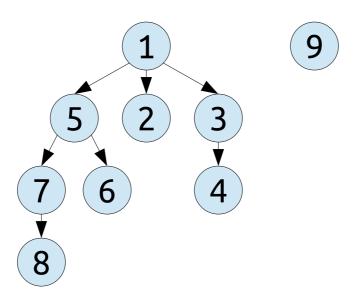


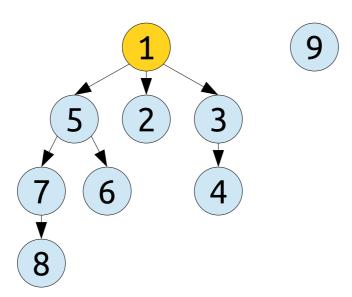


Draw what happens if we *enqueue* the numbers 1, 2, 3, 4, 5, 6, 7, 8, and 9 into a binomial heap.

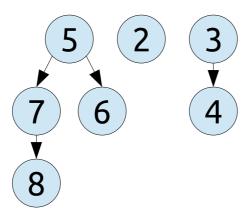


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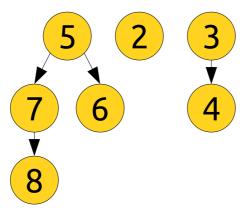


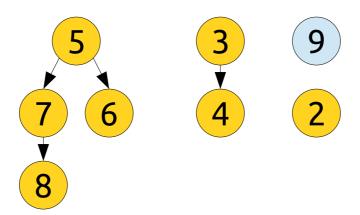


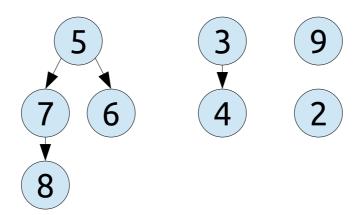


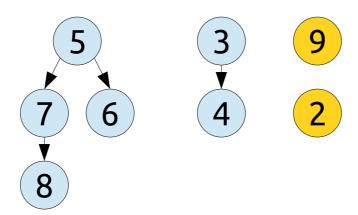


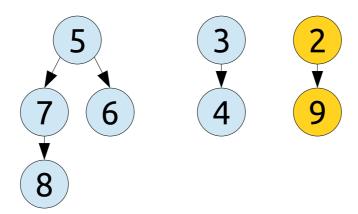


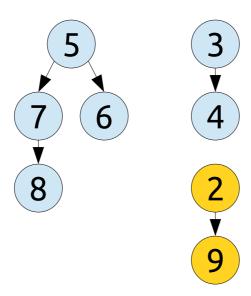


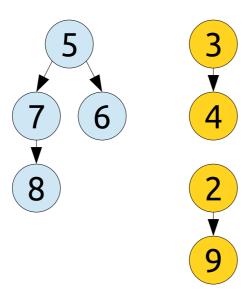


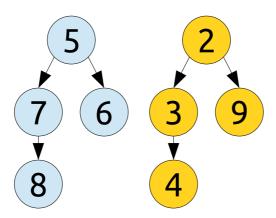


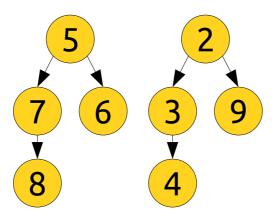


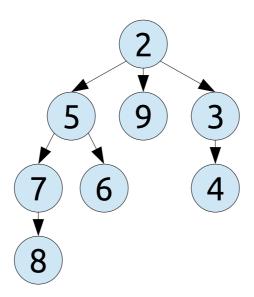












- Here's the current scorecard for the binomial heap.
- This is a fast, elegant, and clever data structure.
- **Question:** Can we do better?

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

- **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 extractmins in time o(n log n). This is impossible with comparison-based algorithms.

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

- 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).

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

- The *enqueue* operation is implemented in terms of *meld*.
- If we want

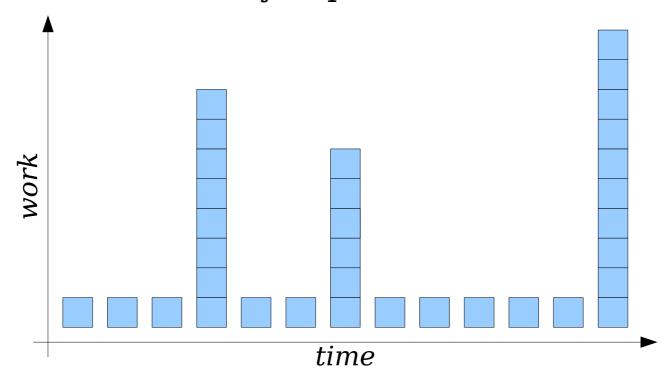
 enqueue to run
 in time O(1),
 we'll need meld
 to take time O(1).
- How could we accomplish this?

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

Thinking With Amortization

Refresher: Amortization

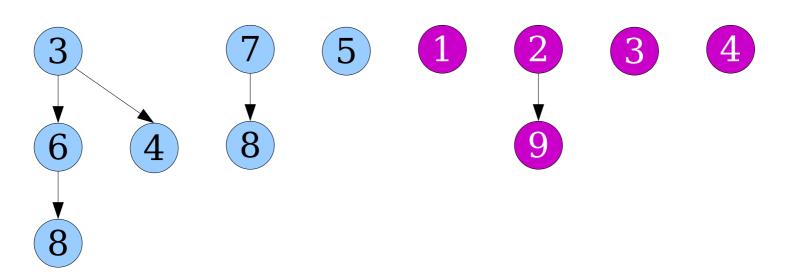
- In an amortized efficient data structure, some operations can take much longer than others, provided that previous operations didn't take too long to finish.
- Think dishwashers: you may have to do a big cleanup at some point, but that's because you did basically no work to wash all the dishes you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



Consider the following lazy melding approach:

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

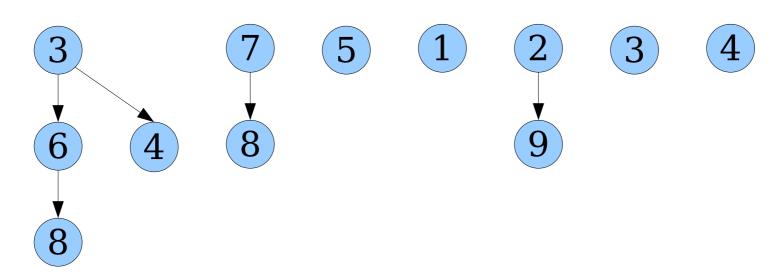
• *Intuition:* Why do any work to organize keys if we're not going to do an *extract-min*? We'll worry about cleanup then.



