

# Search Trees

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Graphs and Trees

Binary Search Trees

AVL-Trees

(a,b)-Trees

Splay-Trees

## Definition

An (undirected) graph  $G = (V, E)$  is defined by a set of nodes  $V$  and a set of edges  $E$ .

$$E \subseteq \binom{V}{2} := \{X : X \subseteq V, |X| = 2\}$$

A directed graph  $G = (V, E)$  is given by a set of nodes and a set of directed edges:

$$E \subseteq V \times V$$

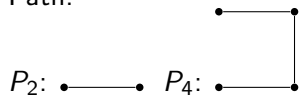
## Definition

The neighborhood of node  $x$  is given by:

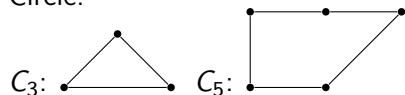
$$N(x) = \{y : x \in V, \{x, y\} \in E\}$$

# Special Graphs

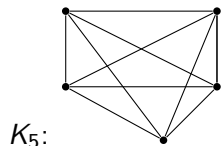
Path:



Circle:



Complete graph/ Clique:



## Definition

A graph  $G = (V, E)$  is called connected, if there is a path from each node  $x$  to each other node  $y$ .

## Definition

A graph  $H = (W, F)$  is called subgraph of  $G = (V, E)$  if

$$W \subseteq V \text{ and } F \subseteq E.$$

## Definition

An acyclic graph  $G = (V, E)$  does not contain any circle as a subgraph.

## Definition

A graph  $G = (V, E)$  is called a tree if it is connected and acyclic.

## Definition

A rooted binary tree  $G = (V, E)$  is a tree with one root node  $r$ .

$$\begin{aligned} |N(r)| &< 3 & r \in V \\ 1 \leq |N(x)| &\leq 3 & \forall x \in V \setminus \{r\} \end{aligned}$$

## Definition

The height of a tree  $G = (V, E)$  with root  $r \in V$  is defined as

$$h = \max_{x \in V} \{\text{distance from } r \text{ to } x\}$$

## Theorem

*The following definitions of a tree  $G = (V, E)$  are equivalent*

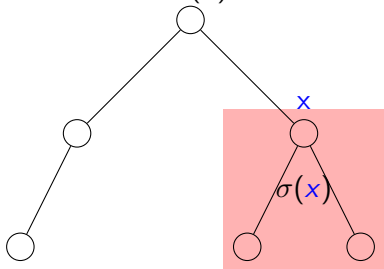
- ▶  *$G$  is connected and acyclic.*
- ▶  *$G$  is connected and  $|V| = |E| + 1$ .*
- ▶  *$G$  is acyclic and  $|V| = |E| + 1$ .*
- ▶ *When adding a new edge to  $G$  the resulting graph will contain a circle.*
- ▶ *When removing an edge from  $G$  the resulting graph is not connected anymore.*
- ▶ *For all two nodes  $x, y \in V$  and  $x \neq y$  there is exactly one path from  $x$  to  $y$ .*

## Definition

A tree  $H = (W, F)$  is called a spanning tree of a graph  $G = (V, E)$  if  $W = V$  and  $F \subseteq E$ .

## Definition

The function  $\sigma(x)$  returns the subtree, which is rooted in  $x$ :





Problem:

For a set of items  $x_1, \dots, x_n$  where each dataset consists of a key and a value, we want to minimize the total access time on an arbitrary sequence of operations.

One operation can perform

- ▶ a test if a key is stored in the data structure (IsElement),
- ▶ the insertion of an item in the data structure (Insert)
- ▶ or a deletion of a key in the data structure (Delete).

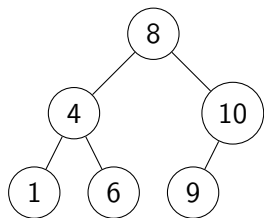
- ▶ An internal search tree stores all keys in internal nodes. The leaves contain no further information. Accordingly there is no need to store them and they can be represented by NIL-pointers.
- ▶ In an external search tree, all keys are stored at the leaves. The internal nodes only contain information for managing the data structure.

A binary search tree is a binary tree, whose internal nodes contain the keys  $k = x.key \forall x \in S$ . For each node  $x$  the following equation must hold if node  $y$  is in the left subtree of  $x$  and node  $z$  is in the right subtree of node  $x$ :

$$y.key < x.key < z.key$$

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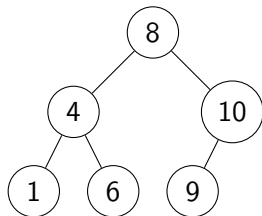
$$y.key < x.key < z.key$$



For making algorithms more understandable, here are more definitions.

A node  $v$  of a search tree stores several values:

- ▶ key – key of the stored item
- ▶ leftChild, rightChild which are pointers to left/right child (only if it is a binary tree)
- ▶ children, the number of children



The items are accessible in pseudocode as follows:

`k=v.key // stores 8 in k if v is the root`

```
IsElement(T,k)
{
  v:=T.root
  while(v!=NIL)
  {
    if(v.key==k)
      return v
    else if(v.key>k)
      v=v.leftChild
    else
      v=v.rightChild
  }
  return v
}
```

```
Insert(T,k)
{
  v=IsElement(T,k)
  if (v==NIL)
  {
    // Inserts a node, updates pointers
    add a node w with w.key=k
    v=w
  }
}
```

```
Delete(T, k)
{
  v=isElement(T, k)
  if(v==NIL)
    return
  else
    replace v by a InOrder-predecessor/successor
}
```



There are sequences of operations, such that each operation requires  $\Theta(n)$  operations, if  $n$  is the number of nodes in the tree.

