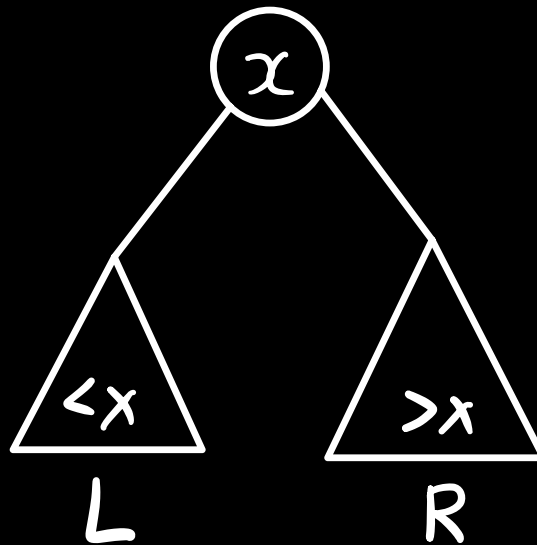


Splay Trees

Roger Fu

Binary Search Trees

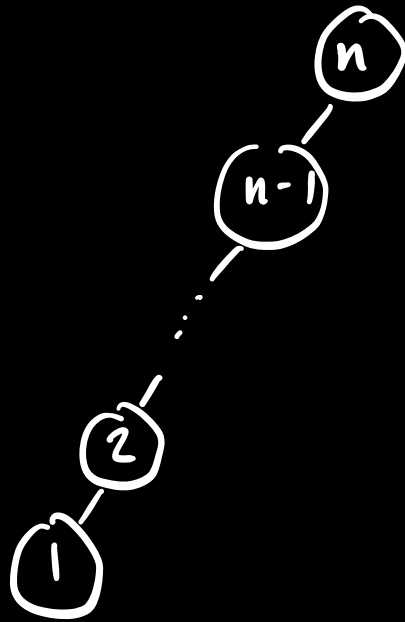
· Empty; or tuple (x, L, R) :



x is the key, L the left subtree, R the right subtree

Binary search tree property: all keys in $L < x$, all keys in $R > x$.

Binary Search Tree: A Weakness



Searching for $1, 2, \dots, n-1, n$ takes

$$n + (n-1) + \dots + 2 + 1 = \frac{n(n+1)}{2} \text{ operations}$$

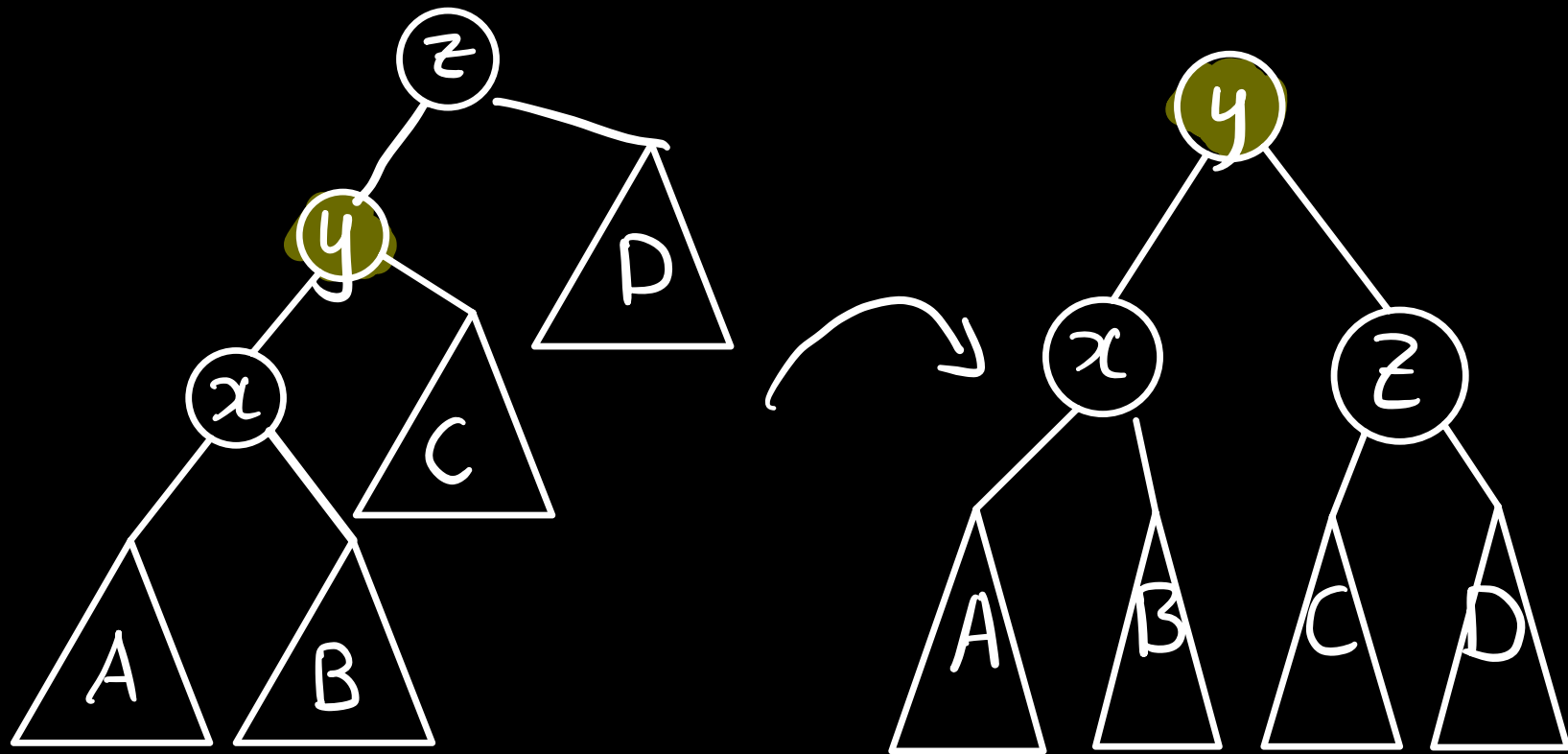
Idea: introduce splay trees to get subquadratic time