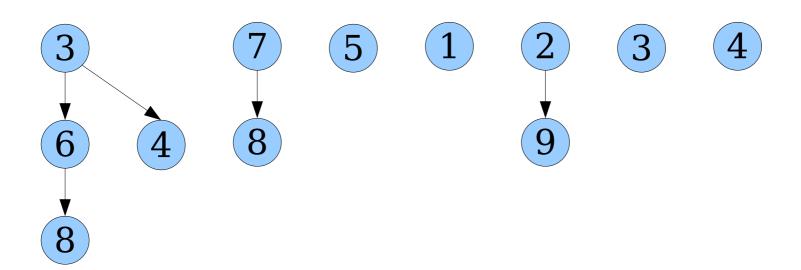
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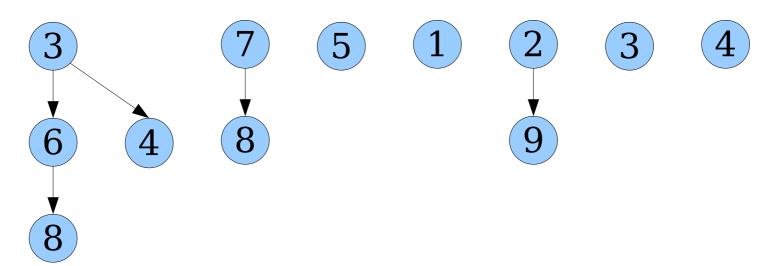
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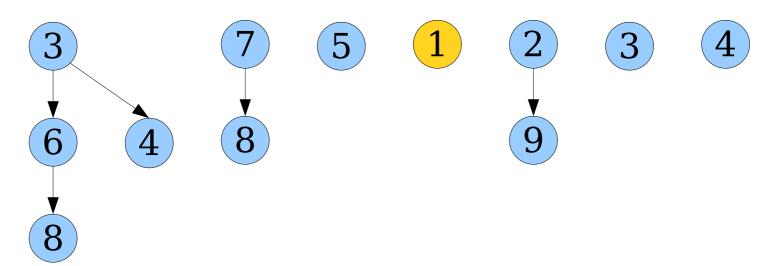
- If we store our list of trees as circularly, doubly-linked lists, we can concatenate tree lists in time O(1).
 - Cost of a *meld*: **O(1)**.
 - Cost of an *enqueue*: **O(1)**.
- If it sounds too good to be true, it probably is.



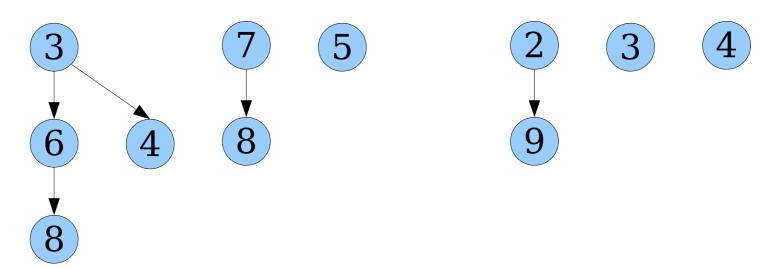
- Imagine that we implement *extract-min* the same way as before:
 - Find the packet with the minimum.
 - "Fracture" that packet to expose smaller packets.
 - Meld those packets back in with the master list.
- What happens if we do this with lazy melding?



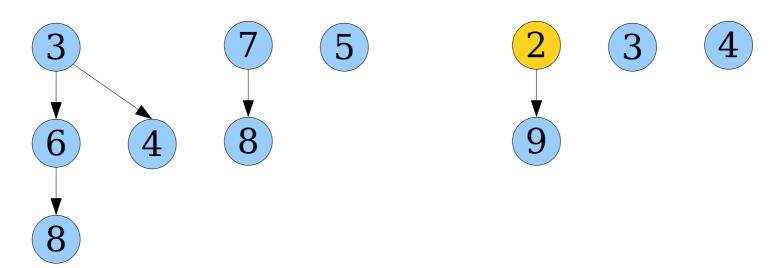
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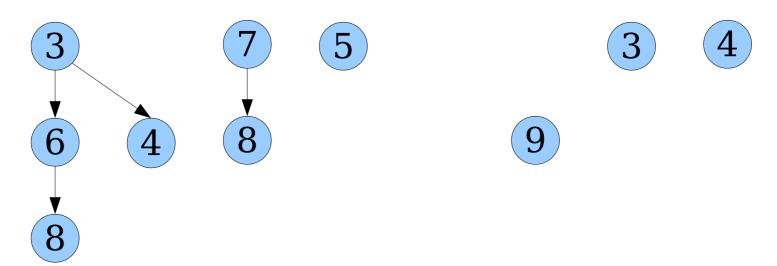
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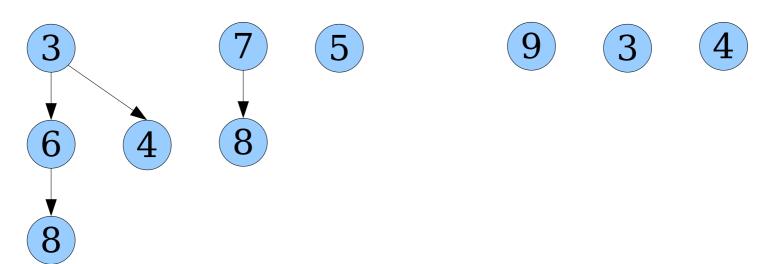
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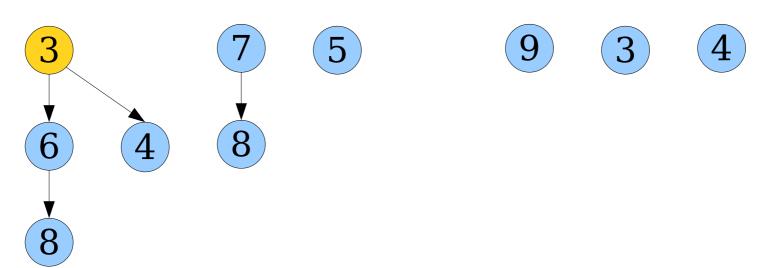
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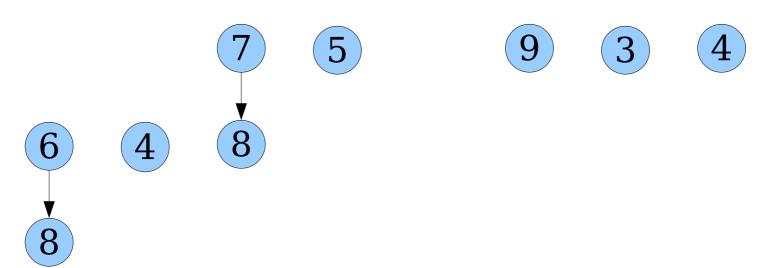
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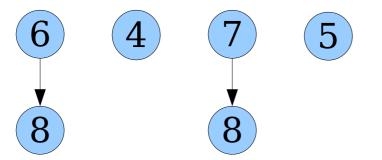
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Each pass of finding the minimum value takes time $\Theta(n)$ in the worst case. We've lost our nice runtime guarantees!

