Lecture 8. Binary Search Trees

CpSc 8400: Algorithms and Data Structures
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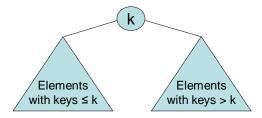
School of Computing Clemson University Spring, 2016

Dictionary / Set Data Structures

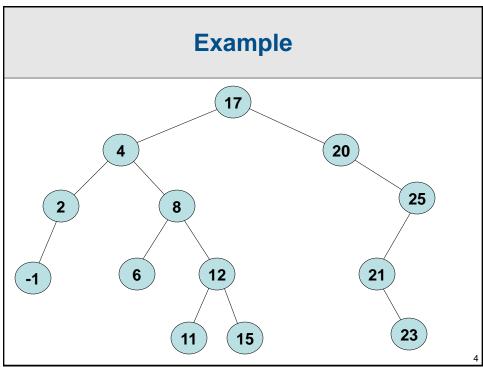
- A dictionary maintains a set of elements with associated keys, supporting the operations:
 - find(k): return a pointer to the element having key k, or indicates that such an element does not exist.
 - insert(e, k): insert a new element e with key k.
 - delete(e): delete element e, given a <u>pointer</u> to e.
 To delete an element given its key: delete(find(k)).
- Common dictionary implementations:
 - Arrays and linked lists: not very efficient (at least one of the above operations takes $\Theta(n)$ worst-case time).
 - Balanced BSTs: all ops take O(log n) time.
 - Universal Hash tables: O(1) (expected) time per op.

Binary Search Trees

 A BST is a binary tree satisfying the binary search tree property:



- Each node typically maintains a pointer to its parent, left child, and right child.
- Tree is **balanced** if height = O(log n).

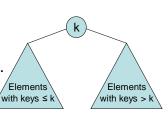


Fundamental Operations

Most BST operations can be implemented in a simple recursive fashion in O(height) time (O(log n) if balanced). For example:

• Find(T, k):

if T == NULL, then return NULL.
if k == T.key, then return T.
if k < T.key, then return Find(T.left, k).
else return Find(T.right, k).</pre>



Insert(T, e, k):

if T == NULL, then return NewNode(e, k). if k < T.key, then T.left = Insert(T.left, e, k). else T.right = Insert(T.right, e, k). return T.

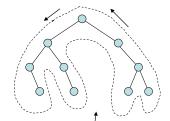
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Tree Traversals

 We often enumerate the contents of an n-element BST in O(n) time using an inorder tree traversal:

Inorder(T):

if T == NULL, then return.
Inorder(T.left).
print T.key.
Inorder(T.right).



- Other types of common traversals:
 - Preorder: root, left subtree, right subtree.
 - Postorder: left subtree, right subtree, root.
 - Eulerian: root, left subtree, root, right subtree, root.
 (sequence of nodes encountered when we "walk around" a BST)

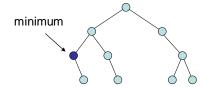
Sorting with a BST

- An inorder traversal prints the contents of an n-element BST in sorted order in O(n) time.
- Therefore, we can use BSTs to sort: *insert* n elements, then do an inorder traversal.
- Since the BST is a comparison-based data structure, this means *insert* must run in Ω(log n) time in the worst case; otherwise we could sort faster than O(n log n) in the comparison model.

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Minimum and Maximum

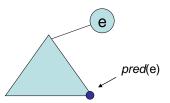
- It's easy to locate the minimum and maximum elements in a BST in O(h) time.
- For the minimum, start at the root and walk left as long as possible:



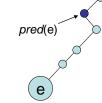
· Vice-versa for the maximum...

Predecessor and Successor

• The *pred*(e) operation takes a pointer to e and locates the element immediately preceding e in an inorder traversal.



If e has a left child, then pred(e) is the maximum element in e's left subtree.



If e has no left child, then pred(e) is the first "left parent" we encounter when walking from e up to the root.

 The successor operation, succ(e), is completely symmetric to this.