Splay Trees

Intuition: make recently accessed elements easy to access again

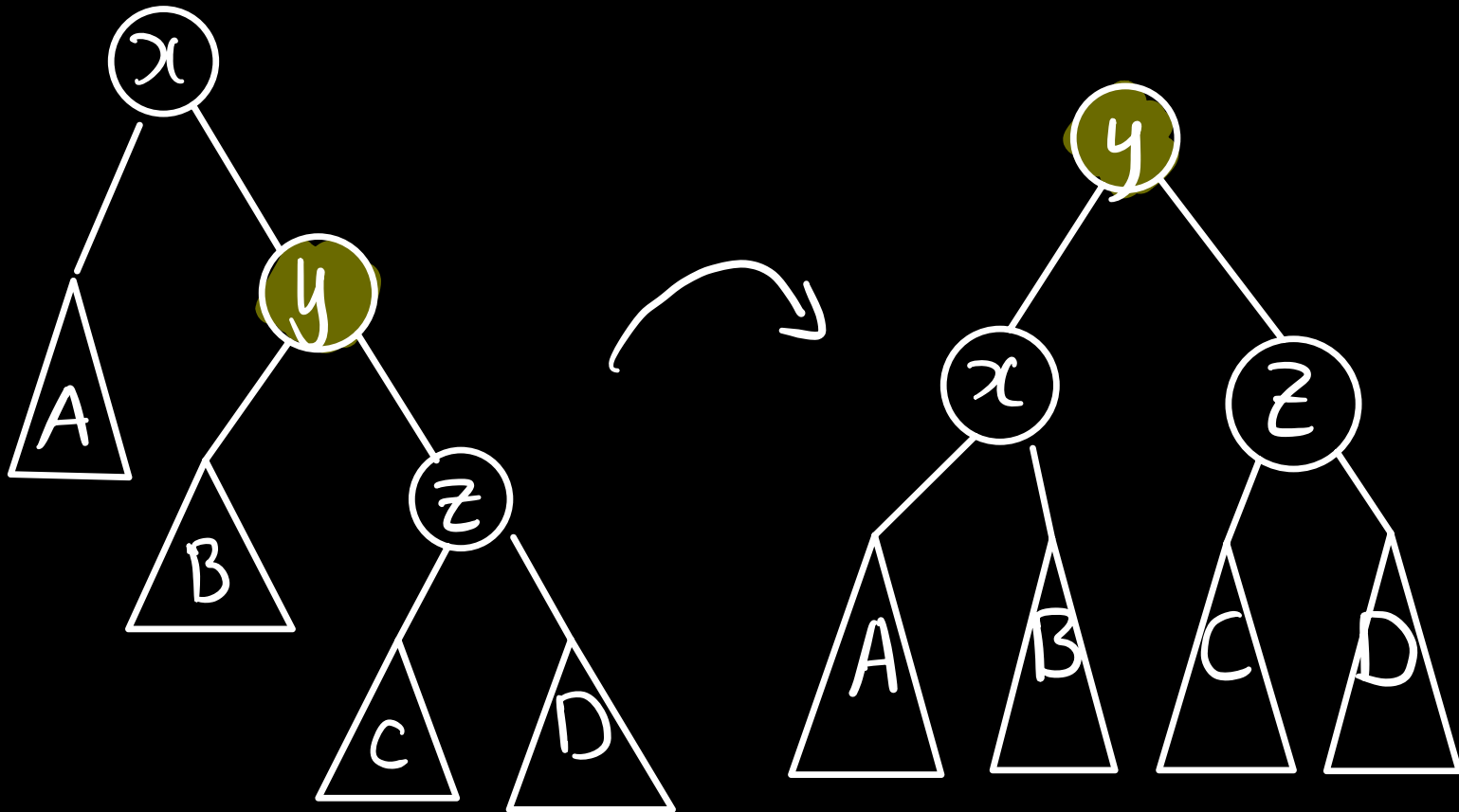
We will **splay** the last accessed node to the root after each operation