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.



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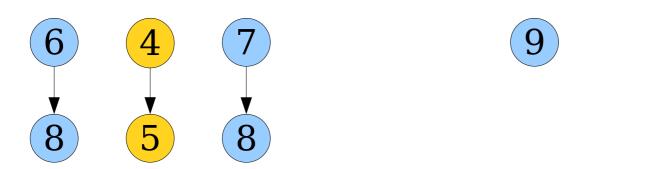
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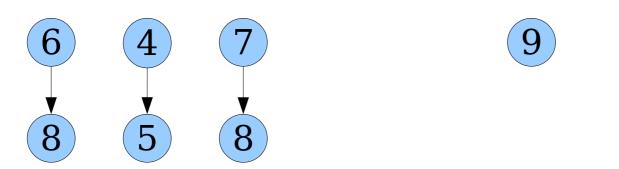
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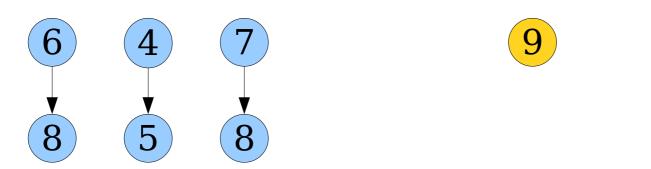
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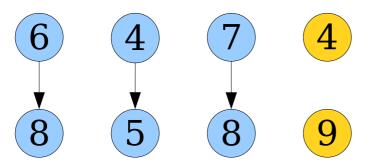
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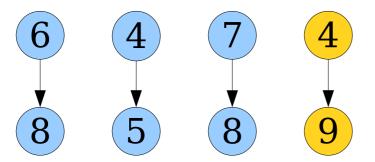
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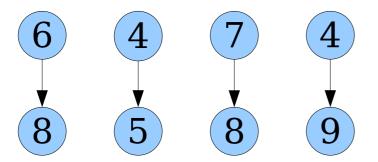
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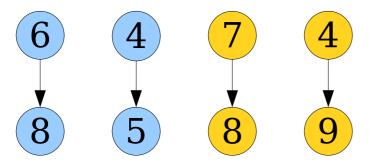
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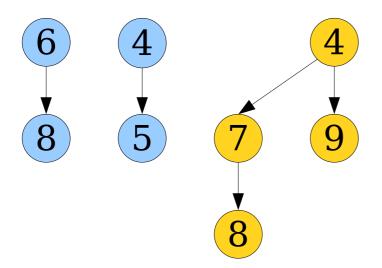
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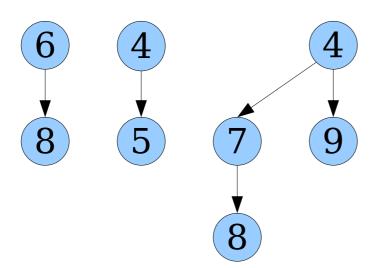
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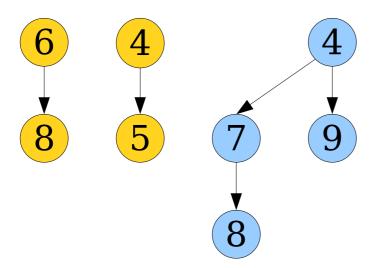
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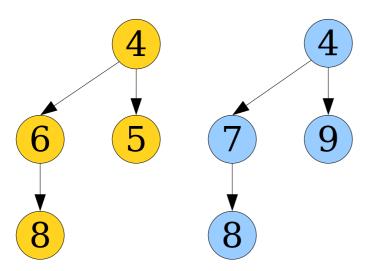
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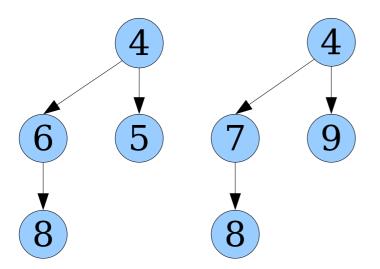
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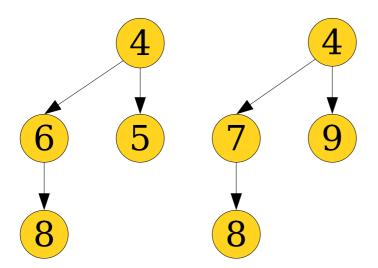
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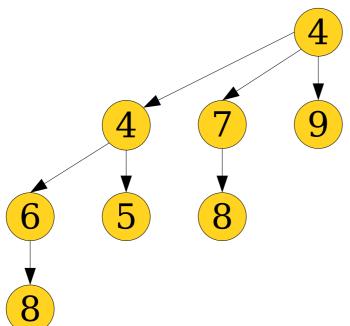
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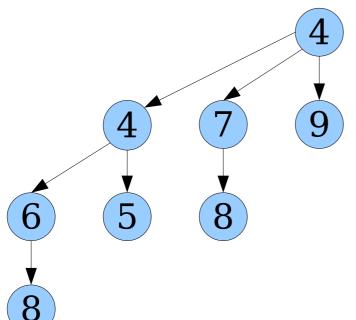
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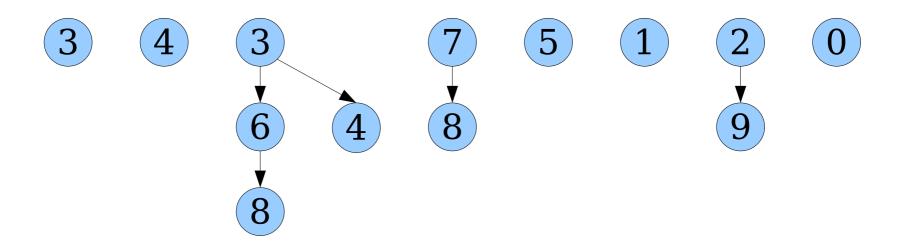


