



CSE332: Data Abstractions

Lecture 6: Dictionaries; Binary Search Trees

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Where we are

Studying the absolutely essential ADTs of computer science and classic data structures for implementing them

ADTs so far:

```
1. Stack: push, pop, isEmpty, ...
```

2. Queue: enqueue, dequeue, isEmpty, ...

3. Priority queue: insert, deleteMin, ...

Next:

- 4. Dictionary (a.k.a. Map): associate keys with values
 - Probably the most common, way more than priority queue

The Dictionary (a.k.a. Map) ADT

Data: djg set of (key, value) pairs Dan keys must be comparable Grossman insert(djg,) Operations: trobison - insert(key,value) **Tyler** - find(key) **Robison** - delete(key) find(trobison) Tyler, Robison, ... **snwang Stanley** Will tend to emphasize the keys; Wang don't forget about the stored values

Comparison: The Set ADT

The Set ADT is like a Dictionary without any values

A key is *present* or not (no repeats)

For find, insert, delete, there is little difference

- In dictionary, values are "just along for the ride"
- So same data-structure ideas work for dictionaries and sets

But if your Set ADT has other important operations this may not hold

- union, intersection, is_subset
- Notice these are binary operators on sets

Dictionary data structures

Will spend the next several lectures implementing dictionaries with three different data structures

1. AVL trees

Binary search trees with guaranteed balancing

2. B-Trees

- Also always balanced, but different and shallower
- B!=Binary; B-Trees generally have large branching factor

3. Hashtables

Not tree-like at all

Skipping: Other balanced trees (e.g., red-black, splay)

But first some applications and less efficient implementations...

A Modest Few Uses

Any time you want to store information according to some key and be able to retrieve it efficiently

Lots of programs do that!

Networks: router tables

Operating systems: page tables

Compilers: symbol tables

Databases: dictionaries with other nice properties

Search: inverted indexes, phone directories, ...

Biology: genome maps

• ...

Simple implementations

For dictionary with *n* key/value pairs

insert find delete

- Unsorted linked-list
- Unsorted array
- Sorted linked list
- Sorted array

We'll see a Binary Search Tree (BST) probably does better, but not in the worst case unless we keep it balanced

Simple implementations

For dictionary with *n* key/value pairs

		insert	find	delete
•	Unsorted linked-list	<i>O</i> (1)	<i>O</i> (<i>n</i>)	O(<i>n</i>)
•	Unsorted array	<i>O</i> (1)	O(<i>n</i>)	O(<i>n</i>)
•	Sorted linked list	<i>O</i> (<i>n</i>)	<i>O</i> (<i>n</i>)	<i>O</i> (<i>n</i>)
•	Sorted array	<i>O</i> (<i>n</i>)	$O(\log n)$	<i>O</i> (<i>n</i>)

We'll see a Binary Search Tree (BST) probably does better, but not in the worst case unless we keep it balanced

Lazy Deletion

10	12	24	30	41	42	44	45	50
✓	æ	✓	√	✓	√	×	√	✓

A general technique for making delete as fast as find:

Instead of actually removing the item just mark it deleted

Plusses:

- Simpler
- Can do removals later in batches
- If re-added soon thereafter, just unmark the deletion

Minuses:

- Extra space for the "is-it-deleted" flag
- Data structure full of deleted nodes wastes space
- find O(log m) time where m is data-structure size (okay)
- May complicate other operations

Some tree terms (mostly review)

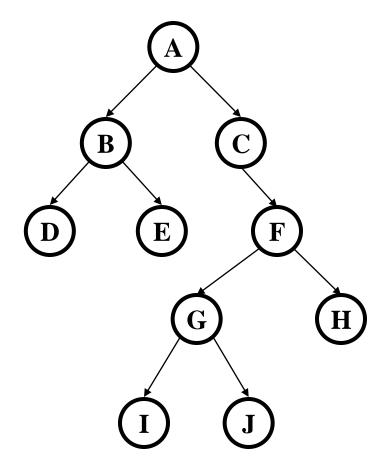
- There are many kinds of trees
 - Every binary tree is a tree
 - Every list is kind of a tree (think of "next" as the one child)
- There are many kinds of binary trees
 - Every binary min heap is a binary tree
 - Every binary search tree is a binary tree
- A tree can be balanced or not
 - A balanced tree with n nodes has a height of O(log n)
 - Different tree data structures have different "balance conditions" to achieve this

