6.006 Recitation

Build 2008.16

Coming up next...

Sorting

- Scenic Tour: Insertion Sort, Selection Sort, Merge Sort
- New Kid on the Block: Merge Sort
- Priority Queues
 - Heap-Based Implementation

Sorting

• Input: array **a** of **N** keys

- Output: a permutation as of a such that
 as[k] < as[k+1]
- Stable sorting:

Sorting

- Maybe the oldest problem in CS
- Reflects our growing understanding of algorithm and data structures
- Who gives a damn?
 - All those database tools out there

Sorting Algorithms: Criteria

What	Why
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Speed	That's what 6.006 is about
Auxiliary Memory	External sorting, memory isn't that cheap
Simple Method	You're learning / coding / debugging / analyzing it
# comparisons, data moving	Keys may be large (strings) or slow to move (flash memory)

Insertion Sort

- Base: a[0:1] has 1
 element ⇒ is sorted
- Induction: a[0:k] is sorted, want to grow to a[0:k+1] sorted
 - find position of a[k+1] in a[0:k]
 - insert a[k+1] in a[0:k]

5	8	2	7	I	4	3	6
5	8	2	7	Ι	4	3	6
2	5	8	7	Ι	4	3	6
2	5	7	8	Ι	4	3	6
Ι	2	5	7	8	4	3	6
I	2	4	5	7	8	3	6
	2	4	5	7	8	3	6
	2	3	4	5	6	7	8

Insertion Sort: Costs

- Find position for a[k+1] in a[0:k] - O(log(k))
 - use binary search
- Insert a[k+1] in a[0:k]:
 O(k)
 - shift elements
- Total cost: $O(N \cdot log(N))$ + $O(N^2) = O(N^2)$

- Pros:
 - Optimal number of comparisons
 - O(I) extra memory (no auxiliary arrays)
- Cons:
 - Moves elements around a lot

Selection Sort

- Base case: a[0:0] has the smallest 0 elements in a
- Induction: a[0:k] has the smallest k elements in a, sorted; want to expand to a[k+1]
 - find min(a[k+1:N])
 - swap it with a[k+1]

5	8	2	7		4	3	6
I	8	2	7	5	4	3	6
I	2	8	7	5	4	3	6
Ι	2	3	7	5	4	8	6
Ι	2	3	4	5	7	8	6
Ι	2	3	4	5	7	8	6
I	2	3	4	5	6	8	7
	2	3	4	5	6	7	8

Selection Sort: Costs

- find minimum in a[k+1:N]) - O(N-k)
 - scan every element
- swap with a[k] O(I)
 - need help for this?
- Total cost: $O(N^2) + O(N) = O(N^2)$

- Pros:
 - Optimal in terms of moving data around
 - O(I) extra memory (no auxiliary arrays)

• Cons:

• Compares a lot

Merge-Sort

I. Divide

- Break into 2 sublists
- 2. Conquer
 - I-elements lists are sorted
- 3. Profit
 - Merge sorted sublists

5	8	2	7		4	3	6
5	8	2	7		4	3	6
5	8	2	7	1	4	3	6
2	5	7	8	1	3	4	6
Ι	2	3	4	5	6	7	8
	Tł	nere	e is	no	step	o 6	
	Th	nere	e is	no	step	7	
	Th	nere	e is_	no	ster	8	

Merge-Sort: Cost

- You should be ashamed of if you don't know!
- $T(N) = 2T(N/2) + \Theta(N)$
- Recursion tree
 - O(log(N)) levels,
 O(N) work / level
- Total cost: $O(N \cdot log(N))$

- Pros:
 - Optimal number of comparisons
 - Fast
- Cons:
 - O(N) extra memory (for merging)

BST Sort

- Build a BST out of the keys
- Use inorder traversal to obtain the keys in sorted order
 - Or go to minimum(), then call successor() until it returns None



BST Sort: Cost

- Building the BST -O(N · log(N))
 - Use a balanced tree
- Traversing the BST O(N)
 - Even if not balanced
- Total cost: $O(N \cdot log(N))$

- Pros:
 - Fast (asymptotically)
- Cons:
 - Large constant
 - O(N) extra memory (left/right pointers)
 - Complex code

Best of Breed Sorting

Speed

 $O(N \cdot log(N))$

Auxiliary Memory

O(I)

Code complexity

Simple

Comparisons

 $O(N \cdot log(N))$

Data movement

O(N)

Heap-Sort

Speed	$O(N \cdot log(N))$	\checkmark
Auxiliary Memory	O(I)	\checkmark
Code complexity	Simple	\checkmark
Comparisons	$O(N \cdot \log(N))$	\checkmark
Data movement	O(N)	X

Heap-Sort uses a... Heap (creative, eh?)