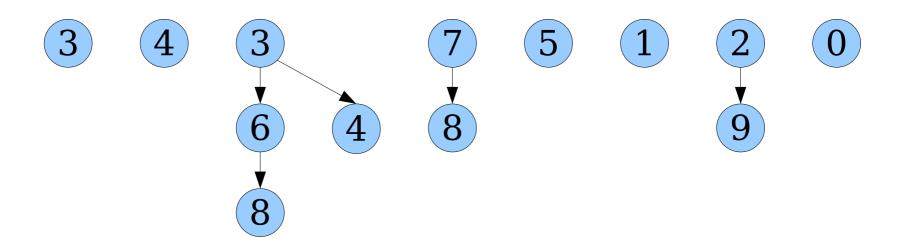
At this point, the mess is cleaned up, and we're left with what we would have had if we had been cleaning up as we go.

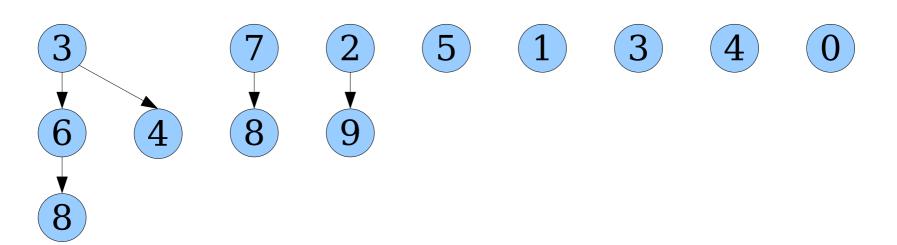
Where We're Going

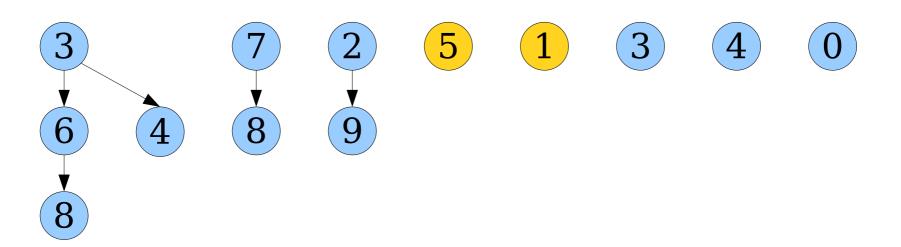
- A lazy binomial heap is a binomial heap, modified as follows:
 - The *meld* operation is lazy. It just combines the two groups of trees together.
 - After doing an extract-min, we do a coalesce to combine together trees until there's at most one tree of each order.
- Intuitively, we'd expect this to 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?

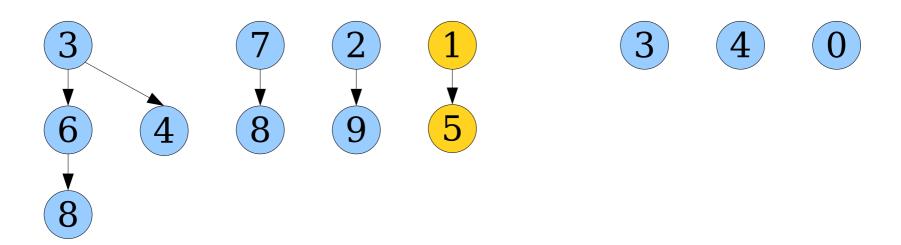
- The *coalesce* step repeatedly combines trees together until there's at most one tree of each order.
- How do we implement this so that it runs quickly?