Binary Trees

- Binary tree is empty or
 - A root (with data)
 - A left subtree (may be empty)
 - A right subtree (may be empty)
- Representation:

Data				
left	right			
pointer	pointer			

 For a dictionary, data will include a key and a value



Binary Trees: Some Numbers

Recall: height of a tree = longest path from root to leaf (count edges)

For binary tree of height *h*:

- max # of leaves:
- max # of nodes:
- min # of leaves:
- min # of nodes:

Binary Trees: Some Numbers

Recall: height of a tree = longest path from root to leaf (count edges)

For binary tree of height *h*:

- max # of nodes:
$$2^{(h+1)} - 1$$

- min # of nodes:
$$h+1$$

For n nodes, we cannot do better than $O(\log n)$ height, and we want to avoid O(n) height

Calculating height

What is the height of a tree with root root?

```
int treeHeight(Node root) {
     ???
}
```

Calculating height

What is the height of a tree with root root?

Running time for tree with n nodes: O(n) – single pass over tree

Note: non-recursive is painful – need your own stack of pending nodes; much easier to use recursion's call stack

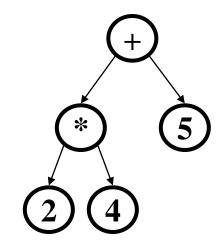
Tree Traversals

A traversal is an order for visiting all the nodes of a tree

Pre-order. root, left subtree, right subtree

• *In-order*. left subtree, root, right subtree

• Post-order: left subtree, right subtree, root



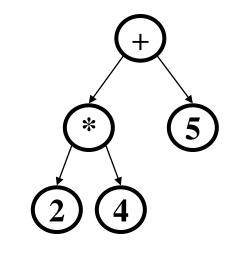
(an expression tree)

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Tree Traversals

A traversal is an order for visiting all the nodes of a tree

- Pre-order. root, left subtree, right subtree
 + * 2 4 5
- In-order. left subtree, root, right subtree
 2 * 4 + 5
- Post-order. left subtree, right subtree, root
 2 4 * 5 +



(an expression tree)

More on traversals

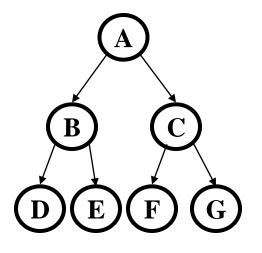
```
void inOrderTraversal(Node t) {
  if(t != null) {
    inOrderTraversal(t.left);
    process(t.element);
    inOrderTraversal(t.right);
  }
}
```

Sometimes order doesn't matter

Example: sum all elements

Sometimes order matters

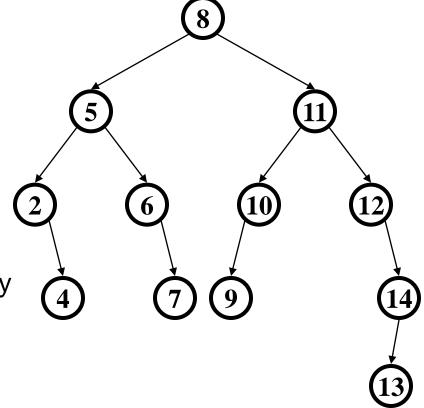
- Example: print tree with parent above indented children (pre-order)
- Example: evaluate an expression tree (post-order)



