# Binary Search Trees: Splay Trees

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# Data Structures Data Structures and Algorithms

#### Learning Objectives

- Implement a splay tree.
- Understand the ideas behind the runtime analysis.
- Know some other properties of splay tree runtimes.

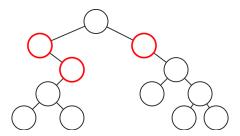
# Outline

## Non Uniform Inputs

Search for random elements  $O(\log(n))$  best possible.

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- Search for random elements  $O(\log(n))$  best possible.
- If some items more frequent than others, can do better putting frequent queries near root.



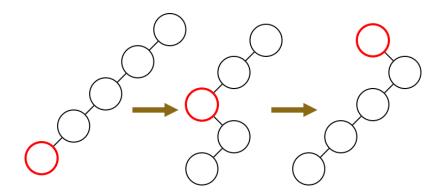
#### Idea

Bring query node to the root.

# Simple Idea

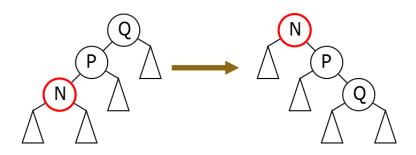
Just rotate to top.

Doesn't work



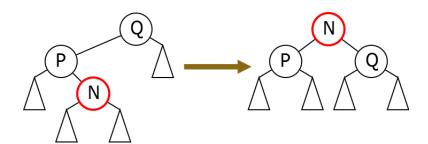
## Modification

Zig-Zig



## Modification

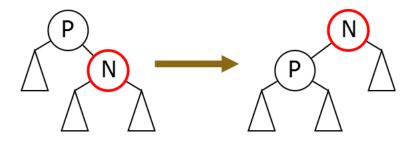
Zig-Zag



#### Modification

If just below root:

Zig



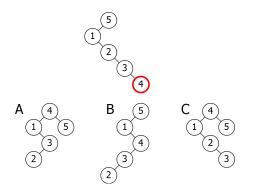
# Splay

#### Splay(N)

```
Determine proper case
Apply Zig-Zig, Zig-Zag, or Zig as
appropriate
if N.Parent ≠ null:
Splay(N)
```

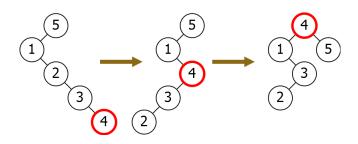
#### Problem

Which of the following is the result of splaying the highlighted node?



#### Problem

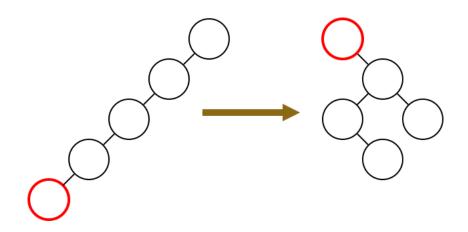
Which of the following is the result of splaying the highlighted node?



# Outline

#### Sometimes Slow

Splay operation is sometimes slow:



## Amortized Analysis

Need to amortize. Pick correct potential function.

#### Rank

 $R(N) = \log_2(\text{Size of subtree of } N).$ Potential function

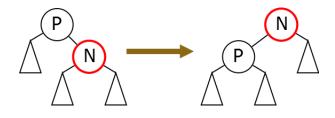
$$\Phi = \sum_{N} R(N).$$

## Zig Analysis

$$\Delta\Phi = R'(N) + R'(P) - R(N) - R(P)$$

$$= R'(P) - R(N)$$

$$\leq R'(N) - R(N).$$



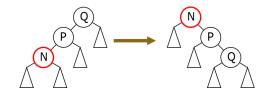
## Zig-Zig Analysis

$$\Delta\Phi = R'(N) + R'(P) + R'(Q)$$

$$- R(N) - R(P) - R(Q)$$

$$= (R'(P) - R(P)) + (R'(Q) - R(N))$$

$$\leq 3(R'(N) - R(N)) - 2$$



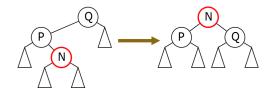
## Zig-Zag Analysis

$$\Delta\Phi = R'(N) + R'(P) + R'(Q)$$

$$- R(N) - R(P) - R(Q)$$

$$= (R'(P) - R(P)) + (R'(Q) - R(N))$$

$$\leq 2(R'(N) - R(N)) - 2$$



## Total Change

$$\Delta \Phi \le 3(R_k(N) - R_{k-1}(N)) - 2$$
 $+ 3(R_{k-1}(N) - R_{k-2}(N)) - 2 + \cdots$ 
 $= 3(R'(N) - R(N)) - \Omega(\text{Depth}(N))$ 
 $= O(\log(n)) - \text{Work}$ 

Amortized cost of Find+Splay is  $O(\log(n))$ .

# Outline

## Find

```
STFind(k, R)
```

 $N \leftarrow \text{Find}(k, R)$ Splay(N) return N

#### Insert

Insert, then splay

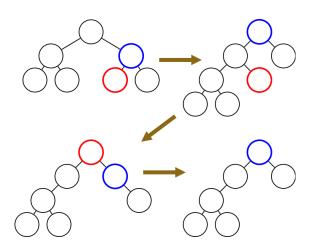
STInsert(k, R)

Insert(k, R)

 $\mathsf{STFind}(k,R)$ 

#### Delete

Bring N and successor to top. Deletes easily.



#### Delete

#### STDelete(N)

Splay(Next(N))
Splay(N)
Delete(N)

## Split

```
STSplit(R, x)
```

```
N \leftarrow \text{Find}(x, R)

\text{Splay}(N)

\text{split off appropriate subtree of } N
```

## Merge

## $STMerge(R_1, R_2)$

$$N \leftarrow ext{Find}(\infty, R_1)$$
  
 $ext{Splay}(N)$   
 $N. ext{Right} \leftarrow R_2$ 

# Summary

Performs all operations in  $O(\log(n))$  amortized time.

# Outline

#### Other Bounds

Splay trees have many other wonderful properties.

## Weighted Nodes

If you assign weights so that

$$\sum_{N} \operatorname{wt}(N) = 1,$$

accessing N costs  $O(\log(1/\mathrm{wt}(N)))$ .

# Dynamic Finger

Cost of accessing node  $O(\log(D+1))$  where D is distance between last access and current access.

# Working Set Bound

Cost of accessing N is  $O(\log(t+1))$  where t is time since N was last accessed.

# Dynamic Optimality Conjecture

It is conjectured that for any sequence of binary search tree operations that a splay tree does at most a constant factor more work than the best search tree for that sequence.

#### Conclusion

#### Splay Trees

- Easy to implement.
- $O(\log(n))$  time per operation.
- Can be much better if queries have structure.