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Deletion

- It's easy to delete an element with zero or one children.
- To delete an element e with two children, first swap e with either pred(e) or succ(e), then proceed to delete e.
 - Note that if e has two children, then pred(e) and succ(e) can both have at most one child.
 - Also note that replacing e with pred(e) or succ(e) is o.k. according to the BST property.
- Can also replace e with the join of e's two subtrees.

Using pred and succ for Inexact Searches

- In addition to pred(e) and succ(e) that take pointers to elements, we could implement:
 - pred(k): find the element with largest key ≤ k.
 - succ(k): find the element with smallest key ≥ k.
- This allows us to perform inexact searches: if an element with key k isn't present, we can at least find the closest nearby elements in the BST.
- This is one reason a BST is so powerful. Other dictionary data structures, notably hash tables, don't support pred / succ and cannot perform inexact searches.

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Paging Through Nearby Elements

- Starting from element e, we can use *pred* / succ to find elements near e very quickly.
- Prototypical application: library catalog.
 - After searching for an author's name, you are presented with a alphabetized list in which you can scroll through nearby names.
- Can also answer range queries: output all elements whose keys are in the range [a, b].
 - Start at the element e = succ(a). Then repeatedly call succ(e) until you reach b.
 - Running time O(k + log n), where k = # of elements written as output.

Rank-Based Access

- A BST augmented with subtree sizes can support two additional useful operations:
 - select(k): return a pointer to the kth largest element in the tree (k = 1 is the min, k = n is the max, and k = n/2 is the median).
 - rank(e): given a pointer to e, returns the number of elements with keys ≤ e's key.
 (that is, e's index within an inorder traversal).
- The element of rank k in a set is also called the kth order statistic of the set. Accordingly, the CLRS textbook calls a BST augmented this way an order statistic tree.

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Select and Rank

Select(T, k):

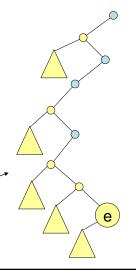
r = size(T.left) + 1. $(r = rank \ of \ root)$ if k = r, then return T. if k < r, then return Select(T.left, k).

else, return Select(T.right, k – r).

Rank(e):

Add up the total # of yellow elements as we walk from e up to the root.

(these are all the elements less than e)



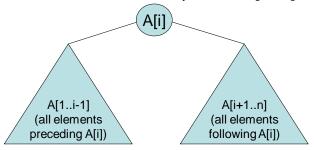
Dynamic Sequences

- Suppose we use an array or linked list to encode a sequence of n elements.
 - We can insert/delete at the endpoints of the sequence in O(1) time,
 - But insertions/deletions in the middle of the sequence take O(n) worst-case time.
- Using a balanced BST augmented with subtree sizes, we can perform all the following operations in O(log n) time:
 - Insert / delete anywhere in the sequence.
 - Access or modify any element by its index in the sequence.

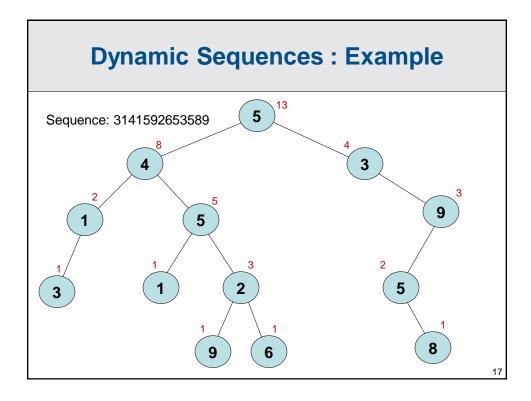
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Dynamic Sequences

 The "BST property" is slightly different when we use a BST to encode a sequence A[1..n]:



• Elements no longer have keys, and we no longer use *find*. Rather, we rely on the *select* to access elements based their index within the sequence.

