CSCI 5454 Final Project: AVL Tree

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Introduction

What is an AVL Tree?

An AVL tree is a binary search tree that is "self-balancing". This means after each operation, like an insertion or deletion, on the tree the heights of each node's children differ by at most 1. The height of a node is the number of nodes in the longest path from it to a leaf node. A leaf node always has a height of 0. Its parent node, if the leaf node was the parent's only child, would have a height of 1. An AVL tree is "self-balancing" because after each tree operation the heights and balance of the nodes are readjusted by the tree [1].

What problems does it solve?

AVL trees, like binary trees, are used for storing and retrieving information. Their advantage is that they can perform these operations faster than if the information was stored in an array. As will be shown later in this paper, storing and retrieving takes $\log n$ time for an AVL tree while performing the same operations on an array could take up to n time.

Where is it used?

Red-black trees, another kind of self-balancing binary search tree, are typically used instead of AVL trees in real world applications [2]. Red-black trees can be found in the C++ Standard Library (std) as the underlying data structure for the std::map and std::set containers and it is reasonable to think that AVL trees could be used instead.

Mathemetical Analysis of Correctness, Runtime, and Space

Correctness

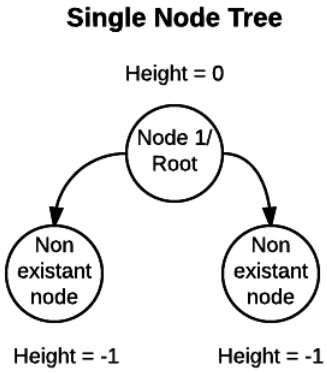
The correctness of the AVL tree relies on maintaining two invariants

1. The height of a node is the number of nodes in the longest path from it to a leaf node.

2. The heights of each node's children differ by at most 1. This is demonstrated by the equation $|x - y| \leq 1$ where x is the height of the left child node and y is the height of the right child node.

Basecase

Figure 1: An AVL tree with a single node



We will use the insertion operation to show these invariants are maintained. We first start with a single, leaf node in a tree as the base case as shown in Figure 1. First, we let the height of any non existent node be equal to -1. This means a single node in a tree has two non existent children nodes each with a height of -1. To find the height of the single node in the tree we take the max of the heights of the children and then add 1 to it to always maintain the first invariant. In this case max(-1, -1) = -1 and adding 1 to this result gives a height of 0 for the single node in the tree. To make sure the second invariant is maintained we use the given equation

$$|x - y| \leq 1$$

In the case of a single, leaf node the height of the left child is -1 and the height of the right child is -1. Substituting these values in for x and y produces the

equation

$$\begin{aligned} |(-1) - (-1)| &= 0 \\ &\leq 1 \end{aligned}$$

Since $0 \leq 1$ the second invariant is maintained.

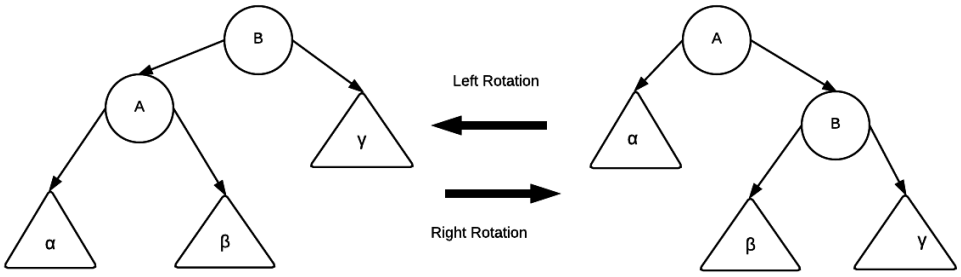
Inductive Proof of the First Invariant

Next it must be inductively shown that the two invariants hold for every AVL tree with size greater than 1 node. Concerning the first invariant, every time a new node is inserted into the AVL tree, it will be a leaf node with a height of 0 as was demonstrated in the base case. Once an insertion takes place, the AVL tree updates the the balances and heights of the new node and all of the nodes above it. This is done by recursing through the parent nodes, starting with the inserted node and moving upwards to the root. The height of each parent node is equal to the max of the childrens' heights plus 1. Since this update of the heights is bottom up and takes place for every node inserted into the tree, the first invariant holds for any tree of size greater than 1 nodes.

