CSE 373: Binary heaps

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Course overview

The course so far...

- ► Reviewing manipulating arrays and nodes
- ► Algorithm analysis
- ► Dictionaries (tree-based and hash-based)

Coming up next:

- ▶ Divide-and-conquer, sorting
- ▶ Graphs
- ► Misc topics (P vs NP, more?)

Timeline

When are we getting project grades/our midterm back?

Tuesday or Wednesday

Timeline

Do we have something due soon?

- ▶ Project 3 will be released today or tomorrow
- ► Due dates:
- ► Part 1 due in two weeks (Fri, Feb 16)

► Full project due in three weeks (Fri, Feb 23)

► Partner selection

- ► Selection form due Fri, Feb 9
- ► You MUST find a new partner..
- ...unless both partners email me and petition to stay together

Today

Motivating question:

Suppose we have a collection of "items".

We want to return whatever item has the smallest "priority".

The Priority Queue ADT

Specifically, want to implement the Priority Queue ADT:

The Priority Queue ADT

A priority queue stores elements according to their "priority". It supports the following operations:

- removeMin: return the element with the smallest priority
- ► peekMin: find (but do not return) the smallest element
- insert: add a new element to the priority queue

The Priority Queue ADT

supports the following operations:

An alternative definition: instead of yielding the element with the largest priority, yield the one with the *largest* priority:

The Priority Queue ADT, alternative definition
A priority queue stores elements according to their "priority". It

► removeMax: return the element with the largest priority

 \blacktriangleright peekMax: find (but do not return) the largest element

► insert: add a new element to the priority queue

The way we implement both is almost identical – we just tweak how we compare elements

In this class, we will focus on implementing a "min" priority queue

Initial implementation ideas

Fill in this table with the worst-case runtimes:

Idea	removeMin	peekMin	insert
Unsorted array list	(n)	(n)	Θ(1)
Unsorted linked list	$\Theta(n)$	$\Theta(n)$	$\Theta(1)$
Sorted array list	$\Theta(1)$	$\Theta(1)$	$\Theta(n)$
Sorted linked list	$\Theta(1)$	$\Theta(1)$	$\Theta(n)$
Binary tree	$\Theta(n)$	$\Theta(n)$	$\Theta(\log(n))$
A\/I +===	Q (1(-1)	Q (l==(+))	(A (lum(m))

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Initial implementation ideas

We want something optimized both frequent inserts and removes.

An AVL tree (or some tree-ish thing) seems good enough... right?

Today: learn how to implement a binary heap.

peekMin is $\mathcal{O}(1)$, and insert and remove are still $\mathcal{O}(\log(n))$ in the worst case.

However, insert is $\mathcal{O}\left(1\right)$ in the average case!

Binary heap invariants

Idea: adapt the tree-based method

Insight: in a tree, finding the min is expensive! Rather then having it to the left, have it on the top!

A BST or AVL tree



A binary heap

Binary heap invariants

We now need to change our invariants...

Binary heap invariants

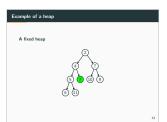
- A binary heap has three invariants:
- Num children: Every node has at most 2 children
- ► Heap: Every node is smaller then its children
- ► Structure: Every heap is a "complete" tree it has no "gaps"

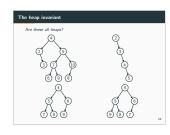
Example of a heap

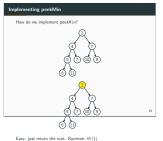
A broken heap



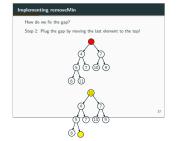
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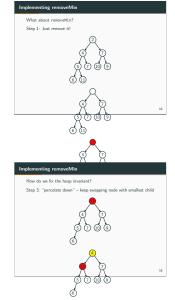


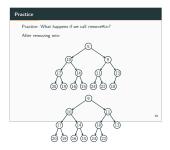




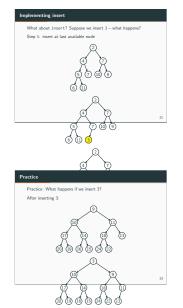
Lasy. Just return the root. Nuntine. () (1

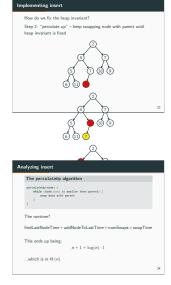












Analyzing removeMin, part 2

Problem: But wait! I promised worst-case $\Theta(\log(n))$ insert and average-case $\Theta(1)$ insert.

This algorithm is $\Theta(\log(n))$ in both the worst and average case!

Why: Finding and modifying the last node is slow: requires traversal!

Can we speed it up?

...

Analyzing removeMin, part 2

Remember this slide?

Idea	removeMax	peekMax	insert
Unsorted array list	(n)	(n)	Θ (1
Unsorted linked list	$\Theta(n)$	$\Theta(n)$	⊕ (1
Sorted array list	$\Theta(1)$	$\Theta(1)$	⊕ (n
Sorted linked list	$\Theta(1)$	$\Theta(1)$	⊕ (n
Binary tree	$\Theta(n)$	$\Theta(n)$	$\Theta(\log(n))$
AVL tree	$\Theta(\log(n))$	$\Theta(\log(n))$	$\Theta(\log(n))$

Analyzing removeMin, part 2

Observation:

- Arrays let us find and append to the end quickly
- ► Trees let us have nice log(n) traversal behavior

The trick: Why pick one or the other? Let's do both!

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The array-based representation of binary heaps

How do we find parent? $parent(i) = \left| \frac{i-1}{2} \right|$

The left child? $\mathsf{leftChild}(i) = 2i + 1$

The right child? leftChild(i) = 2i + 2

And fill an array in the level-order of the tree:



Finding the last node

If our tree is represented using an array, what's the time needed to find the last node now?

 Θ (1): just use this.array[this.size - 1]. ...assuming array has no 'gaps'. (Hey, it looks like the structure invariant was useful after all)

Re-analyzing insert

How does this change runtime of insert?

Runtime of insert:

 $findLastNodeTime + addNodeToLastTime + numSwaps \times swapTime$

...which is:

 $1 + 1 + numSwaps \times 1$

Observation: when percolating, we usually need to percolate up a few times! So, numSwaps ≈ 1 in the average case, and numSwaps \approx height $= \log(n)$ in the worst case!

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Re-analyzing removelylin		
How does this change runtime of renoveMin?		
Runtime of removeMin:		
$findLastNodeTime + removeRootTime + numSwaps \times swapTime$		
which is: $1+1+numSwaps\times 1$		
1 + 1 + num5waps × 1		
Observation: unfortunately, in practice, usually must percolate all		
the way down. So numSwaps $pprox$ height $pprox \log(n)$ on average.		
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