Rotation



Right rotation

Rotation



Left rotation

Rotating Towards Root

If x is left (right) child of p , we say a right (left) rotation rotates x towards the root.

↳ x replaces p as root

Naive Attempt: Rotate-to Root

Natural 1st idea: repeatedly rotate node x to root

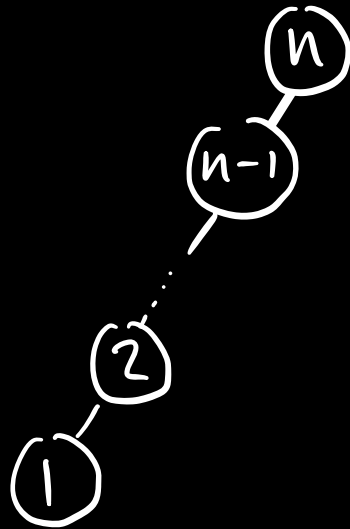
Denote as $\text{rtr}(x)$

Naive Attempt: Rotate-to Root

Natural 1st idea: repeatedly rotate node x to root

Denote as $rtr(x)$

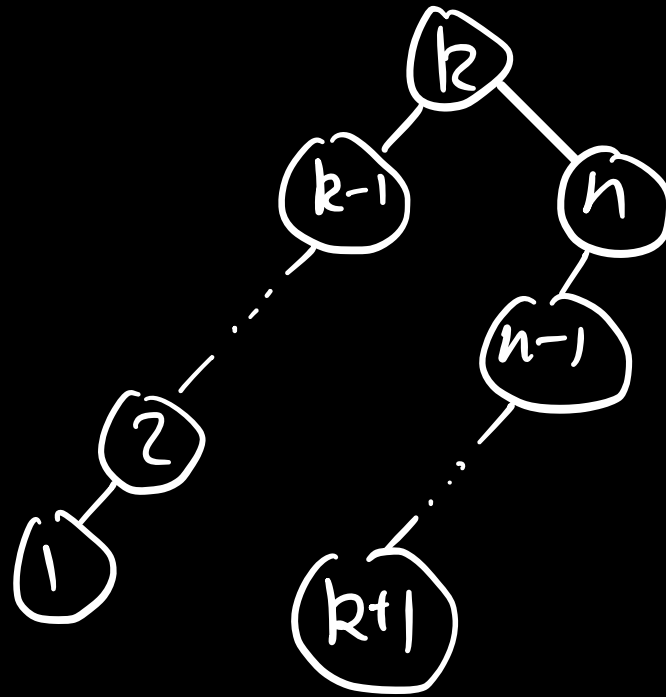
Problem:



Consider $rtr(1), \dots, rtr(n)$

Naive Attempt: Rotate-to-Root

Claim: After $\text{rtr}(1), \dots, \text{rtr}(k)$ we get



Pf: Induct.