Inductive Proof of the Second Invariant

To show the correctness of the second invariant we have to look at the possible cases of a tree being out of balance and how those are corrected. In

Figure 2: Left and Right Tree Rotations

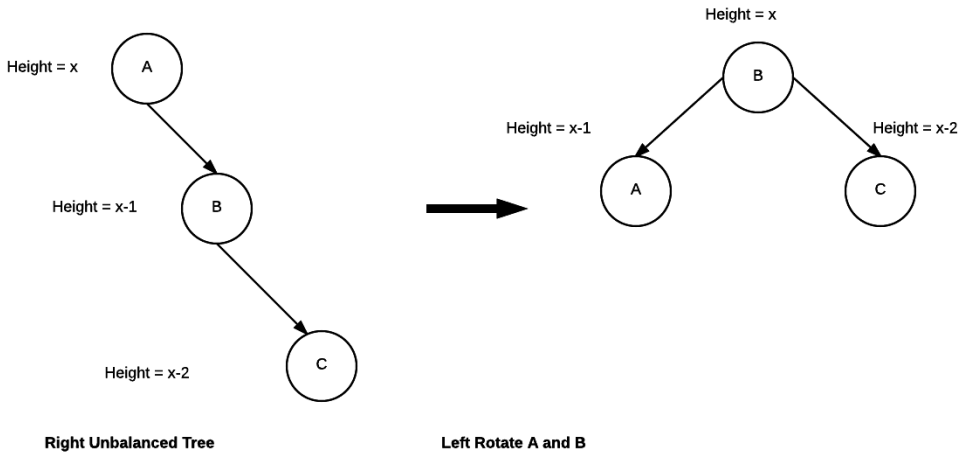


order for an AVL tree to remain in balance, the tree has to perform what are

known as tree rotations after nodes are inserted. [1]. The basic two rotations can be seen in Figure 2. During tree rotations, the shape of a tree is modified in order to change the heights of subtrees at a particular node. For example, doing a right rotation of the tree in Figure 2 "increases" the height of node A while "decreasing" the height of node B [3]. These can be judiciously used to rebalance the tree and maintain the second invariant of an AVL tree.

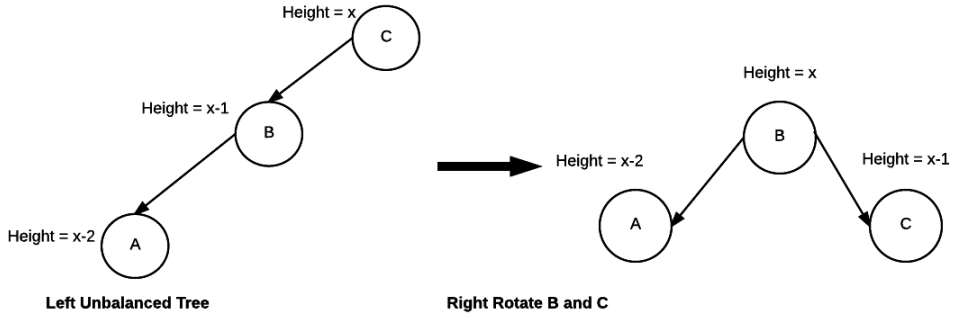
When a node is inserted, the first thing that the tree does is check the balance at the inserted node and all of the nodes above to ensure that the equation for the second invariant is maintained. If the balances of any of the nodes do not satsify the second invariant equation, tree rotations must take place in order to balance the tree. Four cases of an unbalanced tree may occur after a node is inserted into a tree, as shown in Figures 3-6. These may require one or more rotations to rebalance a tree [4]. Once one or more rotations

Figure 3: Right Unbalanced Tree



of the subtrees of a node take place and that node is balanced according to the second invariant equation, the height of that node and any of the other nodes invovled in the rotation are updated. The tree then recursively checks the node's parents, grandparents, etc. until the root of the tree is reached for correct balances, performing any rotations if necessary, and updating the heights as it moves upwards through the tree. This method of updating the

Figure 4: Left Unbalanced Tree



balances and heights ensures that given any number of inserted nodes, the tree with always be rebalanced and each node will have the correct height.