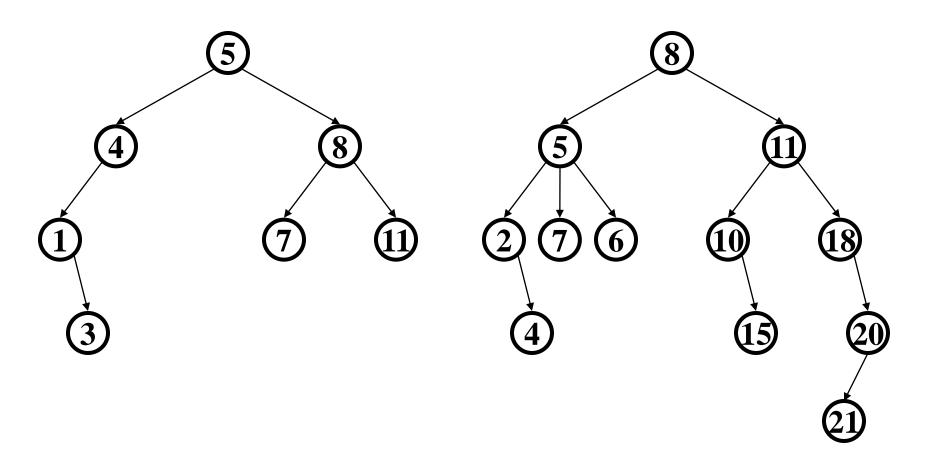
B D E C F G

Binary Search Tree

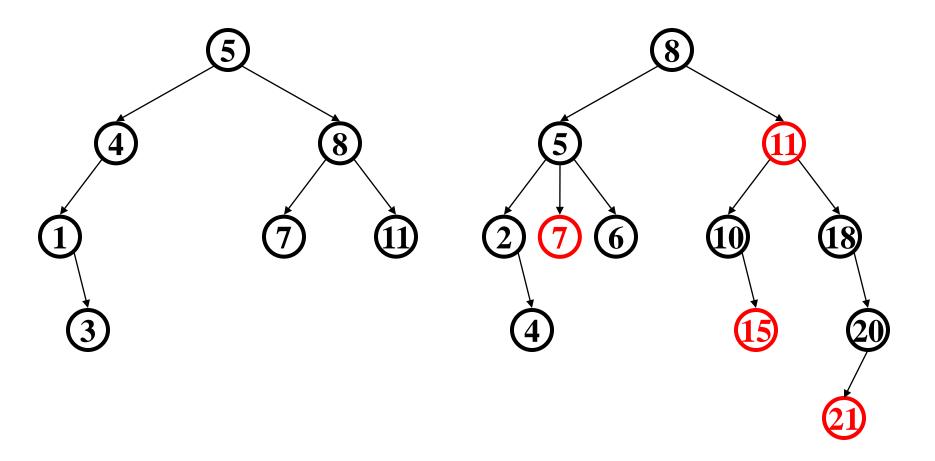
- Structure property ("binary")
 - Each node has ≤ 2 children
 - Result: keeps operations simple
- Order property
 - All keys in left subtree smaller than node's key
 - All keys in right subtree larger than node's key
 - Result: easy to find any given key



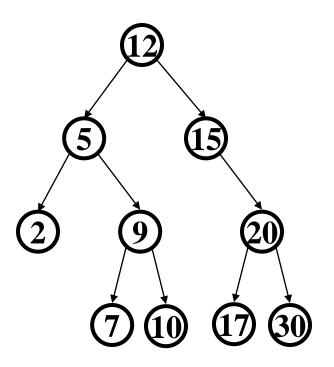
Are these BSTs?



Are these BSTs?

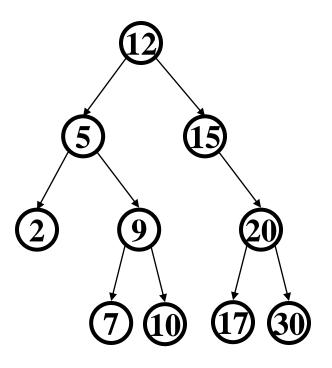


Find in BST, Recursive



```
Data find(Key key, Node root) {
  if(root == null)
    return null;
  if(key < root.key)
    return find(key,root.left);
  if(key > root.key)
    return find(key,root.right);
  return root.data;
}
```

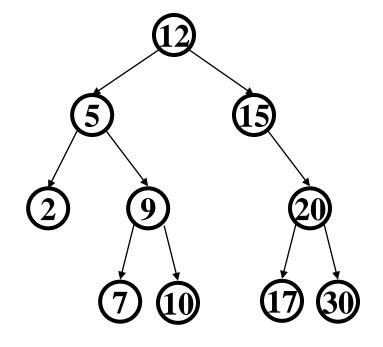
Find in BST, Iterative



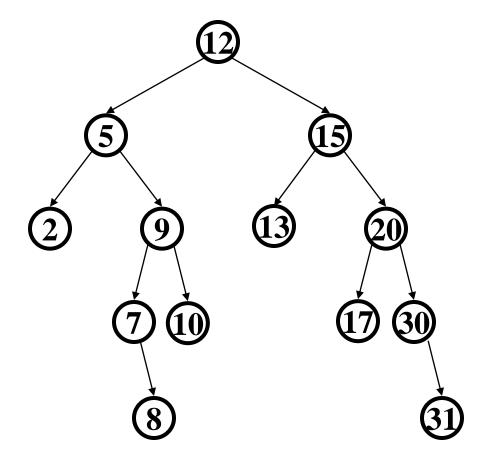
```
Data find(Key key, Node root) {
  while(root != null
          && root.key != key) {
    if(key < root.key)
        root = root.left;
    else(key > root.key)
        root = root.right;
  }
  if(root == null)
    return null;
  return root.data;
}
```