Note that $\text{rtr}(k+1)$ takes $\Theta(n-k)$ rotations: total: $\Theta(n^2)$

The Splay Operation:

splay(x):

while x is not the root:

$p \leftarrow x.$ parent

if p is the root:

rotate x towards root

else:

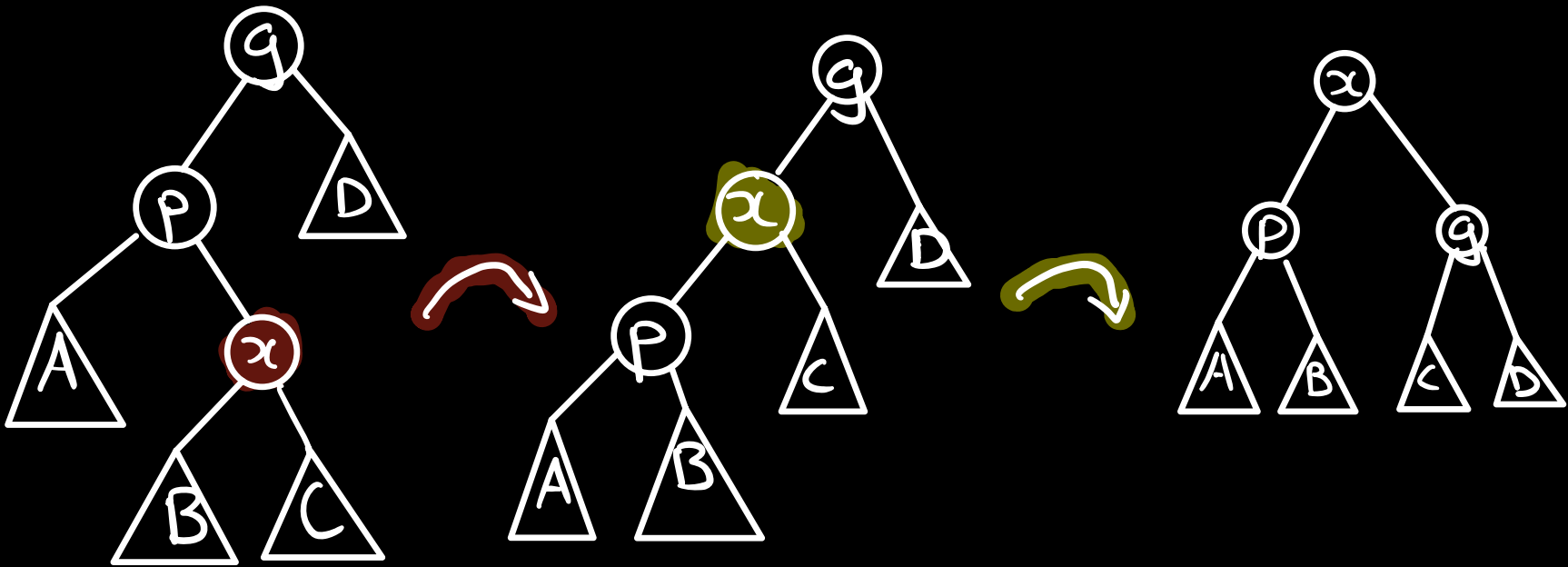
if p and x both left/right children:

zig-zig(x)

