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

## 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 $O(\log n)$; findmin 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 $\left(p q_{1}, p q_{2}\right)$ : Destroy $p q_{1}$ and $p q_{2}$ and combine their elements into a single priority queue. (MSTs via Cheriton-Tarjan)
- pq.decrease-key( $v, k^{\prime}$ ): Given a pointer to element $v$ already in the queue, lower its key to have new value $k^{\prime}$. (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-toall on the next problem set.


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$p q . \boldsymbol{d e c r e a s e - k e y}\left(v, k^{\prime}\right)$ : 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)
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Meldable Priority Queues

## Meldable Priority Queues

- A priority queue supporting the meld operation is called a meldable priority queue.
- meld( $p q_{1}, p q_{2}$ ) destructively modifies $p q_{1}$ and $p q_{2}$ and produces a new priority queue containing all elements of $p q_{1}$ and $p q_{2}$.



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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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$$
\begin{array}{r}
101110 \\
+\quad 1111
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| 1 | 1 | 1 | 1 |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- |
|  | 1 | 0 | 1 | 1 | 0 |
| + |  | 1 | 1 | 1 | 1 |
| 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

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16
$+\quad 16$

| 4 | 1 |
| :--- | :--- |

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> | 64 | 97 | 53 |
| :--- | :--- | :--- |
| 41 | 93 | 58 |
| 84 | 23 | 26 |
| 62 | 59 | 31 |

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6497
41932653
$8 4 \longdiv { 2 3 } 3 1 5 8$
6259


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2653
3158
6497
4193
8423
6259

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$+$

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## 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.


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Time required:
$\mathrm{O}(\log n)$ 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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## Fracturing Packets

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- We can extract-min by fracturing the packet containing the minimum and adding the fragments back in.
- Runtime is $\mathrm{O}(\log n)$ 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 $2^{0}, 2^{1}, 2^{2}, 2^{3}, \ldots, 2^{k-1}$ nodes.
- Question: How can we represent our packets to support the above operations efficiently?


## Binomial Trees

- A binomial tree of order $\boldsymbol{k}$ 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 $0,1,2, \ldots, k-1$.

- Here are the first few binomial trees:

0


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


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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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## The Binomial Heap

- A binomial heap is a collection of binomial trees stored in ascending order of size.
- Operations defined as follows:
- meld $\left(p q_{1}, p q_{2}\right)$ : Use addition to combine all the trees.
- Fuses $\mathrm{O}(\log n+\log m)$ trees. Cost: $\mathrm{O}(\log n+\log m)$. Here, assume one binomial heap has $n$ nodes, the other $m$.
- pq.enqueue( $v, k$ ): Meld $p q$ 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 $n$ ).



























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


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


9

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


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## 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: $\mathrm{O}(\log n)$
- find-min: $\mathrm{O}(\log n)$
- extract-min: $\mathrm{O}(\log n)$
- meld: $\mathrm{O}(\log m+\log n)$.


## 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

Binomial Heap

- enqueue: $\mathrm{O}(\log n)$
- find-min: $\mathrm{O}(\log n)$
- extract-min: $\mathrm{O}(\log n)$
- meld: $\mathrm{O}(\log m+\log n)$. perform $n$ enqueues and then $n$ extractmins in time $o(n \log n)$. This is impossible with comparison-based algorithms.


## Where We Stand

- We can't make both enqueue and extractmin 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


## Binomial Heap

- enqueue: $\mathrm{O}(\log n)$
- find-min: $\mathrm{O}(\log n)$
- extract-min: $\mathrm{O}(\log n)$
- meld: $\mathrm{O}(\log m+\log n)$. remove all added items.
- Goal: Make enqueue take time $\mathrm{O}(1)$.


## 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 $\mathrm{O}(1)$.
- How could we accomplish this?


## 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

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## 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?



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3
4

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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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?
(3)


1

0


## 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 $\mathrm{O}(t \log t)$ 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 $0,1,2, \ldots,\left\lceil\log _{2}(n+1)\right\rceil$.
- 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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| Order 3 | Order 2 | Order 1 | Order 0 |
| :--- | :--- | :--- | :--- |



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## Analyzing Coalesce

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


## The Story So Far

- A binomial heap with lazy melding has these worst-case time bounds:
- enqueue: $\mathrm{O}(1)$
- 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.



## An Amortized Analysis

- Recall: We assign amortized costs as

$$
\text { amortized-cost }=\text { real-cost }+k \cdot \Delta \Phi
$$

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.



## 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: $\mathrm{O}(1) . \Delta \Phi:+1$
- Amortized cost: O(1).



## Analyzing a Meld

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



## Analyzing a Meld

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



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



## Analyzing extract-min

Find tree with minimum key.

$$
\begin{gathered}
\text { Work: } \mathrm{O}(t) \\
\Phi=t
\end{gathered}
$$



Remove min. Add children to list of trees.

Work: O(log $n$ )

Run the coalesce algorithm.

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

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Work: O(log $n$ )

Run the coalesce algorithm.

Work: $\mathrm{O}(t+\log n)$
$\Phi=\mathrm{O}(\log n)$
Amortized cost: $\mathbf{O}(\log \boldsymbol{n})$.

## 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 $t+\mathrm{O}(\log n)$.
- The runtime for coalescing these trees is $\mathrm{O}(t+\log n)$.
- When we're done merging, there will be O(log $n$ ) trees remaining, so $\Delta \Phi=-t+\mathrm{O}(\log n)$.
- Amortized cost is

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

## 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 worstcase efficient.
- This is possible! Check out bootstrapped skew

Lazy Binomial Heap

- Insert: $\mathrm{O}(1)$
- 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.

