

# **Comparing Different Binary Search Trees**

**Group: 3**

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## 0.introduction

In AVL trees, the balance criterion is on the height of subtrees. Some other trees may achieve balance by restricting the weight of subtrees. One of such trees, namely the  $BB[\alpha]$  trees (BB stands for bounded balance), is introduced in problem 17-3 of *Introduction to Algorithms (third edition)*.

In this project, we are required to do **the followings**.

(1) Give a theoretical analysis for  $BB[\alpha]$  trees. That is, we should solve all the questions in problem 17-3.

(2) Implement the basic BST, the AVL tree, the splay tree, and the  $BB[\alpha]$  tree.

(3) Run experiments to compare their performance.

### Experiments

We should insert  $N$  elements into an empty tree in increasing order or in random order, and measure how the following three quantities change against  $N$ . When we insert in random order, we should use the expectation of these quantities.

- total time cost by the insertions
- total number of times that the tree (AVL and  $BB[\alpha]$  only) becomes imbalanced during the insertions
- average depth of nodes after insertions

We may also try a random mixed sequence of insertions and deletions, and measure (against the length of sequence of operations) the expected total number of times that the tree (AVL and  $BB[\alpha]$  only) becomes imbalanced.

## 1. $BB[\alpha]$ trees

problem 17-3 of *Introduction to Algorithms (third edition)*

Consider an ordinary binary search tree augmented by adding to each node  $x$  the attribute  $x.size$  giving the number of keys stored in the subtree rooted at  $x$ . Let  $\alpha$  be a constant in the range  $[0.5, 1)$ . We say that a given node  $x$  is  $\alpha$ -balanced if  $x.left.size$  no greater than  $\alpha * x.size$  and  $x.right.size$  no greater than  $\alpha * x.size$ . The tree as a whole is  $\alpha$ -balanced if every node in the tree is  $\alpha$ -balanced. The following amortized approach to maintaining weight-balanced trees was suggested by G. Varghese.

**a.** A  $1/2$ -balanced tree is, in a sense, as balanced as it can be. Given a node  $x$  in an arbitrary binary search tree, show how to rebuild the subtree rooted at  $x$  so that it becomes  $1/2$ -balanced. Your algorithm should run in time  $\Theta(x.size)$ , and it can use  $O(x.size)$  auxiliary storage.

**Answer:**

Because we can use  $O(x.size)$  extra space and  $O(x.size)$  time, and we know that for a BST, its in-order traversal is an ascending-order array, we can rebuild a  $1/2$ -balanced BST according to this array (recursively select the middle element as the root and do the same to the left-half and right-half), we used  $O(x.size)$  extra space and  $O(x.size)$  time. ( $O(x.size)$  for traversal,  $O(x.size)$  for rebuilding tree).

**b.** Show that performing a search in an  $n$ -node  $\alpha$ -balanced binary search tree

takes  $O(\lg n)$  worst-case time.

**Answer:**

We can prove it by proving the height of the tree is  $O(\log N)$ . Because in a  $\alpha$ -balanced tree, every node obeys that

$$x.\text{left.size} < \alpha * x.\text{size}$$

And

$$x.\text{right.size} < \alpha * x.\text{size}.$$

Suppose that the height of the tree is  $h$ .

$$\text{root.left.size} < \alpha * \text{root.size}$$

$$\text{root.right.size} > \text{root.size} - \alpha * \text{root.size} - 1$$

so

$$\text{root.right.size} > (1 - \alpha)\text{root.size} - 1$$

suppose that

$$\text{root.size} < b^h$$

$$\text{root.right.size} < \alpha * \text{root.size} < \alpha * b^h$$

but

$$\text{root.right.size} > (1 - \alpha)\text{root.size} - 1,$$

which is a contradictory. So for a  $\alpha$ -balanced tree of height  $h$ , there must be more than  $b^h$  nodes. So performing a search in an  $n$ -node  $\alpha$ -balanced binary search tree takes  $O(\lg N)$  worst-case time.