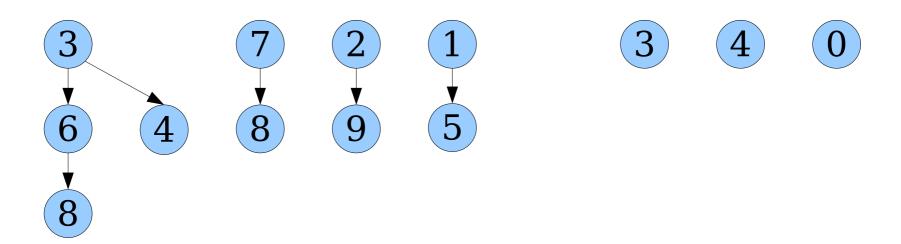


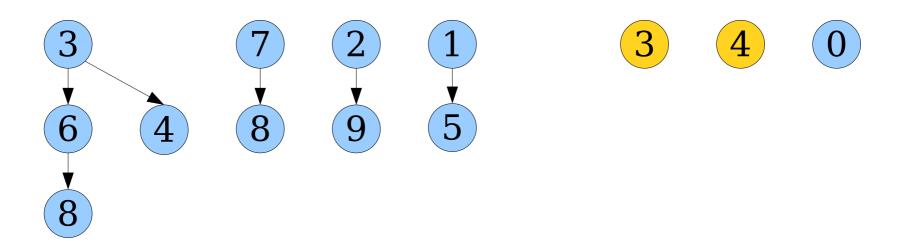


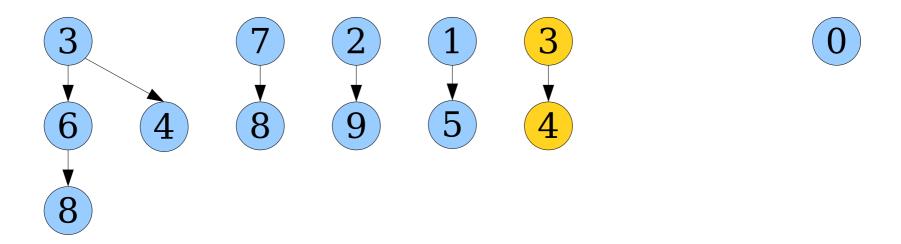


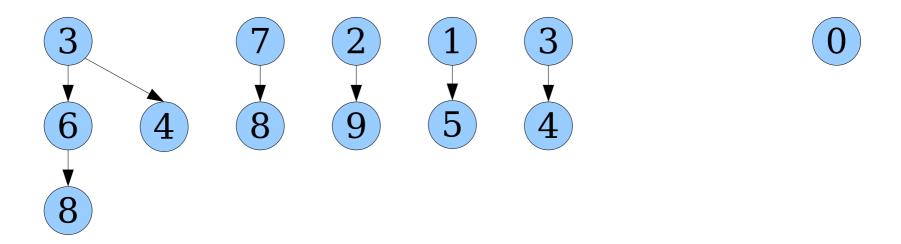


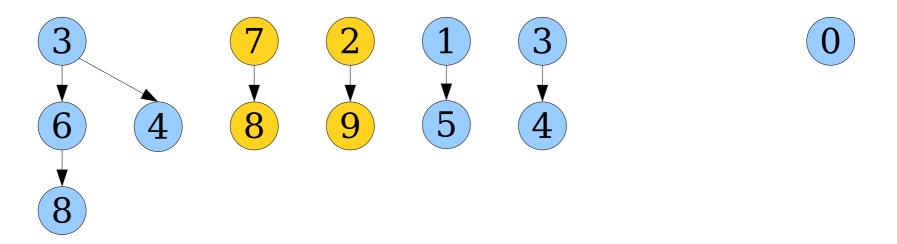


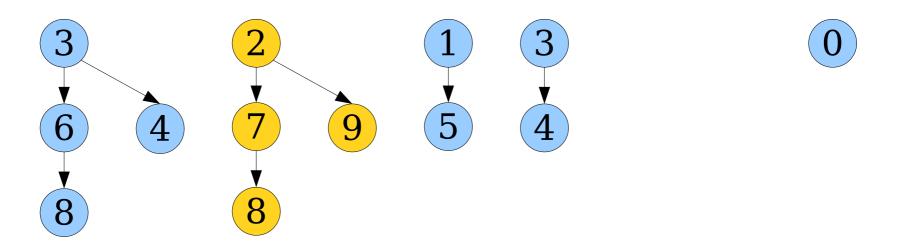


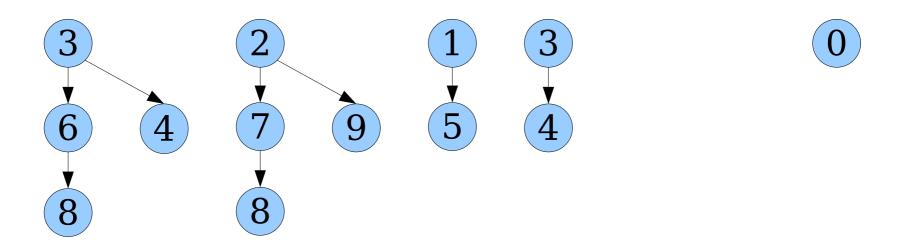


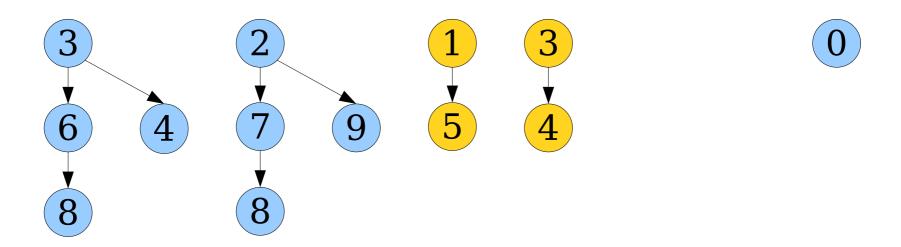


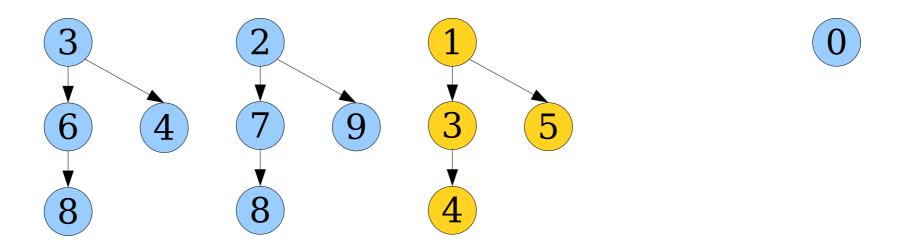


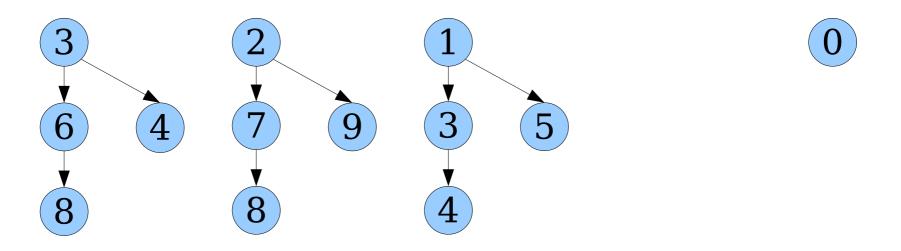


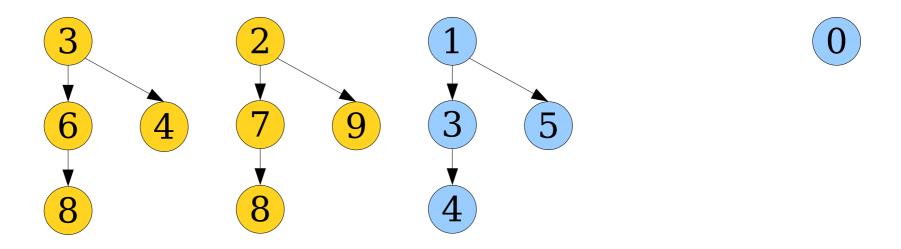


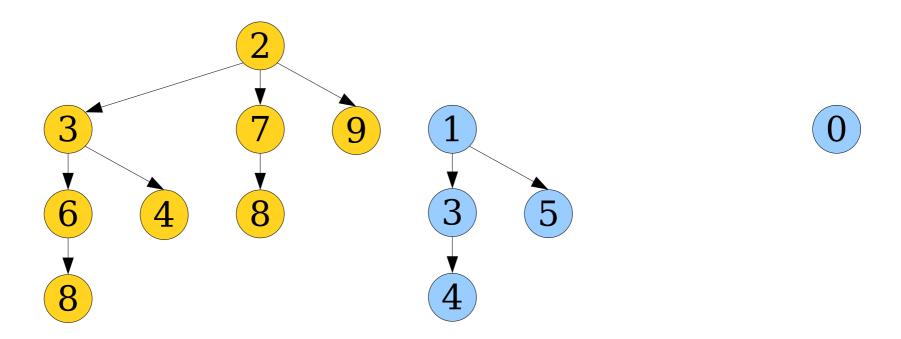