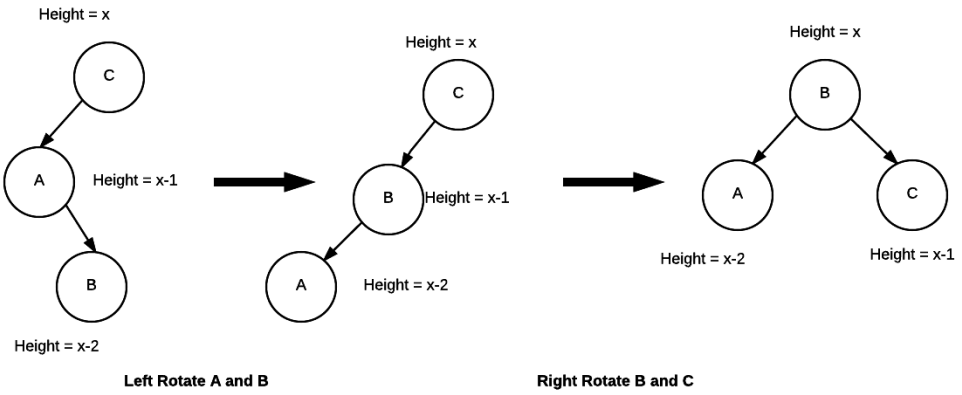
Runtime

Since an AVL tree is a binary search tree any AVL tree operations' runtime depends on the trees height, which is the longest path from the root node to a leaf – also known as its longest branch [5]. In order to prove the runtime any AVL tree operations is O(log n) we need to show that the height of an AVL tree can be bounded by O(log n).

Let the root of the AVL tree have a height of h. One child of the root will have a height of $h-1$ and the other, if we follow the property of an AVL tree that says the heights of the children differ by at most 1, will have a minimum height of $h-2$. Let n_h be the number of nodes in the tree including the root. Let n_{h-1} be the number of nodes underneath and including the child of the root with height $h-1$ and n_{h-2} be the number of nodes underneath and including the child with height $h-2$. We can construct the total size of the tree n_h by using the recurrence relation

$$n_h = n_{h-1} + n_{h-2} + 1$$

Figure 5: Right Left Unbalanced Tree



We can bound the height of the AVL tree by unrolling the recurrence relation with the following setps:

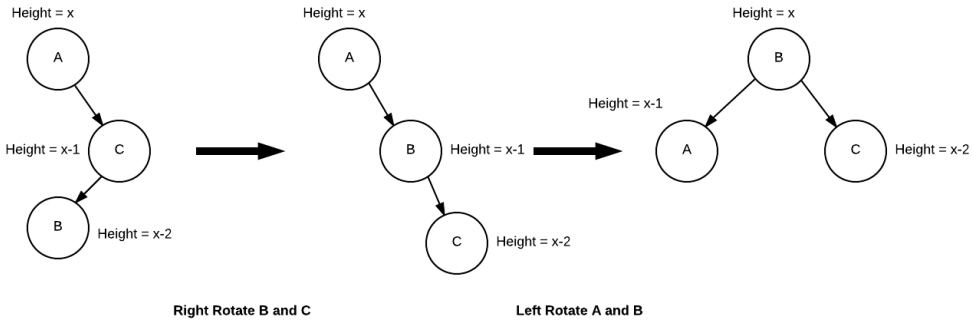
$$n_h = n_{h-1} + n_{h-2} + 1 \tag{1}$$
$$n_{h-1} = n_{h-2} + n_{h-3} + 1 \tag{2}$$
$$n_h = (n_{h-2} + n_{h-3} + 1) + n_{h-2} + 1 \tag{3}$$
$$n_h > 2n_{h-2} > 2(2n_{h-4}) > 2(2(2n_{h-6})) > \ldots > 2^{\frac{h}{2}} \tag{4}$$
$$\log\, n_h > \log\, 2^{\frac{h}{2}} \tag{5}$$
$$\log\, n_h > \frac{h}{2} \tag{6}$$
$$2\log\, n_h > h \tag{7}$$

This shows that the height h of an AVL tree will always be bounded by O(log n). This means the runtime for any AVL tree operation will be bounded by O(log n).

Space

The space an AVL tree uses is equal the number of the nodes in the tree. If there are $n + 1$ nodes inserted into the tree, the space used will be $n + 1$. If there are $n - 1$ nodes removed from the tree, the space used will be $n - 1$. The space is equal to $\theta(n)$.