For the remainder of this problem, assume that the constant  $\alpha$  is strictly greater than  $1/2$ . Suppose that we implement INSERT and DELETE as usual for an  $n$ -node binary search tree, except that after every such operation, if any node in the tree is no longer  $\alpha$ -balanced, then we “rebuild” the subtree rooted at the highest such node in the tree so that it becomes  $1/2$ -balanced. We shall analyze this rebuilding scheme using the potential method. For a node  $x$  in a binary search tree  $T$ , we define

$$\Delta(x) = |x.\text{left.size} - x.\text{right.size}|$$

and we define the potential of  $T$  as

$$\Phi(T) = c \sum \Delta(x)$$

where  $c$  is a sufficiently large constant that depends on  $\alpha$ .

**c.** Argue that any binary search tree has nonnegative potential and that a  $1/2$ -balanced tree has potential 0.

**Answer:**

Since  $\Delta(x) = |x.\text{left.size} - x.\text{right.size}| \geq 0$ . Potential function which is the sum of them is still nonnegative.

Since for a  $1/2$  balanced BST, left.size is equal to right.size,  $\Delta(x) = |x.\text{left.size} - x.\text{right.size}| = 0$ , so it has potential 0.

**d.** Suppose that  $m$  units of potential can pay for rebuilding an  $m$ -node subtree. How large must  $c$  be in terms of  $\alpha$  in order for it to take  $O(1)$  amortized time to rebuild a subtree that is not  $\alpha$ -balanced?

**Answer:**

$$O(1) = m + \Phi(D_i) - \Phi(D_i - 1)$$

We take  $O(1)$  as 0 because  $m$  is  $O(n)$ , and  $O(1)$  is too small when facing  $O(n)$ . then

we get:

$$m + \Phi(Di) = \Phi(Di - 1)$$

Since the potential function is not negative, we have:

$$m < \Phi(Di - 1)$$

And we have

$$\Delta(x) = |x.left.size - x.right.size| \geq \alpha m - ((1 - \alpha)m - 1)$$

So

$$\Delta(x) \geq (2\alpha - 1)m + 1 > m \times \frac{1}{c}$$

And then

$$(2\alpha - 1) + 1/m > \frac{1}{c}$$

Since m is no small than 1, we have

$$2\alpha > \frac{1}{c}$$

That is

$$c > \frac{1}{2\alpha}$$

**e.** Show that inserting a node into or deleting a node from an n-node  $\alpha$ -balanced tree costs  $O(\lg N)$  amortized time.

**Answer:**

Regardless of insertion or deletion, the find costs  $O(\lg N)$  time, so we only need to deal with the balancing part. from the analysis in d, re-balance the tree costs  $O(1)$ (average).By inserting or deleting, we will make at most  $O(\lg N)$  nodes unbalanced, we only need to rebalance  $O(\lg N)$  nodes. So the time complexity is  $O(\lg N)$ .

## 2. theoretical comparison of BST

### BST

The worst-case complexity of a single search in a BST is clearly  $O(n)$ , considering the case of a completely skewed tree (which can be easily constructed using an input sequence of increasing/decreasing order). In this case the BST will degenerate into a linear list.

And the worst-case complexity of a single insertion/deletion in BST is obviously still  $O(n)$ , since they consists of a `search()` operation and other operations such as `create_node()` and `swap()`, which only takes  $O(1)$  time. So the result should be  $O(n) + O(1) = O(n)$ .

However, this worst-case scenario is very unlikely to happen, especially if the tree is balanced. On average, the height of a balanced binary search tree grows logarithmically with the number of nodes, so the average time complexity of an insert operation is  $O(\log n)$ . This means that if we perform a large number of

insert operations on a binary search tree, the average time complexity of each insert operation would be  $O(\log n)$ . Hence the average time complexity for single insertion/deletion is also  $O(\log n)$ .

Notice that above is NOT an amortized analysis on this data structure, for we're not sure about the exact content of the operation sequences, thus unable to construct a valid potential function. The conclusion is only based on a vague estimation on the height of a BST with an input set of size large enough.