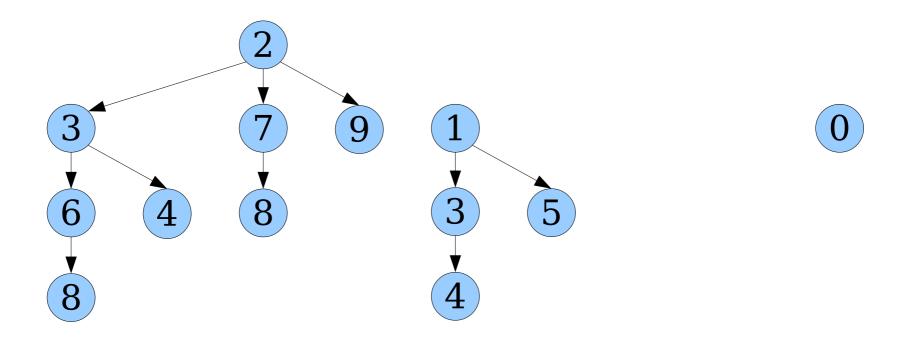




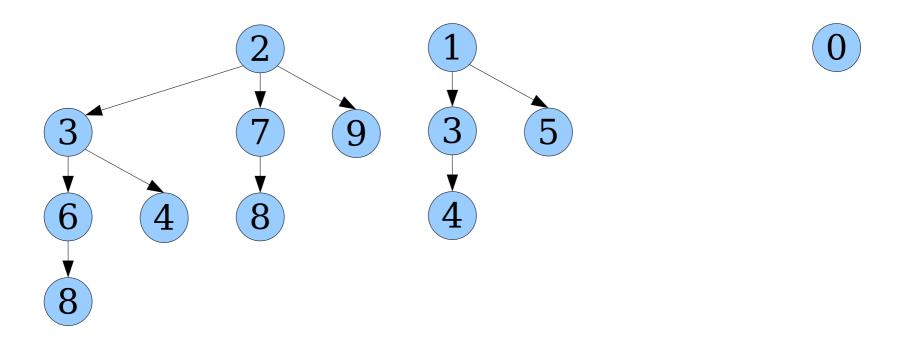






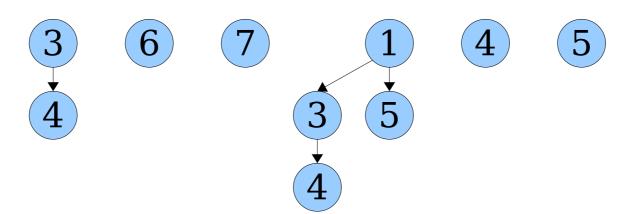


• *Observation:* This would be a *lot* easier to do if all the trees were sorted by size.

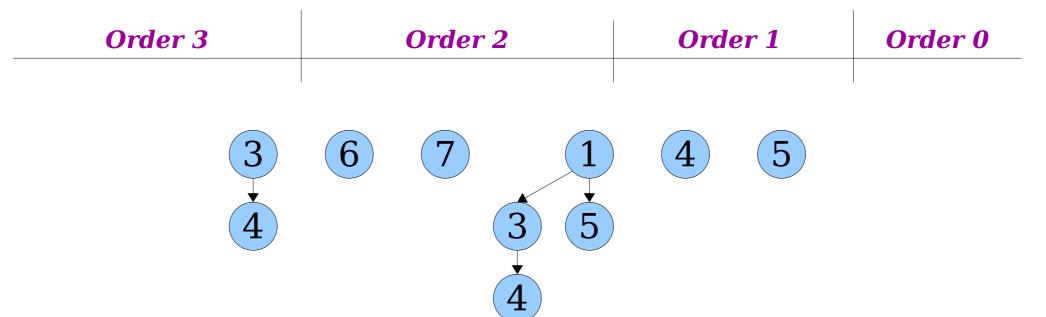


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- We can sort our group of t trees by size in time $O(t \log t)$ using a standard sorting algorithm.
- **Better idea:** All the sizes are small integers. Use counting sort!

- 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.



