

## Other/Better Criteria?

Expected case: Some keys more popular than others

Self-adjusting: Tree adapts as popularity changes

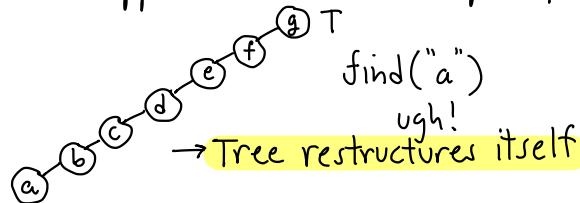
## How to design/analyze?

Splay Tree: A self-adjusting binary search tree

- No rules! (yay anarchy!)
  - No balance factors
  - No limits on tree height
  - No colors/levels/priorities

- Amortized efficiency:
  - Any single op - slow
  - Long series - efficient on avg.

Intuition: Let  $T$  be an unbalanced BST + suppose we access its deepest key



Recap: Lots of search trees

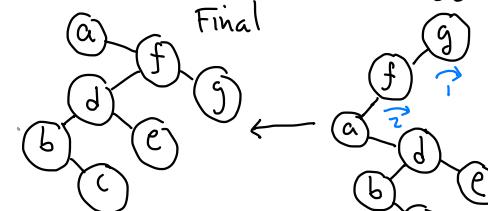
- Unbalanced BSTs
- AVL Trees
- 2-3, Red-black, AA Trees
- Treaps + Skip lists

→ Focus: Worst-case or randomized expected case

## SPLAY TREES I

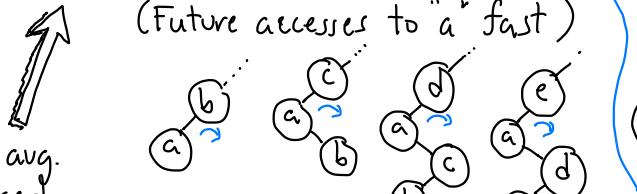
Lesson: Different combinations of rotations can:

- bring given node to root
- significantly change (improve) tree structure.

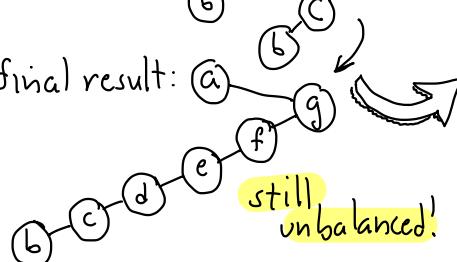


Tree's height has reduced by ~ half!

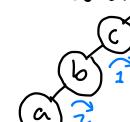
Idea I: Rotate "a" to top  
(Future accesses to "a" fast)



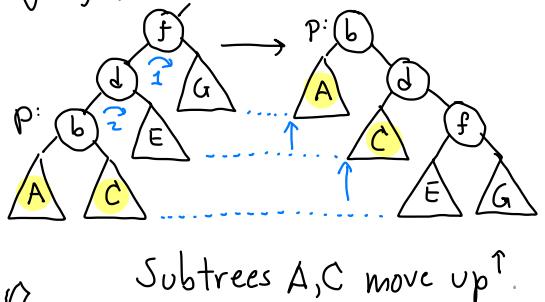
....final result:



Idea II: Rotate 2 at a time - upper + lower



ZigZig(p): [LL case]



Splay(Key x):

Node p  $\leftarrow$  find x by standard BST search  
while ( $p \neq$  root) {

if ( $p ==$  child of root) zig(p)

else /\* p has grand parent \*/

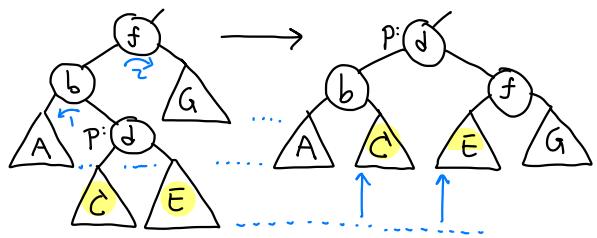
if ( $p$  is LL or RR grand child) zigzag(p)

else /\* p is LR or RL gr. child\*/ zigzag(p)

insert(x):