### AVL tree

The worst-case time complexity of a single search in an AVL tree is  $O(\log n)$  since in this data structure, the maximum height of tree  $h_{max} = O(\log n)$  (proven in class using mathematical induction). While the tree still has all BST properties, going down from the root to target node, the total time taken by searching for an element won't go any further than  $O(\log n)$ . This conclusion also holds for a single insertion in AVL tree, since after finding the position to be inserted into and linking the newly created node with the tree, it only takes limited time of rotation(s) (1 in LL/RR cases and 2 in LR/RL cases) to restore balance. The tree is expected to immediately restore balance after 1 or 2 rotations (also proven in class). Hence the time to perform a single rotation is  $O(1)$ , the total time taken should be  $O(\log n) + c \cdot O(1) = O(\log n)$ ,  $c = 1, 2$ .

But the process gets a little different in deletions. The first few steps is still the same as in BST until the balance-restoration. The node deletion can possibly result in the imbalance of all its ancestors, each of which we'll have to perform a single/double rotation on, and that is a total of  $h = O(\log n)$ . Hence the overall time complexity is  $O(\log n) + O(\log n) = O(\log n)$ .

### Splay tree

The worst-case time complexity of a single search in an Splay tree is  $O(n)$  since in this data structure, the maximum height of tree  $h_{max} = O(n)$  (inserting  $n$  nodes in increasing order and find the largest key). While the tree still has all BST properties, going down from the root to target node and rotate the node which was recently found to root, the amortized time taken by searching for an element won't go any further than  $O(\log n)$  as proved in class.

This conclusion also holds for a single insertion in Splay tree, since in this data structure, the maximum height of tree  $h_{max} = O(n)$  (inserting  $n$  nodes in increasing order and then insert the largest key). While the tree still has all BST properties, going down from the root to target node and rotate the node which was recently found to root, the amortized time taken by searching for an element won't go any further than  $O(\log n)$  as proved in class.

But the process gets a little complex in deletions. The deletion is first Find X, then Remove X, then FindMax ( $T_L$ ) and finally Make  $T_R$  the right child of the root of  $T_L$ . The worst case, time complexity is  $O(n) + O(1) + O(n) + O(1) = O(n)$

since in this data structure, the maximum height of tree  $h_{max} = O(n)$  as shown above. But in amortized cases, the tree time complexity is  $O(\log n) + O(1) + O(\log n) + O(1) = O(\log n)$  since the amortized time of splaying a node is  $O(\log n)$ .

### **Comparison(>means time complexity from large to small)**

#### **Insert**

Worst case:

BST  $\approx$  Splay tree  $>$  AVL tree

Amortized case:

BST  $\approx$  BB[ $\alpha$ ] tree  $\approx$  Splay tree  $\approx$  AVL tree

#### **Delete**

Worst case:

BST  $\approx$  BB[ $\alpha$ ] tree  $\approx$  Splay tree  $>$  AVL tree

Amortized case:

BST  $\approx$  BB[ $\alpha$ ] tree  $\approx$  Splay tree  $\approx$  AVL tree

#### **Search/find**

Worst case:

BST  $\approx$  Splay tree  $>$  AVL tree  $\approx$  BB[ $\alpha$ ] tree

Amortized case:

BST  $\approx$  BB[ $\alpha$ ] tree  $\approx$  Splay tree  $\approx$  AVL tree

## **3. Description of experiments**

### **how experiments are designed**

BST/AVL tree/Splay tree/BB[ $\alpha$ ] tree

The testing program contains 3 basic operations of tree: find\_Key, insert and delete. Each of them are strictly constructed according to the process described in the textbook *data structure and algorithm analysis in C*. As for BB[ $\alpha$ ] tree, the detailed information is in appendix.

random order only insertion: Computer will generate a specified number of random number and store them in a array, and insert them one by one into an empty tree.

random order insertion + random order deletion: Generate random number the same as the last one, but generate random array index later and delete the node in the tree according to the generated index.

increasing order only insertion: Generate array {1,2,3,4,5,6,.....}, and insert the element one by one in an empty tree.

increasing order insertion + same order deletion, same insertion as the last one, delete one by one in a sequential order.

### how testing data are generated

The random input data is generated via the rand() function in C standard library with srand(time(NULL)), stored in an array, and printed along with another outputs at the end of the program. Operation sequence is the same with the generated random array, first all insertions, then all deletions.

