## CS120: Intro. to Algorithms and their Limitations

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# Lecture 5: Binary Search Trees

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# 1 Announcements

- PS2 posted tomorrow
- PS1 due tomorrow
- SRE next Thursday
- Anurag's Wed OH: Zoom only
- New Office hours have been added: See Ed post and class calendar for latest updates
- Late policy reminder: 8 late days total. 3 can be used on a single pset.

# 2 Binary Search Trees

### 2.1 Motivation

Our consideration of dynamic interval scheduling in the previous lecture led us to seek a dynamic data structure to solve the following problem:

### **Updates**:

- Insert(K,V) for  $K \in \mathbb{R}$ : add one copy of (K,V) to the multiset S of key-value pairs being stored. (S is initially empty.)
- Delete(K) for  $K \in \mathbb{R}$ : delete one key-value pair of the form (K, V) from the multiset S (if there are any remaining).

#### Queries:

- Search(K) for  $K \in \mathbb{R}$ : output  $(K', V') \in S$  such that K' = K.
- Next-smaller(K) for  $K \in \mathbb{R}$ : output  $(K', V') \in S$  such that  $K' = \max\{K'' : \exists V'' \ (K'', V'') \in S, \ K'' < K\}.$

### Data-Structure Problem Dynamic Predecessors

From the last lecture and the lecture notes, we can infer an implementation of this data structure in which the times for insertions, deletions, search queries, and next-smaller queries are: O(n), O(n),  $O(\log n)$ , and  $O(\log n)$ , respectively. This data structure simply maintains a sorted array (we will discuss it briefly in this lecture). In this lecture, we'll look at another implementation, binary trees, in which some of those operations are faster.

## 2.2 Binary Search Tree Definition and Intro

**Definition 2.1.** A binary search tree (BST) is a recursive data structure. Every nonempty BST has a root vertex r, and every vertex v has:

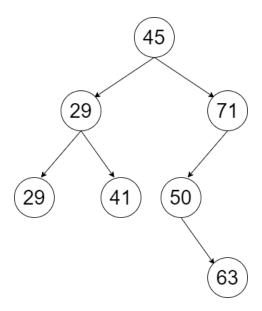
- a key  $K_v$
- a value  $V_v$
- a pointer to a left child  $v_L$ , which is another binary tree (possibly None, the empty BST)
- a pointer to a right child  $v_R$ , which is another binary tree (possibly None, the empty BST)

Crucially, we also require that the keys satisfy the BST Property:

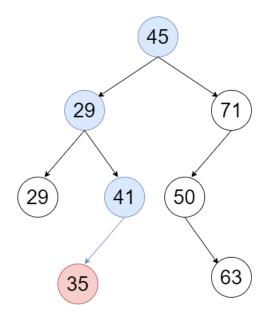
If v has a left-child  $v_L$ , then the keys of  $v_L$  and all its descendants are no larger than  $K_v$ , and similarly, if v has a right-child, then the keys of  $v_R$  and all of its descendants are no smaller than  $K_v$ .

Note that the empty set satisfies all the properties above, and is a BST.

## Example:



If we need to insert 35, then we color the vertices that we compare 35 to:



We remark that the description of binary search trees in Roughgarden and CLRS differs slightly from ours. First, instead of considering the data associated with a key to be a value stored at the same vertex, they treat the vertex itself as the data associated with a key. So, for example, to replace one data element with another, in our formulation, we could just rewrite the key and value attributes of the corresponding vertex, whereas the texts would require removing the vertex from the tree and rewiring pointers to the new vertex.

Relatedly, the texts (and our pset 0) also include a parent pointer at every vertex. It's not necessary for anything we are going to do today, but it can be useful in implementations, as you may have discovered in pset 0 (so feel free to include it if you prefer).

It is also worth noting that in practice, trees with fanout (the number of children, aka maximum outdegree) greater than 2 are often used for greater efficiency in terms of memory accesses. Specifically, in large data systems, the fanout is often chosen so that the size of each vertex is a multiple of the size of a "page" in memory or on disk.

**Theorem 2.2.** Given a binary search tree of height (equivalently, depth) h, all of the dynamic predecessors operations (queries and updates) can be performed in time O(h).

**Definition 2.3** (Height of a tree). The height h of a BST is the length of the longest path from the root vertex to any leaf node. The height of an empty tree is -1.

Proof.

### • Insertions:

```
1 Insert(T,(K,V))
2 if T = \emptyset then
3 | Let T = new BST with key K, value V, no children
4 | return T
5 Let v be the root of T.
6 if K \leq K_v then
7 | T.left = Insert(T_L,(K,V))
8 | return T
9 else
10 | T.right = Insert(T_R,(K,V))
11 | return T
```

**Runtime:** The runtime of this operation is O(h) for the same reason as in pset 0: we walk down the tree at most once, doing O(1) work at each step.