Figure 6: Left Right Unbalanced Tree



Numerical Characterization of Runtime and Space

Description of the code invloved in the Numerical Characterization

To show the numerical characterization of the space and time performance of an AVL tree, I created a randomized input generator to show the runtime and space of the three operations that can be performed on the AVL tree: insertion, deletion, and search. I ran 12 iterations for each of these tree operations, varying the number of nodes n in the tree by a factor of 2. I ran each of these iterations 50 times each to find the average number of operations given a value of n. n takes on all of the values in the set $\{2^4, 2^5, \ldots, 2^{16}\}$ for each tree operation. The keys for each of the nodes in the n node-sized tree were randomly chosen from the set $\{0, 1, \ldots, n\}$ and then inserted into the tree to create a tree of size n.

Once the tree was generated for an iteration, I then chose a random key from the set of keys that were already in the tree, and I either searched for it, inserted it, or deleted it. To keep track of the runtime when performing these operations on the tree, I kept a global variable as a counter and incremented it every time an atomic operation occurred. To keep track of the space used when performing these operations on the tree, I kept a global variable as a counter and incremented it when a node was inserted and decre-

mented it when a node was removed from the tree. These global variables were reset after each iteration.

Once all the iterations were complete for a specific operation, I took the n for each iteration and the values of the global variables for each iteration and graphed them as a function $f(x)$. The values of n were placed on the x-axis and the values of the globabl variables were placed on the y-axis. The graphs use the logarithmic scale instead of the linear scale because the logarithmic scale more clearly shows what n and the values of the global variables are doing. I then graphed one other line representing the function $g(x)$, which is the function that bounds $f(x)$. $g(x)$ is then multiplied by a constant, c_2, producing a separate line.

Characterization of Runtime

Runtime of Search

Figure 7: Graph of the runtime for searching for a key in different-sized AVL trees

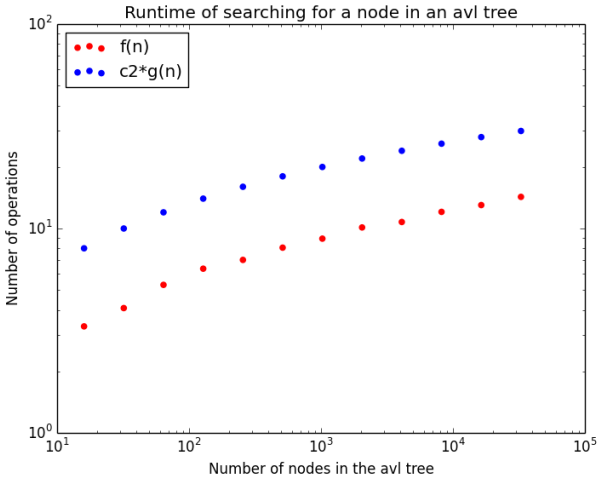


Figure 7 shows the runtime for searching for a key in different-sized avl trees. The function that bounds the runtime of search is $g(n) = \log n$ multiplied by the constant $c_2 = 2$. As can be seen in the graph, the runtime of search is O(log n), with the function log n bounding the search runtime from above.

This means given an AVL tree of size n, it will take O(log n) operations to search for key in the tree.

Runtime of Insert

Figure 8: Graph of the runtime for inserting a key into different-sized AVL trees