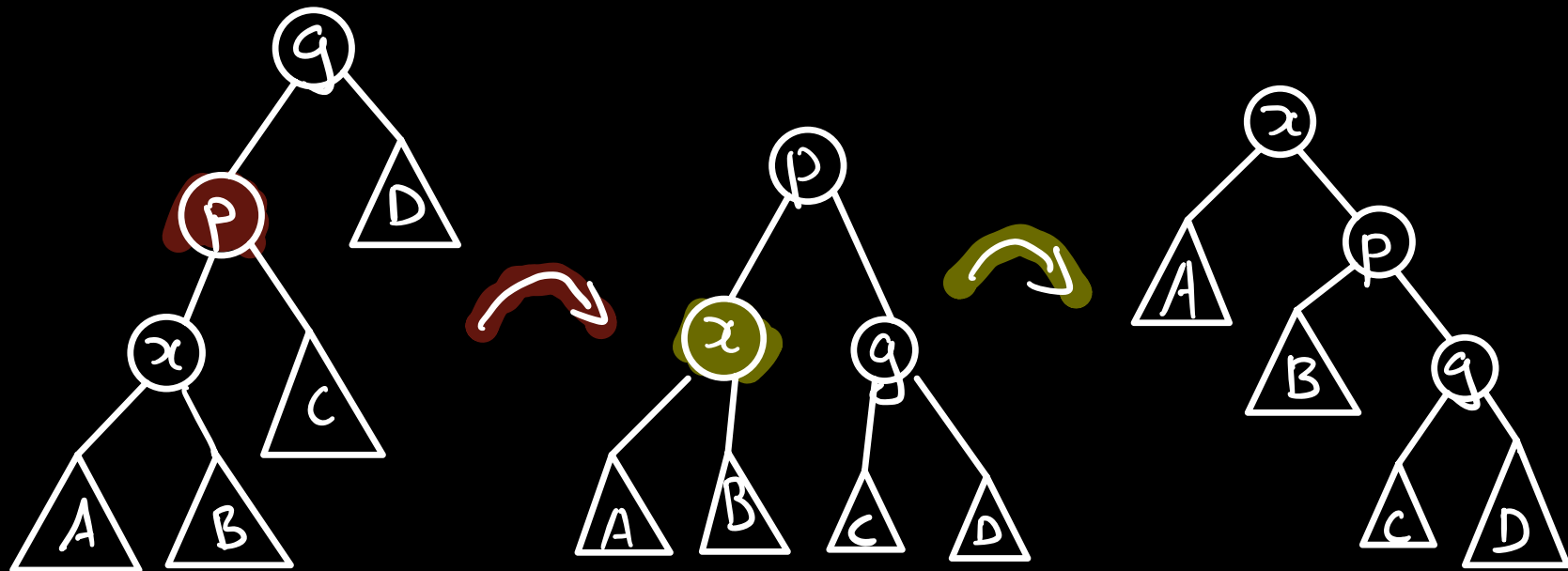
else:

zig-zag(x)

Zig-zag



Zig-zig

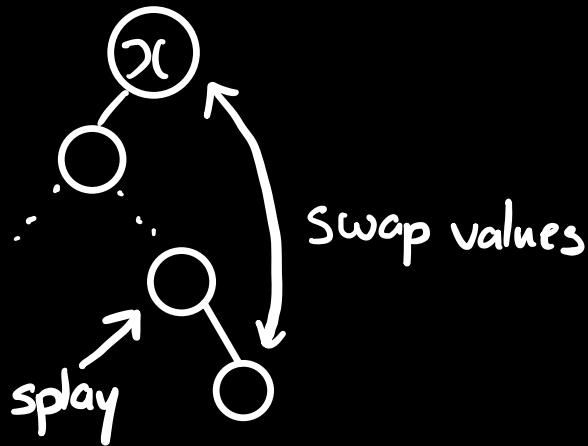


Remark: This is the difference between rotate-to-root & splay

Binary Search Tree Operations

Usual binary search tree operations augmented by splaying last accessed node (insertion, searching)

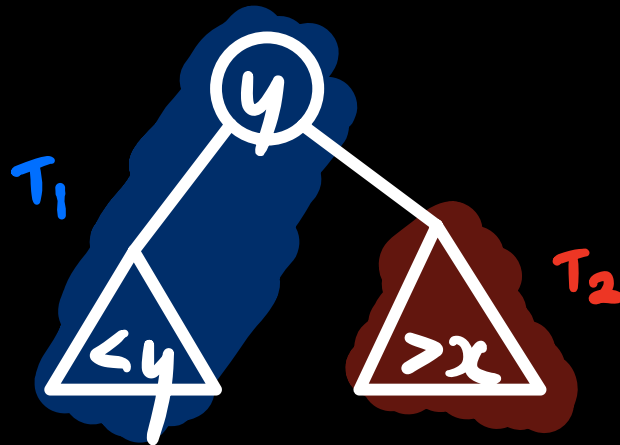
Deletion: if node is deleted, splay the parent:



Splitting

Split: given splay tree T , partition into $T = T_1 \cup T_2$ where all keys in $T_1 \leq x$, all keys in $T_2 > x$

