## Splay Trees

## Static Optimality

## Balanced BSTs

- We've explored balanced BSTs this quarter because they guarantee worst-case $\mathrm{O}(\log n)$ operations.
- Claim: Depending on the access sequence, balanced BSTs may not be optimal BSTs.



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## Static Optimality

- Let $S=\left\{x_{1}, x_{2}, \ldots, x_{n}\right\}$ be a set with access probabilities $p_{1}, p_{2}, \ldots, p_{n}$.
- Goal: Construct a binary search tree $T^{*}$ that minimizes the total expected access time.
- $T^{*}$ is called a statically optimal binary search tree.


## Static Optimality

- There is an $\mathrm{O}\left(n^{2}\right)$-time dynamic programming algorithm for constructing statically optimal binary search trees.
- Knuth, 1971
- There is an $\mathrm{O}(n \log n)$-time greedy algorithm for constructing binary search trees whose cost is within 1.5 of optimal.
- Mehlhorn, 1975
- These algorithms assume that the access probabilities are known in advance.

Challenge: Can we construct an optimal BST without knowing the access probabilities in advance?

## The Intuition

- If we don't know the access probabilities in advance, we can't build a fixed BST and then "hope" it works correctly.
- Instead, we'll have to restructure the BST as operations are performed.
- For now, let's focus on lookups; we'll handle insertions and deletions later on.


## Refresher: Tree Rotations



## An Initial Idea

- Begin with an arbitrary BST.
- After looking up an element, repeatedly rotate that element with its parent until it becomes the root.
- Intuition:
- Recently-accessed elements will be up near the root of the tree, lowering access time.
- Unused elements stay low in the tree.


## The Problem



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## The Problem

- The "rotate to root" method might result in $n$ accesses taking time $\Theta\left(n^{2}\right)$.
- Why?
- Rotating an element $x$ to the root significantly "helps" $x$, but "hurts" the rest of the tree.
- Most of the nodes on the access path to $x$ have depth that increases or is unchanged.


## A More Balanced Approach

- In 1983, Daniel Sleator and Robert Tarjan invented an operation called splaying.
- Rotates an element to the root of the tree, but does so in a way that's more "fair" to other nodes in the tree.
- There are three cases for splaying.


## Case 1: Zig-Zig



## Case 2: Zig-Zag




First, rotate $x$ with $p$
Then, rotate $x$ with $g$
Continue moving $x$ up the tree

## Case 3: Zig

(Assume $r$ is the tree root)


## Splaying, Empirically

- Splaying nodes that are deep in the tree tends to correct the tree shape.
- Why is this?
- Is this a coincidence?


## Why Splaying Works

- Claim: After doing a splay at $x$, the average depth of any nodes on the access path to $x$ is halved.
- Intuitively, splaying $x$ benefits nodes near $x$, not just $x$ itself.
- This "altruism" will ensure that splays are efficient.

The average depth of $x$, $p$, and $g$ is unchanged.


The average height of $x, p$, and $g$ decreases by $1 /{ }_{3}$.


There is no net change in the height of $x$ or $r$.


## An Intuition for Splaying

- Each rotation done only slightly penalizes each other part of the tree (say, adding +1 or +2 depth).
- Each splay rapidly cuts down the height of each node on the access path.
- Slow growth in height, combined with rapid drop in height, is a hallmark of amortized efficiency.


## Making Things Easy

- Splay trees provide make it extremely easy to perform the following operations:
- lookup
- insert
- delete
- predecessor / successor
- join
- split
- Let's see why.


## Lookups

- To do a lookup in a splay tree:
- Search for that item as usual.
- If it's found, splay it up to the root.
- Otherwise, splay the last-visited node to the root.


## Insertions

- To insert a node into a splay tree:
- Insert the node as usual.
- Splay it up to the root.


## Join

- To join two trees $T_{1}$ and $T_{2}$, where all keys in $T_{1}$ are less than the keys in $T_{2}$ :
- Splay the max element of $T_{1}$ to the root.
- Make $T_{2}$ a right child of $T_{1}$.



## Split

- To split $T$ at a key $k$ :
- Splay the successor of $k$ up to the root.
- Cut the link from the root to its left child.



## Delete

- To delete a key $k$ from the tree:
- Splay $k$ to the root.
- Delete $k$.
- Join the two resulting subtrees.



## The Runtime

- Claim: All of these operations require amortized time O(log $n$ ).
- Rationale: Each has runtime bounded by the cost of O(1) splays, which takes total amortized time $\mathrm{O}(\log n)$.
- Contrast this with red/black trees:
- No need to store any kind of balance information.
- Only three rules to memorize.