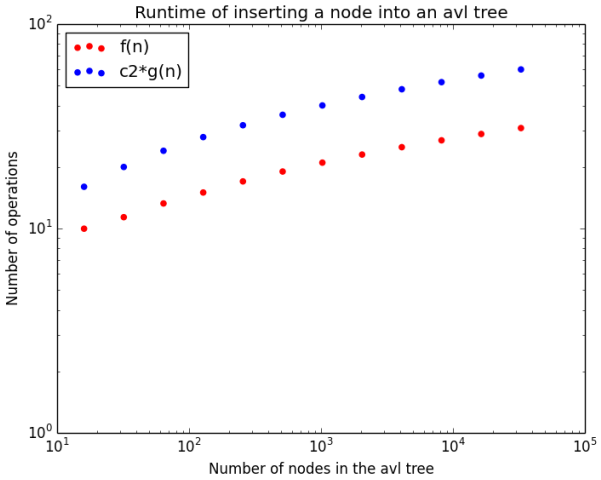
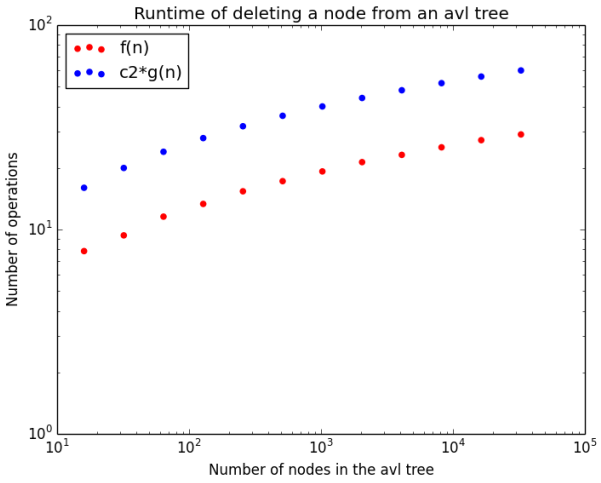


Figure 8 shows the runtime for inserting a key into different-sized AVL trees. The function that bounds the runtime of insert is $g(n) = \log n$ multiplied by the constant $c_2 = 4$. As can be seen in the graph the runtime of insert is O(log n), with the function log n bounding the insert runtime from above. This means given an AVL tree of size n, it will take O(log n) operations to insert a key into the tree.

Runtime of Delete

Figure 9 shows the runtime of deleting a key from different-sized AVL trees. The function that bounds the runtime of delete is $g(n) = \log n$ multiplied by the constant $c_2 = 4$. As can be seen in the graph the runtime of delete is O(log n), with the function log n bounding the delete runtime from above. This means given an AVL tree of size n, it will take O(log n) operations to delete a key from the tree.

Figure 9: Graph of the runtime for deleting a key from different-sized AVL trees



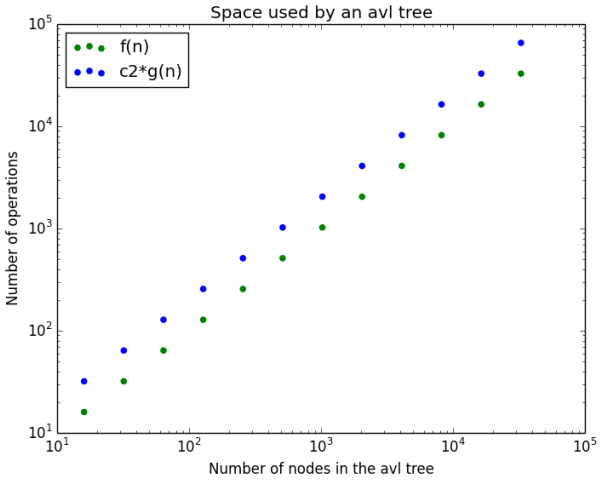
Characterization of Space Usage

Figure 10 shows the space used by different-size AVL trees. The function that bounds the space used by an AVl tree is $g(n) = n$ multiplied by the constant $c_2 = 2$. As can be seen in the graph the space that is used by an AVL tree is O(n), with the function n bounding the space used from above. This means that given an AVL tree of size n, there will be exaclty n nodes in the tree. With a tree of size n, if a node is inserted there will be $n + 1$ nodes in the tree. With a tree of size n, if a node is deleted there will be $n - 1$ nodes in the tree.

Extensions, Improvements, and Recent Work

There has been work done recently to improve the time it takes to rebalance an AVL tree. A relaxation of AVL tree balance constraints produces a new data structure called a weak AVL tree (wavl). It has been shown that insert and delete take at most "two rotations" in the worst case and "O(1)" amortized time. This is fewer rotations than the number of rotations a regular AVL tree and red-black tree need to perform to rebalance [6].

Figure 10: Graph of the space used by different-size AVL trees



A relaxed version of an AVL tree was also proposed to be used in a "concurrent tree algorithm" to improve thread throughput. In this case, an AVL tree acts like a map object with operations like "get, put, remove." An AVL tree is relaxed because balancing of the tree is delayed until after the map operations take place on the tree [7].

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