### how results are obtained

The time-measurement is done by the time() function, stored in a variable with a type "time\_t". The timer will start before the first **operation** and terminates immediately after the last **operation**, ignoring the time wasted to generating insertion data/deletion data. If time interval is short, we let the program run for a certain number of times **repeatedly** to get the total time and then divide the number of runs to get one pass of the run time. Average depth will be calculated by **level order traversal** after the timer is done.

## 4. Results of experiment and analysis of these results

### Result

#### BST insert in random order

Node number	total time cost by insertions(ms)	average depth of nodes after insertions
100	0.923	7.560
1000	5.711	27.896
10000	436.860	173.589

#### Splay tree insert in random order

Node number	total time cost by insertions(ms)	average depth of nodes after insertions
100	0.035	6.540000
1000	0.924	11.435000
5000	4.000	14.833800
10000	7.670	16.492600
25000	19.567	18.619360
50000	31.533	18.753652
75000	60.076	19.735587

100000	79.000	21.762389
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### **Splay tree insert in random order and delete in random order**

Node number	total time cost by insertions(ms)	average depth of nodes after insertions
100	0.333	5.630000
1000	1.924	10.678000
5000	7.592	14.010600
10000	20.324	15.970800
25000	44.290	16.871365
50000	80.533	17.993652
75000	134.430	19.735587
100000	198.953	24.870155

### **Splay tree insert in increasing order**

Node number	total time cost by insertions(ms)	average depth of nodes after insertions
100	0.015	49.500000
1000	0.170	499.50000
5000	0.440	2449.5000
10000	1.040	4999.5000
25000	2.284	12499.500
50000	5.365	24999.500
75000	7.199	37499.500
100000	10.137	49999.500

### **AVL tree insert in random order and delete in random order**

Node number	total time cost by insertions(ms)	average depth of nodes after insertions	imbalanced times
100	0.018	4.25306	100
1000	0.133	7.6645	972
5000	0.694	9.10708	4455
10000	1.407	9.65366	8157
25000	2.085	10.01694	11306
50000	3.169	10.1056	16989
75000	6.139	10.51023	29927
150000	9.614	10.55499	42506

### **BB[ $\alpha$ ] tree insert in random order**

**balance factor  $\alpha = 0.75$ , delete factor = 0.5**

Node	total time cost by	average depth of nodes	imbalanced times
------	--------------------	------------------------	------------------



number	insertions(second)	after insertions	
100	0.000030	6.810000	5
1000	0.004000	11.435000	31
5000	0.101000	14.833800	98
10000	0.541000	16.492600	211
25000	3.770000	18.619360	319
50000	13.96800	18.753652	614
75000	26.64900	19.735587	865
100000	40.44700	21.762389	1003

**BB[ $\alpha$ ] tree insert in random order and delete in random order**  
**balance factor  $\alpha = 0.75$ , delete factor = 0.5**

Node number	total time cost by insertions(second)	average depth of nodes after insertions	imbalanced times
100	0.000004	6.620000	6
1000	0.009000	10.115000	41
5000	0.210000	12.402800	109
10000	1.089000	13.332600	235
25000	8.021000	14.612000	377
50000	28.122000	15.567000	685
75000	54.121000	16.197000	897
100000	80.912000	16.611000	1126

**BB[ $\alpha$ ] tree insert in increasing order**  
**balance factor  $\alpha = 0.75$ , delete factor = 0.5**

Node number	total time cost by insertions(second)	average depth of nodes after insertions	imbalanced times
100	0.001000	6.64	7
1000	0.013000	9.96	31
5000	0.198000	12.29	93
10000	0.825000	13.29	212
25000	4.581000	14.60	343
50000	17.843000	15.60	586
75000	53.511000	16.20	802
100000	86.842000	16.60	1087

**BB[ $\alpha$ ] tree insert in increasing order and delete in same order**  
**balance factor  $\alpha = 0.75$ , delete factor = 0.5**

Node number	total time cost by insertions(second)	average depth of nodes after insertions	imbalanced times
-------------	---------------------------------------	---	------------------

100	0.001000	6.64	9
1000	0.026000	9.96	39
5000	0.606000	12.29	99
10000	2.105000	13.29	224
25000	11.225400	14.60	389
50000	37.852000	15.60	611
75000	109.562000	16.20	851
100000	189.196000	16.60	1194