- Max-Heap DT
 - Almost complete binary tree
 - Root node's key >= its children's keys
 - Subtrees rooted at children are Max-Heaps as well



Max-Heap Properties

- Very easy to find max. value
 - look at root, doh
- Unlike BSTs, it's very hard to find any other value
 - 6 (3rd largest key) at same level as I (min. key)



- THIS IS WHY HEAPS ROCK OVER BSTs
 - No need to store a heap as a binary tree (left, right, parent pointers)
- Store keys inside array, in level-order traversal



- Work with arrays, think in terms of trees
 - Left subtree of 8 is in bold... pretty mindboggling, eh?
 - Prey that you don't have to debug this



• root index: I

• left_child(node_index):

node_index · 2

• right_child(node_index):

• node_index $\cdot 2 + 1$

• parent(node_index):

L node_index / 2]



• How to recall this

- I. draw the damn heap (see right)
- remember the concept (divide / multiply by 2)
- 3. work it out with the drawing



Heaps Inside Arrays: Python Perspective

- Lists are the closest thing to array
- Except they grow
 - Just like our growing hashes
 - Amortized O(I) per operation

	2	3	4	5	6	7	8
7	5	8	3		4	6	2

Messing with Heaps



- I. Change any key
- 2. Restore Max-Heap invariants



• Issue

- key's node becomes smaller than children
- only possible after decreasing a key
- Solution
 - percolate (huh??)



- Percolate:
 - swap node's key with max(left child key, right child key)
 - Max-Heap restored locally
 - the child we didn't touch still roots a Max-Heap



Percolate

- Issue: swapping decreased the key of the child touched
 - child might not root a Max-Heap
- Solution: keep percolating



• Percolating is finite:

- leaves are always Max-Heaps
- Percolate cost:
 - O(heap height node's level)
 - O(log(N) log(node))



• Issue

- key's node becomes larger than parent
- only possible after increasing a key
- Solution
 - sift (huh??)



• Sift

- swap node's key with parent's key
- parent's key was >= node's key, so must be >= children keys
- Max-Heap restored for node's subtree





- Issue: swapping increased the key of the parent
 - parent might not root a Max-Heap
- Solution: keep sifting



- Sifting is finite:
 - root has no parent, so it can be increased at will
- Sift cost:
 - O(height)
 - O(log(node))



Messing with Heaps

Update(node, new_key)

- old_key ← heap[node].key
- heap[node].key new_key
- if new_key < old_key: sift(node)</p>
- else: percolate(node)

Messing with Heaps II

• Goal

- Want to shrink or grow the heap
- Growing:
 - inserting keys
- Shrinking:
 - deleting keys



Messing with Heaps II: One More Node

- Can always insert -∞ at the end of the heap
- Max-Heap will not be violated
 - Can only add to the end, otherwise we wouldn't get an (almost) complete binary tree



Messing with Heaps II: One More Node

- Insert any key
 - insert -∞ at the end of the heap
 - change node's key to desired key
 - sift



Messing with Heaps II: One More Node

Insertion cost

- insert -∞ at the end of the heap - O(I)
- change node's key to new key - O(I)
- sift O(log(N))
- Total cost: O(log(N))



- Can always delete last node
- Max-Heap will not be violated
 - It must be the last node, otherwise the binary tree won't be (almost) complete



Messing with Heaps II: One More Less Node



- Replace root key with last key
- Delete last node
- Percolate



- Deleting root cost
 - Replace root key with last key - O(I)
 - Delete last O(I)
 - Percolate O(log(N))
- Total cost: O(log(N))



- Deleting any node
 - Change key to $+\infty$
 - Sift
 - Delete root



- Deletion cost
 - Change key to +∞ O(1)
 - Sift O(log(N))
 - Remove root O(log(N))

Total cost: O(log(N))



Heap-Sort: Everything Falls Into Place

- Start with empty heap
- Build the heap: insert a[0] ... a[N-1]
- Build the result: delete root until heap is empty, gets keys sorted in reverse order
- Use a to store both the array and the heap (explained in lecture)

Heap-Sort: Slightly Faster

- Build the heap faster: Max-Heapify
 - Explained in lecture
 - O(N) instead of $O(N \cdot \log(N))$
- Total time for Heap-Sort stays O(N·log(N)) because of N deletions
- Max-Heapify is very useful later

Priority Queues

- Data Structure
 - insert(key) : adds to the queue
 - **max**() : returns the maximum key
 - **delete-max**() : deletes the max key
 - **delete**(key) : deletes the given key
 - optional (only needed in some apps)

Priority Queues with Max-Heaps

- Doh? (assuming you paid attention so far)
- Costs (see above line for explanations)
 - insert: O(log(N))
 - max: O(I)
 - delete-max: O(log(N))
 - delete: O(log(N)) only if given the index of the node containing the key

Cool / Smart Problem

- Given an array **a** of numbers, extract the **k** largest numbers
- Want good running time for any **k**

Cool / Smart Problem

• Small cases:

- k = 1: scan through the array, find N
- k small
 - try to scale the scan
 - getting to O(kN), not good

Cool / Smart Problem

- Solution: Heaps!
 - build heap with Max-Heapify
 - delete root k times
 - $O(k \cdot log(N))$
- Bonus Solution: Selection Trees (we'll come back to this if we have time)

Discussion: Priority Queue Algorithms

- BSTs
 - store keys in a BST
- Regular Arrays
 - store keys in an array
- Arrays of Buckets
 - a[k] stores a list of keys with value k

And we're done!

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v. Next

- Get sleep (unmitigated disaster)
- Definitely can't teach both heaps and sorting
 - Convert heaps to a problem, since they should know the basics from lecture
 - Make sorting shorter