### **Proof of Correctness:**

The definition of a dynamic data structure problem doesn't actually impose any requirements on this operation: this is part of the preprocessing step, and the definition of correctness of a data structure implementation only requires that for all  $x \in \mathcal{I}$ ,  $u_0, u_1, \ldots, u_{m-1} \in \mathcal{U}$  and  $q \in \mathcal{Q}$ , if  $f(x, u_0, u_1, \ldots, u_{m-1}) \neq \emptyset$ , we have:

```
\text{EvalQ}(\text{EvalU}(\cdots \text{EvalU}(\text{EvalU}(\text{Preprocess}(x), u_0), u_1), \dots, u_{m-1}), q) \in f(x, u_0, u_1, \dots, u_{m-1}, q),
```

and  $\perp$  otherwise—these are only conditions on the output of *queries*. Nevertheless, there's something we should prove about the insert operation, which will make proving correctness of queries easier: We should prove that the insert operation respects the BST property. Alternately, we can call this a requirement of the BST implementation, which the insert operation must satisfy to be a valid update.

Our proof of correctness uses induction on the height of the tree. Specifically, we prove the following statement by induction on h:

```
(*) for every BST T of height at most h, Insert(T, (K, V)) returns a tree satisfying the BST property whose key-value pairs are (K, V) \cup \bigcup_{(K_i, V_i) \in T} (K_i, V_i).
```

Our base case is h = -1, i.e. an empty tree. In this case, we return a tree containing exactly (K, V), where no vertex has any children, so the BST property is automatically satisfied.

For the induction step, assume that (\*) is true for trees T of height at most h, and we'll prove it for trees of height at most h+1. Let T be an arbitrary tree of height  $h+1 \geq 0$ , so T has a root v. Given a search key K, we split into casework. If  $K \leq K_v$ , then  $T_L$  is a tree of height at most h, so by induction  $\mathbf{Insert}(T_L, (K, V))$  is a tree containing the key-value pairs in  $T_L$  plus (K, V) such that every vertex in it satisfies the BST property. Also, v itself satisfies the BST property: its right child hasn't changed, so it still satisfies the BST property. Every key  $K_L$  in its left child is either a key in  $T_L$ , which was no larger than  $K_v$  because T satisfied the BST property, or is K itself, which is at most  $K_v$  by the assumption of this case.

The other case, where  $K > K_v$ , is similar.

#### • Search:

```
1 Search(T,K)
2 Let v be the root of T.
3 if K_v = K then
4 | return (K_v, V_v)
5 if K > K_v and T_R \neq \emptyset then
6 | return Search(T_R, K)
7 if K \leq K_v and T_L \neq \emptyset then
8 | return Search(T_L, K)
9 return \bot
```

Our proof of correctness uses induction on the height of the tree. Specifically, we prove the following statement by induction on h:

(\*) for every BST T of height at most h, if T contains a vertex with key K, then Search(T,K) returns a pair  $(K_w,V_w)$  such that  $K_w=K$ , and otherwise Search(T,K) returns  $\perp$ .

Our base case is height 0, i.e. a tree whose only vertex is the root v. In this case, T contains K if and only if  $K = K_v$ , which is exactly the condition under which  $\operatorname{Search}(T,K)$  returns  $(K_v, V_v)$ . Otherwise,  $\operatorname{Search}(T,K)$  returns  $\bot$ , since  $T_R = T_L = \emptyset$ . So  $\operatorname{Search}(T,K)$  is correct.

For the induction step, assume that (\*) is true for trees T of height at most h, and we'll prove it for trees of height at most h+1. Given an arbitrary tree T of height h+1 and a search key K, we split into casework. Letting v be the root of T, if  $K_v = K$  we are successful because we return  $(K_v, V_v)$ . If  $K_v < K$  then by the BST property K is in T if and only if K is in the right subtree  $T_R$  (which is of height at most h), so the recursive call Search  $(T_R, K)$  succeeds by induction, and the left call is similar.

For the runtime, note that we do a constant amount of work at each level and recurse at most h times on a tree of height h. To formally analyze this, let T(h) be the maximum amount of work done by a search call on a tree of height at most h. Then for h > 0, we have

$$T(h) = T(h-1) + c,$$

for a constant c. Unrolling, we get

$$T(h) = T(h-1) + c = (T(h-2) + c) + c = \cdots = T(0) + h \cdot c = O(h)$$

#### • Minimum (and Maximum):

For Minimum, we always go left until there is no left child.

#### • Next-smaller:

Given a query Next-smaller(q) on a tree T rooted at vertex v with key  $K_v$ , we again use a search type strategy. If  $K_v \geq q$ , return Next-smaller(q) on the left subtree  $T_L$  (if it exists if not, return  $\bot$ ). If  $K_v < q$ , call Next-smaller(q) on the right subtree  $T_R$  (if it exists). If we receive  $\bot$ , return  $(K_v, V_v)$  and otherwise return the result of the call on  $T_R$ .