Other "Finding" Operations

- Find minimum node
 - "the liberal algorithm"
- Find maximum node
 - "the conservative algorithm"
- Find predecessor of a non-leaf
- Find successor of a non-leaf
- Find predecessor of a leaf
- Find successor of a leaf



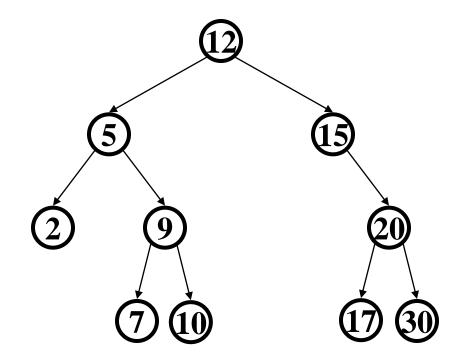
Insert in BST



insert(13)
insert(8)
insert(31)

(New) insertions happen only at leaves – easy!

Deletion in BST

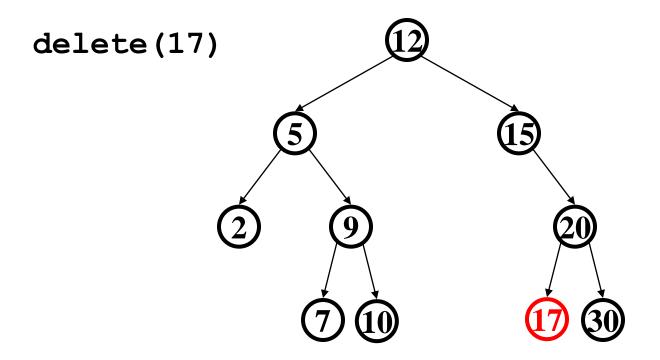


Why might deletion be harder than insertion?

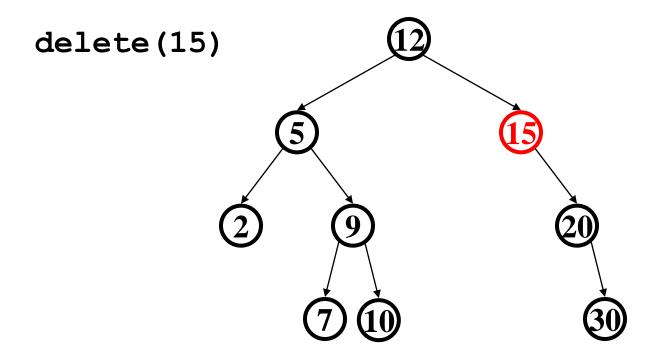
Deletion

- Removing an item disrupts the tree structure
- Basic idea: **find** the node to be removed, then "fix" the tree so that it is still a binary search tree
- Three cases:
 - Node has no children (leaf)
 - Node has one child
 - Node has two children

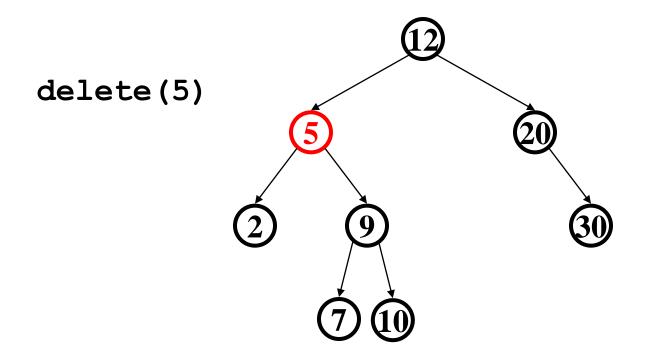
Deletion – The Leaf Case



Deletion - The One Child Case



Deletion - The Two Child Case



What can we replace 5 with?