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	1	3 5	4
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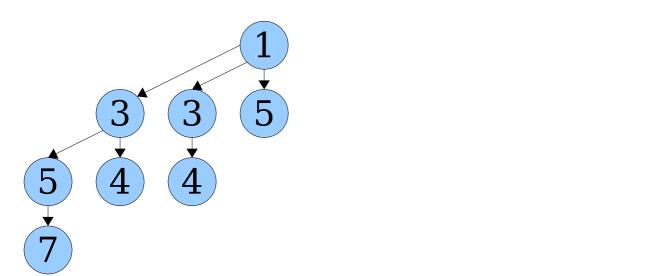
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Analyzing Coalesce

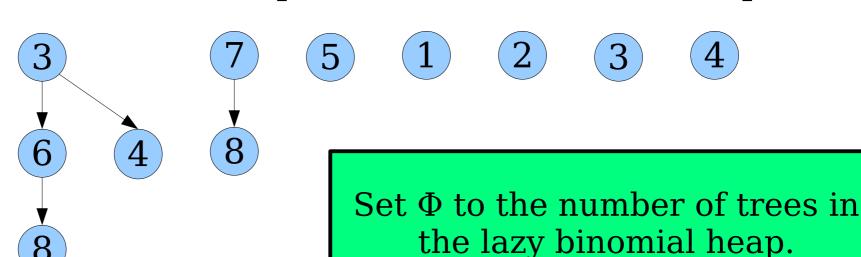
- *Claim:* Coalescing a group of t trees takes time $O(t + \log n)$.
 - Time to create the array of buckets: $O(\log n)$.
 - Time to distribute trees into buckets: O(t).
 - Time to fuse trees: $O(t + \log n)$
 - Number of fuses is O(t), since each fuse decreases the number of trees by one. Cost per fuse 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 Φ .
- 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.



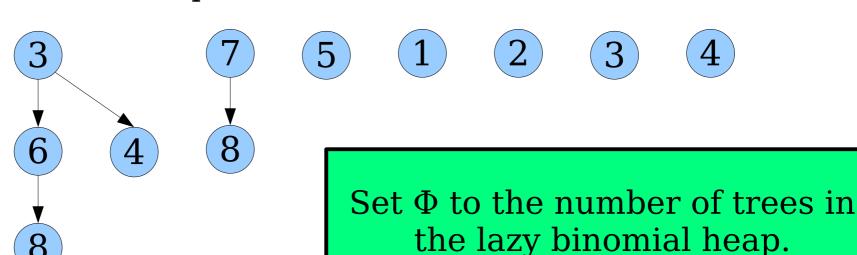
An Amortized Analysis

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

$$amortized-cost = real-cost + k \cdot \Delta \Phi$$
,

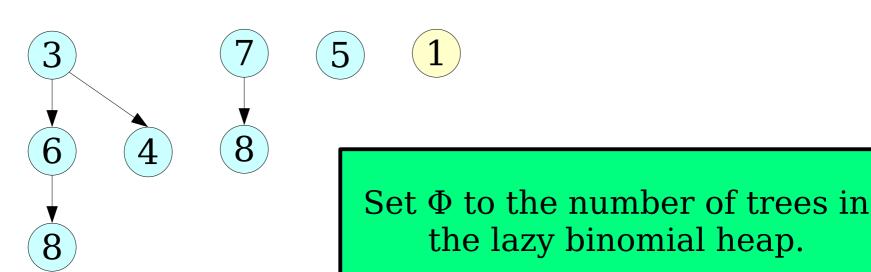
where $\Delta \Phi = \Phi_{after} - \Phi_{before}$.

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



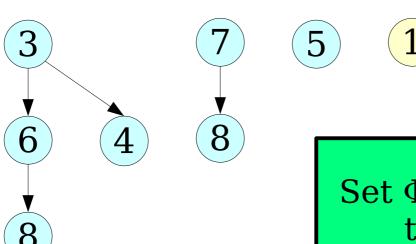
Analyzing an Insertion

• To *enqueue* a key, we add a new binomial tree to the forest.



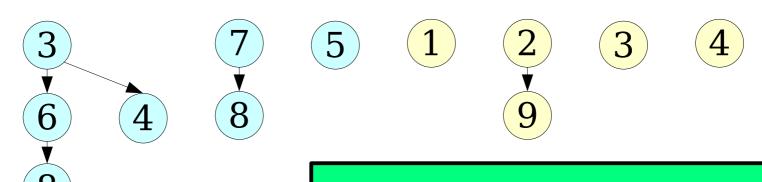
Analyzing an Insertion

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



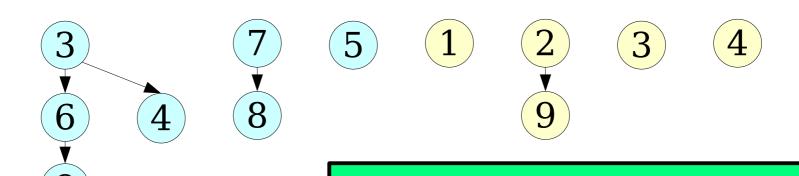
Set Φ to the number of trees in the lazy binomial heap.

- What is the amortized cost of *meld*?
- The real cost is O(1).
- What's $\Delta\Phi$?
- That's trickier there are two separate collections of trees here.



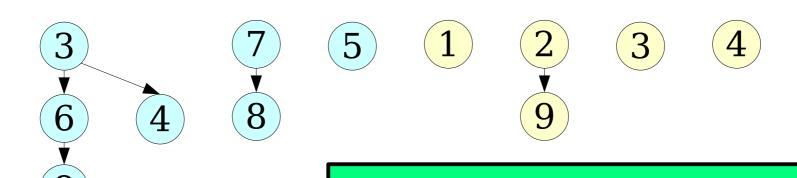
Set Φ to the number of trees in the lazy binomial heap.

- What is the amortized cost of *meld*?
- Common trick: When working with mergeable data structures, define Φ globally across all instances of the data structure.



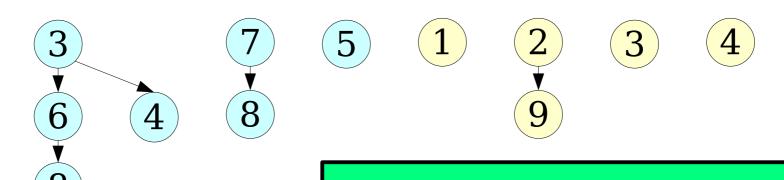
Set Φ to the number of trees in the lazy binomial heap.