• **Deletions:** next SRE

# 3 Balanced Binary Search Trees

So far we have seen that a variety of operations (search, insertion, deletion, min, max, next-smaller, and next-larger) can be performed on Binary Search trees in time O(h), where h is the height of the tree. Thus, we are now motivated to figure out how we can ensure that our binary search trees remain shallow (e.g. of height  $O(\log n)$  where n is the number of items currently in the dataset). While doing so, we need to be sure that we retain the binary-search-tree property (the ordering of keys between a vertex and its left-descendants and its right-descendants).

There are several different approaches for retaining balance in binary search trees. We'll focus on one, called AVL trees, after mathematicians named Adelson-Velsky and Landis.

**Definition 3.1** (AVL Trees). An AVL Tree is a binary search tree in which:

- Every vertex has an additional attribute containing its *height* in the tree (length of the longest path from the vertex to a leaf). ("data structure augmentation")
- Every pair of siblings in the tree have heights differing by at most 1 (where we define the height of the an empty subtree to be -1). ("height-balanced")

**Lemma 3.2.** Every AVL Tree with n vertices has height (=depth) at most  $2\log_2 n$ .

Proof.

Let  $n(h) = \min$  number of vertices in an AVL tree of height  $\geq h$ . Then, by the second AVL property, for  $h \geq 2$ ,  $n(h) \geq n(h-1) + n(h-2) + 1$ . If we drop some constants, then

$$n(h) \ge n(h-1) + n(h-2) + 1 \ge 2n(h-2).$$

Now we are at a good place for unenrolling the expression:

$$n(h) \ge 2n(h-2)$$

$$\ge 4n(h-4)$$

$$\ge 8n(h-6)$$

$$\vdots$$

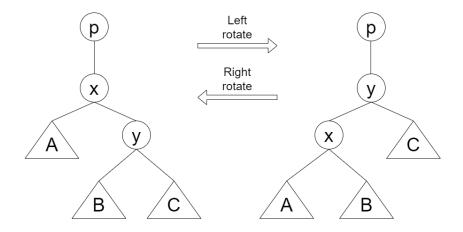
$$\ge n^{n/2}n(0)$$

$$= 2^{h/2}.$$

(The above is for the case when h is even; a similar analysis can be done when h is odd.)

So we can apply all of the aforementioned operations on an AVL Tree in time  $O(\log n)$ . But an insertion or deletion can ruin the AVL property. To fix this, we need additional operations that allow us to move vertices around while preserving the binary search tree property. One important example is a *rotation*.

We first investigate rotations on BSTs.



On the left: all keys in A are  $\leq K_x$ , all keys in B are  $\in [K_x, K_y]$ , and all keys in C are  $\geq K_y \implies$  BST property is preserved in the right.

From left to right we have the operation Left.rotate(x). From right to left we have the operation Right.rotate(y).

How much time does a rotation take?

Rotations only take O(1) time! One only needs to change a constant number of pointers – there is no traversing or other expensive operations.

**Theorem 3.3.** We can insert a new key-value pair into an AVL Tree while preserving the AVL Tree property in time  $O(\log n)$ .

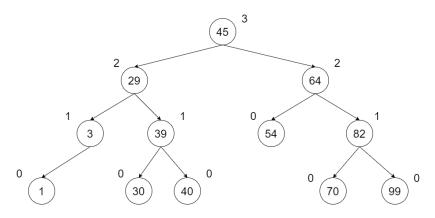
Proof. (sketch)

The algorithm runtime is as follows:

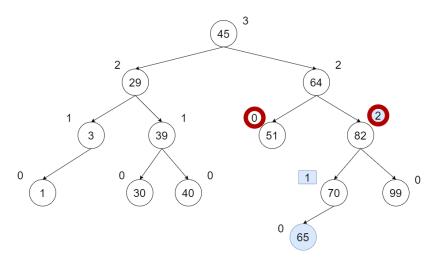
- Do an ordinary BST insert  $\rightarrow$  adding a leaf L.
- Update heights working all the way from L to the root.
- When we see that we violate the AVL tree property we use rotations to fix it.

In total, this requires time  $O(\text{height}) = O(\log n)$ .

Let us see an example. The following is our original BST tree:



Now we want to insert L=65. We update the height of the corresponding vertices (rectangles in blue), but when we reach the vertex 82 we see that we are now violating the AVL property: vertex 51 has height 0 while vertex 82 has height 2, so their difference is more than 1 (marked in red).



To fix this, we then use AVL rotations. Let y be the vertex of key 64, and z be the vertex of key 82, and to the above tree we apply the operation Right.rotate(z) followed by Left.rotate(y).