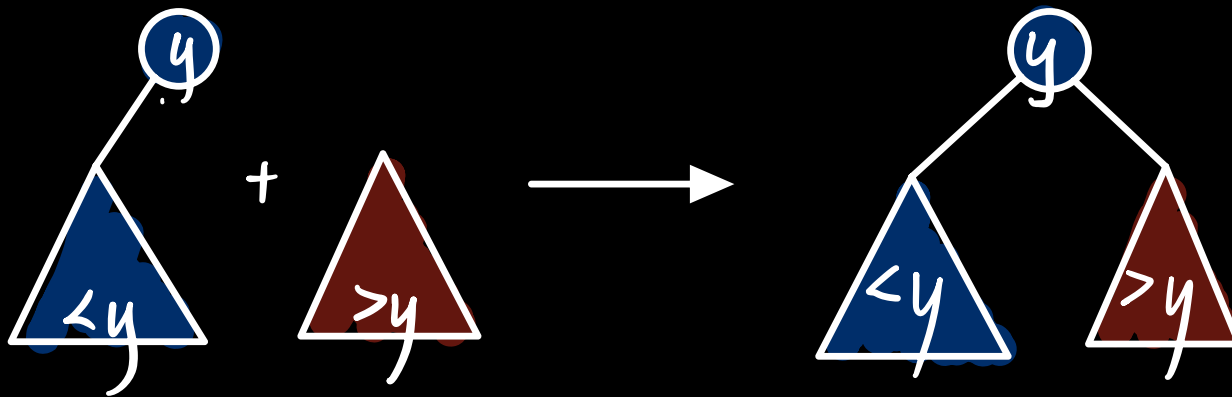
1. Find largest $y \leq x$ s.t. $y \in T$
2. Splay y to root
3. Partition right subtree into T_2 :



Merging

Merge: given T_1, T_2 where $\max(T_1) < \min(T_2)$, combine to form $T = T_1 \cup T_2$.

1. Splay largest element of T_1 to root
2. Attach T_2 as right child of T_1



Time Complexity

- Time complexity dominated by cost of searching for splaying node
- Problem: runtime of splaying depends on structure of tree, which depends on previous operations done
 - ↳ Can establish **amortized bound** of $O(\log n)$

Detour: Amortized Analysis

Let $T^{\text{actual}}(\mathcal{O})$ runtime of operation \mathcal{O} & $T^{\text{amort}}(\cdot)$ function on operations. $T^{\text{amort}}(\cdot)$ upper bounds the amortized run-time if for any sequence of operations $\mathcal{O}_1, \dots, \mathcal{O}_k$ we have

$$\sum_{i=1}^k T^{\text{actual}}(\mathcal{O}_i) \leq \sum_{i=1}^k T^{\text{amort}}(\mathcal{O}_i).$$

Intuition: Amortized time complexity is like average upper bound

Splay Trees & Range Queries I

Range query problem: given list x_1, \dots, x_n support:

1. Calculating $f(x_i, x_{i+1}, \dots, x_j)$
2. Changing x_i .

Will focus on case where there exists g such that

$$g(g(f(x_i, \dots, x_{k-1}), x_k), f(x_{k+1}, \dots, x_j))$$

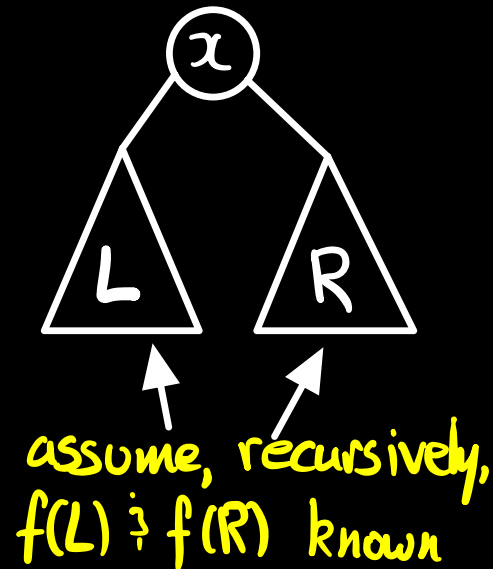
Example:

- f is the sum
- f is the max
- f is max subarray sum

Splay Trees : Range Queries II

Augment splay tree T by storing $f(T)$ at root node

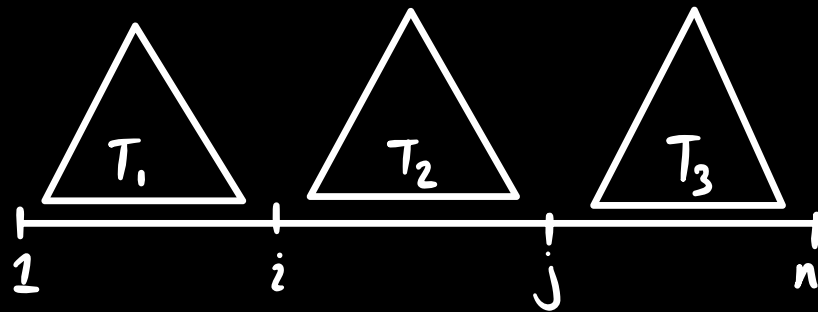
Combining: $g(g(f(L), x), f(R))$



Note: when rotating, f needs to be recomputed

Splay Tree : Range Queries III

1. Construct splay tree w/ keys $1, 2, \dots, n$
2. At node i , store x_i and for each subtree T , compute $f(T)$
3. To query $f(x_i, \dots, x_j)$, split on $i-1$ and j :



Return $f(T_2)$ (and merge the trees back together)

Splay Tree & Range Queries II

4. To update x_i , splay i^{th} node to root & update x_i .
↳ Only need to recompute $f(T)$.

Time Complexity:

Operations dominated by cost of splaying.

Amortized $O(c \log n)$ where c is cost of $g(\cdot)$

Warning! Splay tree is generally slower than segment tree!

*** JUST BECAUSE BOTH ARE $O(\log n)$
DOES NOT MAKE THEM INTERCHANGABLE!**

Implicitly Keyed Splay Trees

Idea: Instead of explicitly using $1, \dots, n$ as keys, use order in in-order traversal as key

Advantages: Supports modifying underlying list:

- Insertion at arbitrary indices
- Deletion of arbitrary indices
- Moving subarrays around
- Reversing subarrays

Implicitly Keyed Splay Trees

```
def get_val(x: node, pos: int):  
    if sz(x.L) + 1 = pos:  
        splay(x)  
        return x  
    else if sz(x.L) + 1 < pos:  
        return get_val(x.R, pos - (sz(x.L) + 1))  
    else:  
        return get_val(x.L, pos)
```

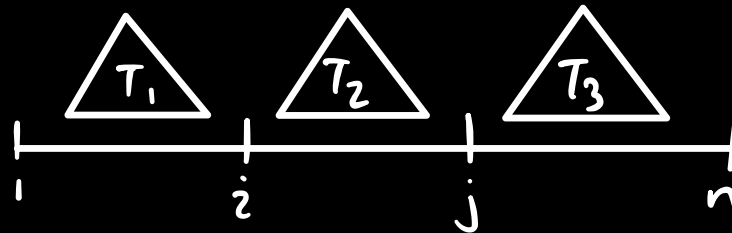
Remark: You can similarly modify **split**. **Join** becomes equivalent to concatenation.

Lazy Propagation

Idea: Instead of applying update to range, update the stored aggregate value and set flag to propagate changes to children

Example: Range sum update & query

- For each subtree T store sum $f(T)$
- To query sum, split:



and return $f(T_2)$

- To update, split and update lazy propagation flag on T_2 (and update $f(T_2)$)

Reversing Ranges

- Lazy propagation to swap left + right child
- Be careful if your aggregate function f is not commutative!
 - ↳ **Example:** largest prefix sum

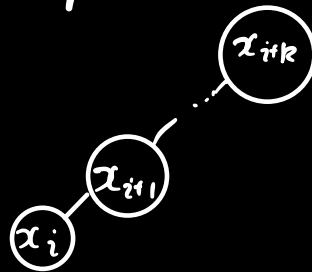
Practice Problems

- spoj.com/problems/SEQ2
- dmoj.ca/problem/ccol6p6
- dmoj.ca/problem/ds4

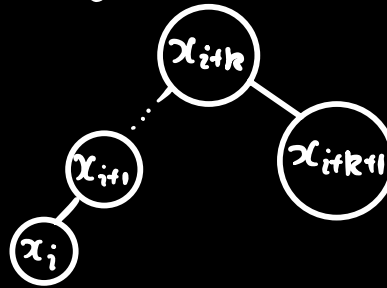
Hints

- SEQ2: you should aim to insert k consecutive numbers in $O(k + \log n)$

Verify inductively the inserted elements form a tree like:



and inserting x_{i+k+1} gives



Bonus: Size-Balanced Tree

- Balanced binary search tree that balances itself by checking invariant on subtree sizes.

- Advantages over splay tree:

- ↳ Doesn't store extra data for rebalancing (splay tree needs parent pointers)

- ↳ "Tendency of perfect BST in practice"

- Disadvantages compared to splay tree:

- ↳ Not clear how to implement split/merge