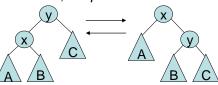


Dynamic Sequences

- Many operations have a nice meaning:
 - select(i) accesses the ith element
 - An inorder traversal prints the sequence in order
 - succ and pred allow us to move between successive elements (as in a linked list)
 - min and max jump to the beginning and end
- You might want to think of a BST as a structure that fundamentally encodes an arbitrary sequence from left to right. In the case of a dictionary, this sequence happens to be the sorted ordering of the elements we are storing.

Balancing a BST

- There are many ways to modify the insert and delete operations of a BST to keep it balanced:
 - Worst-case mechanisms: h = O(log n) always.
 AVL trees, BB[α] trees, red-black trees
 - Amortized mechanisms: starting with an empty tree, any sequence of m operations takes O(m log n) time.
 splay trees, "batch" rebalancing methods
 - Randomized mechanisms: always balanced "with high probability". randomly-balanced BSTs, treaps
- Most of these are based on rotations:



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Example of Balancing with Rotations: AVL Trees

- AVL: <u>A</u>del'son-<u>V</u>el'skiĭ, and <u>L</u>andis '62.
- A binary tree is height balanced if:
 - Its left and right subtrees differ in height by at most 1, and
 - Its left and right subtrees are themselves height balanced.
- It's easy to show that height-balanced trees are balanced (proof on next slide).
- An AVL tree is a height-balanced tree where every node is augmented with the height of its subtree.
 - Note: easy to keep this info up to date after a rotation.
- After each insertion or deletion, we restore the height balance property by performing O(log n) carefully chosen rotations.

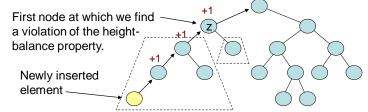
Height Balanced → **Balanced**

- Claim: A height-balanced tree of height h contains ≥ F_h nodes, where F_h is the hth Fibonacci number.
- Recall that $F_h = \Theta(\Phi^h)$, where $\Phi = (1+\sqrt{5})/2$. Therefore, an n-element height-balanced tree can have height at most $O(\log n)$.
- · Easy proof by induction:
 - Consider any arbitrary height-balanced tree of height h.
 - Without loss of generality, suppose its left subtree is tallest.
 - By induction, left subtree contains ≥ F_{h-1} elements.
 - By induction, right subree contains ≥ F_{h-2} elements.
 - So total # of elements is $\ge 1 + F_{h-1} + F_{h-2} = 1 + F_h$.
 - Don't forget the base cases:
 - •Tree of height 0 contains \geq 1 element (F₀ = 0).
 - •Tree of height 1 contains \geq 2 element (F₁ = 1).

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Restoring Balance After an Insertion

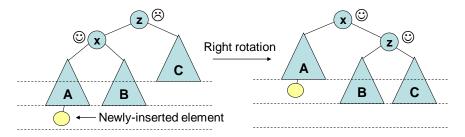
- After an *insert*, walk back up the insertion path and increment augmented subtree heights.
- Stop at the first node (z in the figure below) at which we encounter a height-balancing violation (a height imbalance of exactly 2).



- Using 1 or 2 rotations, we will be able to rebalance our tree at z.
- The height of z's subtree will decrease by 1 to its original value, so we don't need to visit the remaining nodes on the insertion path above z.

Restoring Balance After an Insertion

- Let's suppose we have an imbalance at node z due to an insertion that makes z's left subtree too tall (the same argument will apply to the right subtree as well...)
- Easy case: Insertion into subtree A makes it too tall:

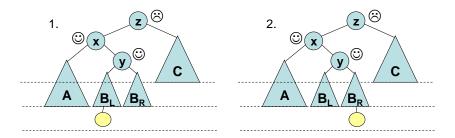


Harder case: Insertion into subtree B makes it too tall...

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Restoring Balance After an Insertion

 Harder case: Subtree B too tall. There are now 2 possible subcases:



- In all both cases, a left rotation about the edge (x, y) brings us back to the previous case (with A being too tall).
- So 2 total rotations needed for these cases.