CSE 373

OCTOBER 9TH - AMORTIZED ANALYSIS

TODAY

- Master Theorem
- Amortized Analysis
- Binary Search Trees

- Algorithm Analysis
 - Asymptotic behavior

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- Loops and iterations

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- Recursive functions

- Algorithm Analysis
 - Asymptotic behavior
- Loops and iterations
- Recursive functions
 - Recurrence relations

- On Friday, we showed the formal recurrence approach
 - Break into recursive, non-recursive
 - Compute non-recursive computation time
 - Produce the recurrence
 - Roll out the recurrence and produce the closed form
 - Upper-bound the closed form with bigO notation



- While this process is important, we can save some steps if all we care about is the upper bound
 - bigO notation eliminates the need for constants
 - Lots of our messing around with c₀ and c₁ doesn't come through to the solution



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Master theorem

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- Therefore, if you can produce a recurrence, there is actually a procedural way to produce solutions
- If T(n) = a*T(n/b)+n^c for n > n₀ and if the base case is a constant
 - Case 1: $\log_{b}(a) < c$: T(n) = O(n^c)
 - Case 2: $\log_{b}(a) = c$: T(n) = O(n^c Ig n)
 - Case 3: $\log_{b}(a) > c$: $T(n) = O(n^{\log a})$



• Recurrences come up all the time

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- Analyze methods and iterative approaches through the normal methods
- Recursive functions use a recurrence
- Possible to get to bigO solution quickly
- Usually for worst-case analysis





- Final analysis type
 - Worst-case



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 - Consider adding to an unsorted array
 - Resizing is the costly O(n) operation
 - This occurs in predictable ways
 - Do these types of operations really slow down the function?

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 - Doubling the array

- Adding to unsorted array
 - What if we only add some constant number to the array?
 - Let's resize and add 10,000 elements every time
 - How long does it take to add *n* elements?
 - n-n/10,000*O(1) + n/10,000*O(n)

- Adding to unsorted array
 - What if we only add some constant number to the array?
 - Let's resize and add 10,000 elements every time
 - How long does it take to add *n* elements?
 - $n-n/10,000*O(1) + n/10,000*O(n) = O(n^2)$
 - This is for any constant, regardless of how large

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 - If the worst case happens in predictable ways (i.e. every *n* inserts), then the costly operation doesn't increase the total asymptotic runtime of multiple operations
 - Over *n* operations, remember to divide the total runtime by *n*

DICTIONARIES

- Back to the dictionary problem
 - Can we apply these analytical tools to some simple implementations?

IMPLEMENTATIONS

Simple implementations

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Simple implementations

Unsorted linked-list	insert	find	delete
	O(n)*	O(n)	O(n)
Unsorted array	O(n)*	O(n)	O(n)
Sorted linked list	O(n)	O(n)	O(n)
Sorted array	O(n)	O(log n)	O(n)

* Because we need to check for duplicates

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 - A rooted tree, where each node has at most two children
 - All elements less than the root are in the left subtree and all elements larger than the root are in the right subtree
 - All, subtrees must also be binary search trees
- With this property, all binary search trees have sorted in-order traversals

- Other implementations?
 - Binary Search Tree (BST)
 - Sort based on keys (which have to be comparable)
 - How do we implement this?
 - What changes need to be made?

- BST Node:
 - Before:

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 - Node left
 - Node right
 - Value data

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 - Value v

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 - All keys are unique, i.e. each key only has one value

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 - Insert()? Delete()? Find()?
 - Consider best and worst-case.
 - What are the inputs for best and worst-case?

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 - Insert():
 - Worst case: O(n). What is this worst case?

- BST Analysis:
 - Insert():
 - Worst case: O(n)
 - Best case:

- BST Analysis:
 - Insert():
 - Worst case: O(n)
 - Best case: O(1)
 - What is the general case here?
 - What does the runtime for a particular insert depend on?
 - Height of the tree

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 - Height = 1 + max(height(left),height(right))

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 - Generally, however: O(log n) when the tree is balanced

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 - Case 3: The node has two children
 - What are some possible strategies?

- Deleting nodes with 2 children
 - How do we delete 12?



- How do we delete 12?
- Can we replace 12 with one of it's children?



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- Can we replace 12 with one of it's children?
- Need to find candidate to replace 12



Deleting nodes with 2 children

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- In order to avoid changing the shape and doing too much work, it must be either the predecessor (the element just before it in sorted order) or the successor (the element just after it in sorted order)

Deleting nodes with 2 children

• What are the predecessor and successor of 12?



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- BST Analysis:
 - Delete():
 - Worst case(): O(n), finding the predecessor/successor takes time. What is this case?
 - Best case(): O(1) if we're deleting the root from a degenerate tree
 - "Degenerate" trees are those that are very unbalanced.



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- Many of our worst cases are when trees are poorly balanced
- Can we enforce this balance?
- What are some possible balance conditions?
 - Number of elements on the left = number on right?
 - What we really care about though is the height of the tree
 - Height of the left = height on the right?



• This doesn't help much



This doesn't help much

- Need the definition to be recursive
- Let height(left) = height(right) for all nodes



• Now what's wrong?



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 - Only perfect trees (with 2^k children) can exist





• For each node in the tree, the height of its left and right subtrees can differ by at most one



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- $|(height(left) height(right)| \le 1$



- For each node in the tree, the height of its left and right subtrees can differ by at most one
- $|(height(left) height(right)| \le 1$
- This is the AVL property, and we can use it to create self balancing trees