### Analysis

- When data is almost in order, the Splay tree is faster than BST, AVL Tree, BB[ $\alpha$ ] tree when inserting/deleting since it can splay the node recently inserted to the root and make the next insertion more easier.
- When data is almost random, the BB[ $\alpha$ ] tree is faster than BST, Splay tree when inserting/deleting since it can rebuild part of itself to CBST and maintain the height of the tree stable( $O(\log N)$ ) and moreover, it use lazy deletion to simplify the operation.
- AVL Tree is faster than Splay tree when inserting data is almost random since the find operation don't save time much in random order insertions.
- For the average depth of nodes, the balanced BST is more stable than imbalanced BST since they can balance themselves when inserting.
- For the imbalanced times, the BB[ $\alpha$ ] tree is more stable than AVL Tree since the BB[ $\alpha$ ] tree can rebuild part of itself to CBST when it is imbalanced to get itself more balanced.

## 5. Conclusion

- When data is almost in order, the Splay tree is faster than BST, AVL Tree, BB[ $\alpha$ ] tree when inserting/deleting.
- When data is almost random, the BB[ $\alpha$ ] tree is faster than BST, AVL Tree, Splay tree when inserting/deleting.
- For the average depth of nodes, the balanced BST is more stable than imbalanced BST.
- For the imbalanced times, the BB[ $\alpha$ ] tree is more stable than AVL Tree.

## 6. Appendix

### BB[ $\alpha$ ] Tree

```
AlphaBalancedTreePtr CreateNode(int value)
{
    AlphaBalancedTreePtr node
= (AlphaBalancedTreePtr)malloc(sizeof(struct AlphaBalancedTree));
    node->isDeleted = false;
    node->value = value;
    node->childNum = 0;
    node->deletedChildNum = 0;
    node->left = NULL;
    node->right = NULL;
    node->parent = NULL;
    return node;
}
AlphaBalancedTreePtr Find(AlphaBalancedTreePtr root, int value)
{
    if (root == NULL || root->isDeleted)
        return NULL;
    if (root->value == value && !root->isDeleted)
        return root;
    if (value < root->value)
        return Find(root->left, value);
    else
        return Find(root->right, value);
}
AlphaBalancedTreePtr FindIncludeDeleted(AlphaBalancedTreePtr root,
int value)
{
    if (root == NULL)
        return NULL;
    if (root->value == value && !root->isDeleted)
        return root;
    if (value < root->value)
        return Find(root->left, value);
    else
        return Find(root->right, value);
}
AlphaBalancedTreePtr Delete(AlphaBalancedTreePtr root, int value)
{
    root = DeleteWithoutAdjust(root, value);
    root = Adjust(root);
    return root;
}
```

```

}
AlphaBalancedTreePtr DeleteWithoutAdjust(AlphaBalancedTreePtr root,
int value)
{
    AlphaBalancedTreePtr temp = FindIncludeDeleted(root, value);
    if (temp == NULL) return root;
    bool isDeleted = temp->isDeleted;
    if (root == NULL) return root;
    else if (value < root->value)
    {
        root->left = DeleteWithoutAdjust(root->left, value);
        isDeleted ? root->deletedChildNum++ : root->childNum--;
    }
    else if (value > root->value)
    {
        root->right = DeleteWithoutAdjust(root->right, value);
        isDeleted ? root->deletedChildNum++ : root->childNum--;
    }
    else
        root->isDeleted = true;
    return root;
}

AlphaBalancedTreePtr Insert(AlphaBalancedTreePtr root, int value)
{
    root = InsertWithoutAdjust(root, value, NULL);
    root = Adjust(root);
    return root;
}

AlphaBalancedTreePtr InsertWithoutAdjust(AlphaBalancedTreePtr root,
int value, AlphaBalancedTreePtr parent)
{
    AlphaBalancedTreePtr temp = FindIncludeDeleted(root, value);
    if (temp != NULL && !temp->isDeleted) return root;
    bool isDeleted = (temp != NULL);
    if (root == NULL)
    {
        root = CreateNode(value);
        root->parent = parent;
    }
    else if (value < root->value)
    {
        root->left = InsertWithoutAdjust(root->left, value, root);
        isDeleted ? root->deletedChildNum-- : root->childNum++;
    }
}

