## Lecture 4: Balanced Binary Search Trees

## Lecture Overview

- The importance of being balanced
- AVL trees
- Definition
- Balance
- Insert
- Other balanced trees
- Data structures in general


## Readings

CLRS Chapter 13. 1 and 13. 2 (but different approach: red-black trees)
Recall: Binary Search Trees (BSTs)

- rooted binary tree
- each node has
- key
- left pointer
- right pointer
- parent pointer

See Fig. 1


Figure 1: Heights of nodes in a BST


Figure 2: BST property

- BST property (see Fig. 2).
- height of node $=$ length ( $\#$ edges) of longest downward path to a leaf (see CLRS B. 5 for details).


## The Importance of Being Balanced:

- BSTs support insert, min, delete, rank, etc. in $O(h)$ time, where $h=$ height of tree ( $=$ height of root).
- $h$ is between $\lg (n)$ and $n$ : (see Fig. 3).


Figure 3: Balancing BSTs

- balanced BST maintains $h=O(\lg n) \Rightarrow$ all operations run in $O(\lg n)$ time.


## AVL Trees:

## Definition

AVL trees are self-balancing binary search trees. These trees are named after their two inventors G.M. Adel'son-Vel'skii and E.M. Landis $\square$
An AVL tree is one that requires heights of left and right children of every node to differ by at most $\pm 1$. This is illustrated in Fig. (4)


Figure 4: AVL Tree Concept
In order to implement an AVL tree, follow two critical steps:

- Treat nil tree as height -1 .
- Each node stores its height. This is inherently a DATA STRUCTURE AUGMENTATION procedure, similar to augmenting subtree size. Alternatively, one can just store difference in heights.

A good animation applet for AVL trees is available at this link. To compare Binary Search Trees and AVL balancing of trees use code provided here.

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## Balance:

The balance is the worst when every node differs by 1 .
Let $N_{h}=\min (\sharp$ nodes $)$.

$$
\begin{aligned}
\Rightarrow N_{h} & =N_{h-1}+N_{h-2}+1 \\
& >2 N_{h-2} \\
\Rightarrow N_{h} & >2^{h / 2} \\
\Longrightarrow h & <2 \lg N_{h}
\end{aligned}
$$

Alternatively:

$$
\begin{array}{rlrl}
N_{h} & >F_{n} & & \left(n^{\text {th }}\right. \text { Fibonacci number) } \\
\text { In fact, } N_{h} & =F_{n+2}-1 & & \text { (simple induction) } \\
F_{h} & =\frac{\phi^{h}}{\sqrt{ } 5} & & \text { (rounded to nearest integer) } \\
\text { where, } \phi & =\frac{1+\sqrt{ } 5}{2} \approx 1.618 & & \text { (golden ratio) } \\
\Longrightarrow \max h & \approx \log _{\phi}(n) \approx 1.440 \lg (n) &
\end{array}
$$

## AVL Insert:

1. insert as in simple BST.
2. work your way up tree, restoring AVL property (and updating heights as you go).

Each Step:

- suppose $x$ is lowest node violating AVL
- assume $x$ is right-heavy (left case symmetric)
- if x's right child is right-heavy or balanced: follow steps in Fig. 5
- else follow steps in Fig. 6
- then continue up to x's grandparent, greatgrandparent ...


Figure 5: AVL Insert Balancing (FIX: Node z should be y)


Figure 6: AVL Insert Balancing

Example: An example implementation of the AVL Insert process is illustrated in Fig. 7 .


Figure 7: Illustration of AVL Tree Insert Process. Note that node x is left-heavy.

Comment 1. In general, process may need several rotations before an Insert is completed.
Comment 2. Delete(-min) harder but possible.

## Balanced Search Trees:

There are many balanced search trees.

| AVL Trees | Adel'son-Velsii and Landis 1962 |
| :--- | :--- |
| B-Trees/2-3-4 Trees | Bayer and McCreight 1972 (see CLRS 18) |
| BB $[\alpha]$ Trees | Nievergelt and Reingold 1973 |
| Red-black Trees | CLRS Chapter 13 |
| Splay-Trees | Sleator and Tarjan 1985 |
| Skip Lists | Pugh 1989 |
| Scapegoat Trees | Galperin and Rivest 1993 |
| Treaps | Seidel and Aragon 1996 |

Note 1. Skip Lists and Treaps use random numbers to make decisions fast with high probability.
Note 2. Splay Trees and Scapegoat Trees are "amortized": adding up costs for several operations $\Longrightarrow$ fast on average.

## Splay Trees

Upon access (search or insert), move node to root by sequence of rotations and/or doublerotations (just like AVL trees). Height can be linear but still $O(\lg n)$ per operation "on average" (amortized)

Note: We will see more on amortization in a couple of lectures.

## Optimality

- For BSTs, cannot do better than $O(\lg n)$ per search in worst case.
- In some cases, can do better e.g.
- in-order traversal takes $\Theta(n)$ time for $n$ elements.
- put more frequent items near root

A Conjecture: Splay trees are O (best BST) for every access pattern.

- With fancier tricks, can achieve $O(\lg \lg u)$ performance for integers $1 \cdots u$ [Van Ernde Boas; see 6.854 or 6.851 (Advanced Data Structures)]


## Big Picture:

Abstract Data Type(ADT): interface spec.
e.g. Priority Queue:

- $\mathrm{Q}=$ new-empty-queue()
- Q.insert(x)
- $\mathrm{x}=$ Q.deletemin()
vS.
Data Structure (DS): algorithm for each op.

There are many possible DSs for one ADT. One example that we will discuss much later in the course is the "heap" priority queue.


[^0]:    ${ }^{1}$ Original Russian article: Adelson-Velskii, G.; E. M. Landis (1962). "An algorithm for the organization of information". Proceedings of the USSR Academy of Sciences 146: 263266. (English translation by Myron J. Ricci in Soviet Math. Doklady, 3:12591263, 1962.)