Deletion – The Two Child Case

Idea: Replace the deleted node with a value guaranteed to be between the two child subtrees

Options:

- successor from right subtree: findMin(node.right)
- predecessor from left subtree: findMax(node.left)
 - These are the easy cases of predecessor/successor

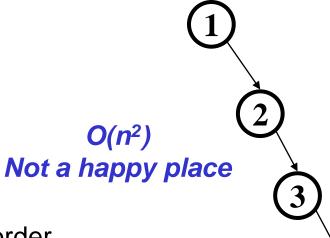
Now delete the original node containing *successor* or *predecessor*

Leaf or one child case – easy cases of delete!

BuildTree for BST

- We had buildHeap, so let's consider buildTree
- Insert keys 1, 2, 3, 4, 5, 6, 7, 8, 9 into an empty BST
 - If inserted in given order, what is the tree?
 - What big-O runtime for this kind of sorted input?

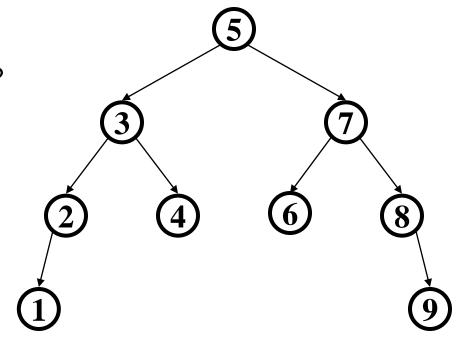
Is inserting in the reverse order any better?



BuildTree for BST

- Insert keys 1, 2, 3, 4, 5, 6, 7, 8, 9 into an empty BST
- What we if could somehow re-arrange them
 - median first, then left median, right median, etc.
 - -5, 3, 7, 2, 1, 4, 8, 6, 9
 - What tree does that give us?
 - What big-O runtime?

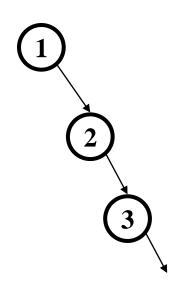
O(n log n), definitely better



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Unbalanced BST

- Balancing a tree at build time is insufficient, as sequences of operations can eventually transform that carefully balanced tree into the dreaded list
- At that point, everything is O(n) and nobody is happy
 - find
 - insert
 - delete



Balanced BST

Observation

- BST: the shallower the better!
- For a BST with n nodes inserted in arbitrary order
 - Average height is $O(\log n)$ see text for proof
 - Worst case height is O(n)
- Simple cases, such as inserting in key order, lead to the worst-case scenario

Solution: Require a Balance Condition that

- 1. Ensures depth is always $O(\log n)$ strong enough!
- 2. Is efficient to maintain not too strong!

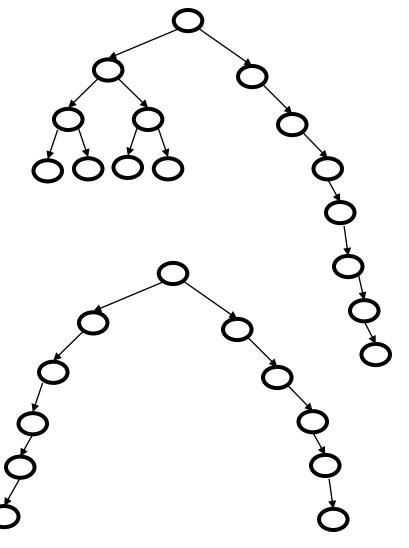
Potential Balance Conditions

 Left and right subtrees of the root have equal number of nodes

Too weak!
Height mismatch example:

 Left and right subtrees of the root have equal height

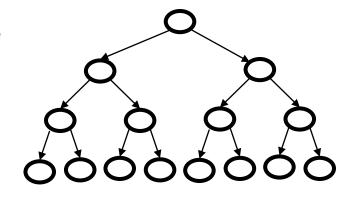
Too weak!
Double chain example:



Potential Balance Conditions

3. Left and right subtrees of every node have equal number of nodes

Too strong!
Only perfect trees (2ⁿ – 1 nodes)



 Left and right subtrees of every node have equal height

Too strong! Only perfect trees (2ⁿ – 1 nodes)

The AVL Balance Condition

Left and right subtrees of *every node* have *heights* **differing by at most 1**

Definition: balance(node) = height(node.left) - height(node.right)

AVL property: for every node x, $-1 \le balance(x) \le 1$

- Ensures small depth
 - Will prove this by showing that an AVL tree of height h must have a number of nodes exponential in h
- Efficient to maintain
 - Using single and double rotations