- What is the amortized cost of *meld*?
- Common trick: When working with mergeable data structures, define Φ globally across all instances of the data structure.



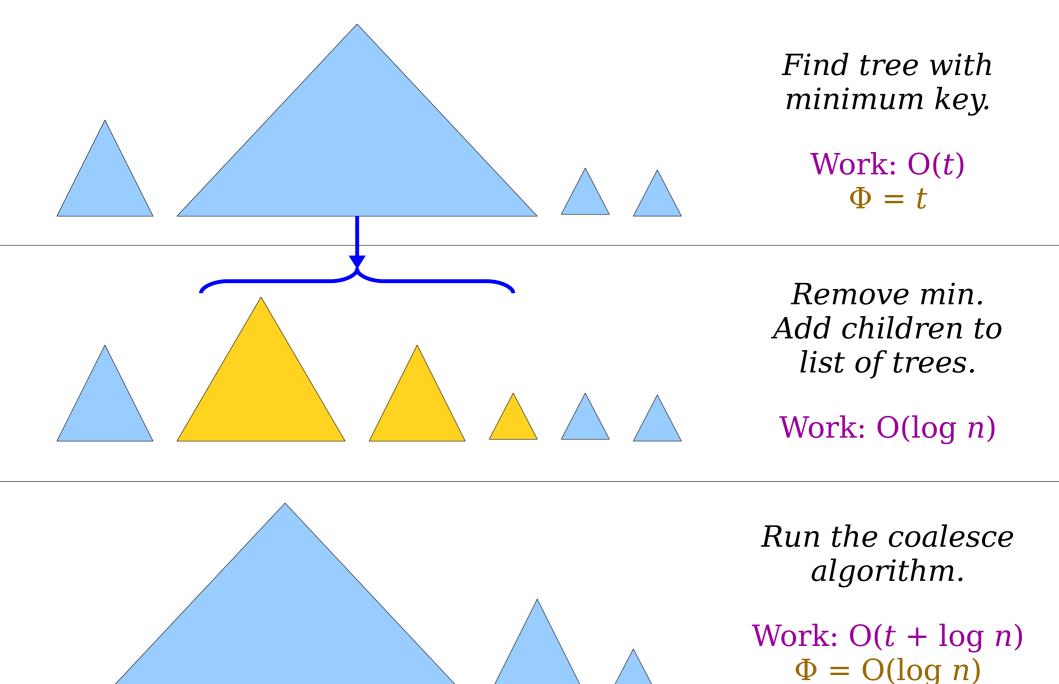
Set Φ to the number of trees in *all* lazy binomial heaps.

- What is the amortized cost of **meld**?
- Common trick: When working with mergeable data structures, define Φ globally across all instances of the data structure.
- Now $\Delta \Phi = 0$ and the amortized cost is O(1).



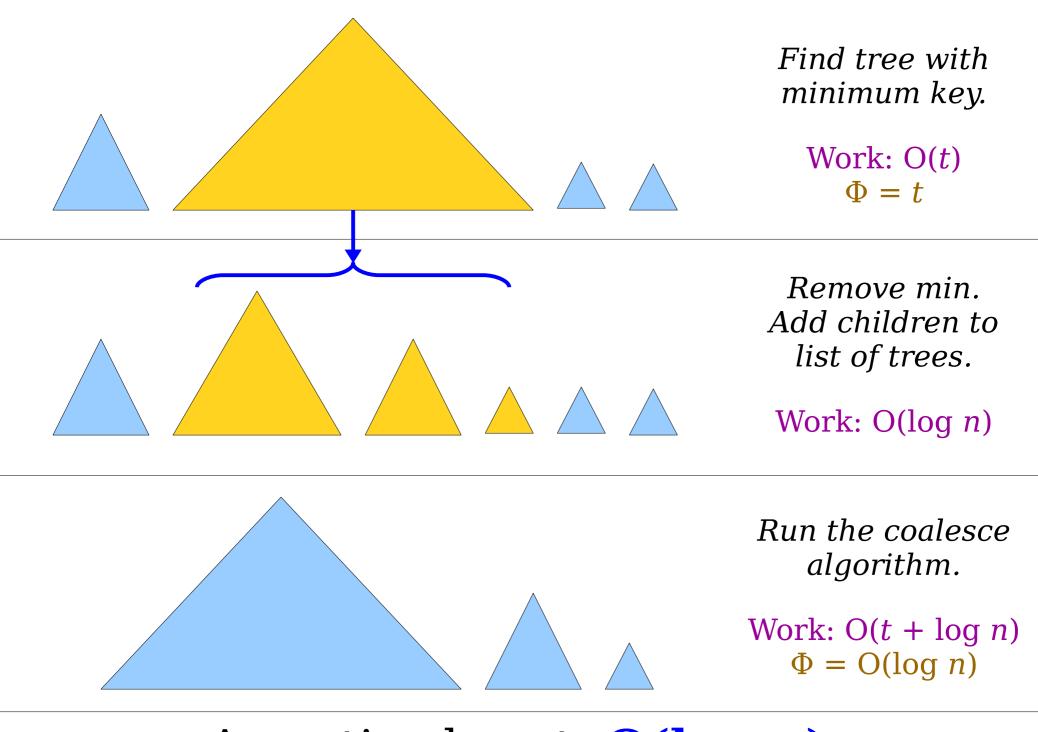
Set Φ to the number of trees in **all** lazy binomial heaps.

Analyzing *extract-min*



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

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



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

Analyzing Extract-Min

- Suppose we perform an extract-min on a lazy binomial heap with t trees in it.
- Initially, we fracture the tree containing the minimum. 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

```
O(t + \log n) + k \cdot (-t + O(\log n))
= O(t) - k \cdot t + k \cdot O(\log n)
= O(\log n).
```

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 worstcase efficient.
 - This is possible!
 Check out
 bootstrapped skew
 binomial 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*.
- Analyzing Fibonacci Heaps
 - A clever analysis.