```

```

        else if (value > root->value)
        {
            root->right = InsertWithoutAdjust(root->right, value, root);
            isDeleted ? root->deletedChildNum-- : root->childNum++;
        }
        else
            root->isDeleted = false;
        return root;
    }
AlphaBalancedTreePtr Adjust(AlphaBalancedTreePtr root)
{
    if (root == NULL) return root;
    else if (isBalanced(root))
    {
        if (root->left != NULL)
            root->left = Adjust(root->left);
        if (root->right != NULL)
            root->right = Adjust(root->right);
        return root;
    }
    else
    {
        AlphaBalancedTreePtr temp = root->parent;
        root = Rebuild(root);
        root->parent = temp;
        if (root->left != NULL) root->left = Adjust(root->left);
        if (root->right != NULL) root->right = Adjust(root->right);
        return root;
    }
}
AlphaBalancedTreePtr Rebuild(AlphaBalancedTreePtr root)
{
    if (root->childNum == 1) return root;
    int* inorderTraversal = GetInorderTraversal(root);
    root = BuildTotallyBalancedTree(inorderTraversal, root->childNum -
root->deletedChildNum + 1);
    UpdateUndeletedChildNum(root);
    return root;
}
AlphaBalancedTreePtr BuildTotallyBalancedTree(int* inorderTraversal,
int num)
{
    AlphaBalancedTreePtr node = CreateNode(inorderTraversal[(num - 1)
/ 2]);

```

```

        if (num == 0)        return NULL;
        node->left = BuildTotallyBalancedTree(inorderTraversal, (num - 1)
/ 2);
        node->right = BuildTotallyBalancedTree(inorderTraversal + (num -
1) / 2 + 1, num - (num - 1) / 2 - 1);
        if (node->left != NULL)node->left->parent = node;
        if (node->right != NULL)node->right->parent = node;
        return node;
    }
void InorderTraversal(AlphaBalancedTreePtr root, int* result, int*
currentIndex)
{
    if (root == NULL || root->isDeleted)return;
    if (root->left != NULL)
        InorderTraversal(root->left, result, currentIndex);
    result[*currentIndex] = root->value;
    (*currentIndex)++;
    if (root->right != NULL)
        InorderTraversal(root->right, result, currentIndex);
    return;
}
float GetBalanceFactor(AlphaBalancedTreePtr root)
{
    if (root == NULL)return 0;
    return root->left == NULL ? 0 : (float)(root->left->childNum + 1)
/ (root->childNum + 1);
}
float GetDeletedScale(AlphaBalancedTreePtr root)
{
    if (root == NULL)return 0;
    return (float)root->deletedChildNum /
(root->childNum + root->deletedChildNum + 1);
}
int* GetInorderTraversal(AlphaBalancedTreePtr root)
{
    int*    inorderTraversal    =    malloc((root->childNum    -
root->deletedChildNum + 1) * sizeof(int));
    int* index = malloc(sizeof(int));*index = 0;
    InorderTraversal(root, inorderTraversal, index);
    return inorderTraversal;
}
int* GetRandomIntArrayWithRange(int num, int min, int max)
{
    if (num == 0)

```

```

    if (num == 0) return NULL;
    srand((unsigned)time(NULL));
    int* result = malloc(num * sizeof(int));
    for (int i = 0; i < num; i++)
        result[i] = rand() % (max - min + 1) + min;
    return result;
}

int UpdateUndeletedChildNum(AlphaBalancedTreePtr root)
{
    int childNum = 0;
    if(root == NULL) return 0;
    childNum += UpdateUndeletedChildNum(root->left);
    childNum += UpdateUndeletedChildNum(root->right);
    root->childNum = childNum;
    return childNum+1;
}

bool isBalanced(AlphaBalancedTreePtr node)
{
    if (node == NULL || node->childNum == 0 || node->childNum == 1)
        return true;
    float balanceFactor = GetBalanceFactor(node);
    return balanceFactor > targetBalanceFactor && balanceFactor < 1 -
targetBalanceFactor;
}

```