Node p  $\leftarrow$  splay(x)  
if ( $p.key == x$ ) Error!!  
q  $\leftarrow$  new Node(x)  
if ( $p.key < x$ )  
q.left  $\leftarrow$  p  
q.right  $\leftarrow$  p.right  
p.right  $\leftarrow$  null  
else ... (symmetrical)...  
root  $\leftarrow$  q

ZIGZAG(p): [LR case]

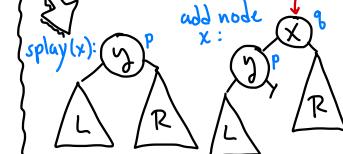


Splay Trees II

find(x):

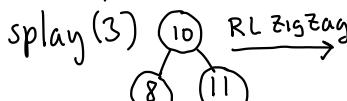
root  $\leftarrow$  splay(x)  
if ( $root.key == x$ )  
return  $root.value$   
else return null

insert(x): add node x:



Example:

splay(3)



RL zigzag



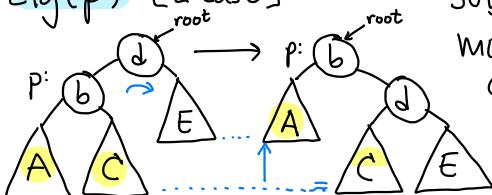
LL zigzag

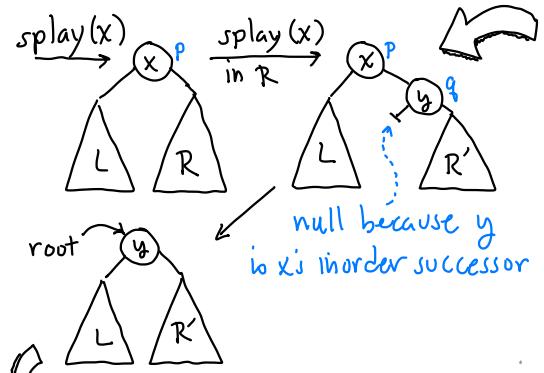


L zig



Zig(p): [L case]





**delete(x):**

- splay(x) [x now at root]
- p = root
- if (p.key ≠ x) **error!**
- splay(x) in p's right subtree
- q = p.right [q's key is x's successor]
- q.left = p.left [q.left == null]
- root = q

**Dynamic Finger Theorem:**  
Keys:  $x_1 < \dots < x_n$ . We perform accesses  $x_{i_1}, x_{i_2}, \dots, x_{i_m}$   
Let  $\Delta_j = i_j - i_{j-1}$ : distance between consecutive items

**Thm:** Total access time is  $O(m + n \log n + \sum_{j=1}^m (1 + \lg \Delta_j))$

### Analysis:

- Amortized analysis
- Any one op might take  $\Theta(n)$
- Over a long sequence, average time is  $\Theta(\log n)$  each
- Amortized analysis is based on a sophisticated potential argument
- Potential: A function of the tree's structure
- Balanced  $\Rightarrow$  Low potential.
- Unbalanced  $\Rightarrow$  High potential
- Every operation tends to reduce the potential

### SPLAY TREES III

Splay Trees are Amazingly Adaptive!

**Balance Theorem:** Starting with an empty dictionary, any sequence of  $m$  accessed takes total time  $\Theta(m \log n + n \log n)$  where  $n = \max.$  entries at any time.

### Static Optimality:

- Suppose key  $x_i$  is accessed with prob  $p_i$ :  $(\sum_{i=1}^n p_i = 1)$
- **Information Theory:** Best possible binary search tree answers queries in expected time  $\Theta(H)$  where  $H = \sum p_i \lg \frac{1}{p_i}$   $\leftarrow$  Entropy

**Static Optimality Theorem:** Given a seq. of  $m$  ops. on splay tree with keys  $x_1, \dots, x_n$ , where  $x_i$  is accessed  $q_i$  times. Let  $p_i = q_i/m$ . Then total time is  $\Theta(m \sum p_i \lg \frac{1}{p_i})$