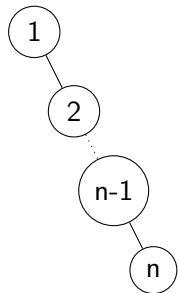
Thus the worst-case complexity of a binary search tree is

$$\Theta(n)$$

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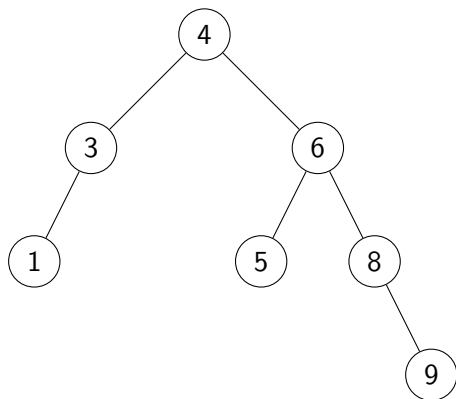


AVL-trees have been invented in 1962 and are internal binary search trees. They are named after their inventors: Georgy Adelson-Velsky and Yevgeniy Landis.

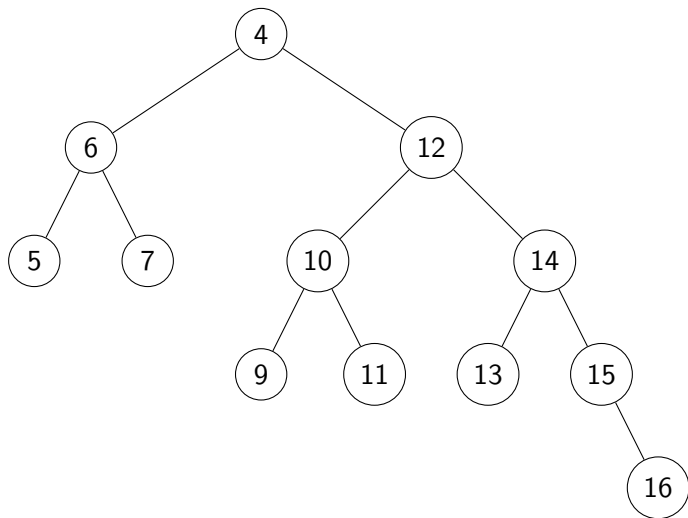
The main idea of AVL-trees is to keep the tree height balanced. This means

$$|\text{height}(\sigma(v.\text{leftchild})) - \text{height}(\sigma(v.\text{rightChild}))| \leq 1$$

has to be valid for every node  $v$  in an AVL-tree.



... is an AVL tree.



... is not an AVL tree.

## Theorem

*An internal binary search tree with height  $h$  contains at most  $2^h - 1$  nodes.*

Proof.

$$\sum_{i=0}^{h-1} 2^i = 2^h - 1$$



## Theorem

An AVL-tree with height  $h$  consists at least of  $F_{h+2} - 1$  internal nodes.

## Proof.

How could an AVL-tree  $T_h$  with height  $h$  and a minimal number of nodes be constructed?

AVL-condition:  $height(\sigma(r.leftchild)) - height(\sigma(r.rightchild)) = 1$

Height should be  $h \Rightarrow$

$height(\sigma(r.leftChild)) = h - 1, height(\sigma(r.rightChild)) = h - 2$

$\Rightarrow n(T_h) = 1 + n(T_{h-1}) + n(T_{h-2})$

$$\begin{array}{llll}
 n(T_1) & = 1 & = 2 - 1 & = F_3 - 1 \\
 n(T_2) & = 2 & = 3 - 1 & = F_4 - 1 \\
 n(T_3) & = 4 & = 5 - 1 & = F_5 - 1 \\
 n(T_h) & = 1 + n(T_{h-1}) + n(T_{h-2}) & = 1 + F_{h+1} - 1 + F_h - 1 & = F_{h+2} - 1
 \end{array}$$

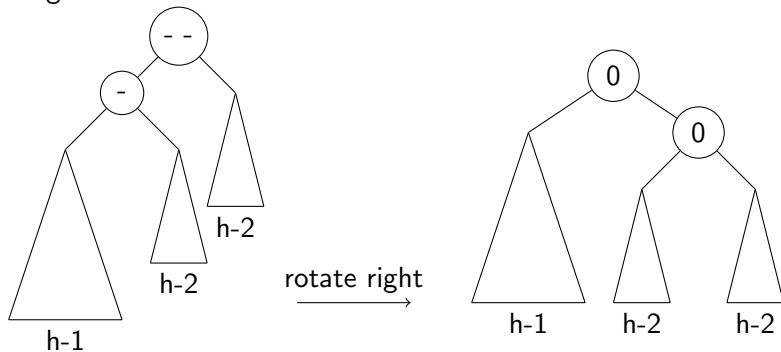
We know:

$$n \geq \frac{1}{\sqrt{5}} \left( \frac{1 + \sqrt{5}}{2} \right)^{h+2}$$

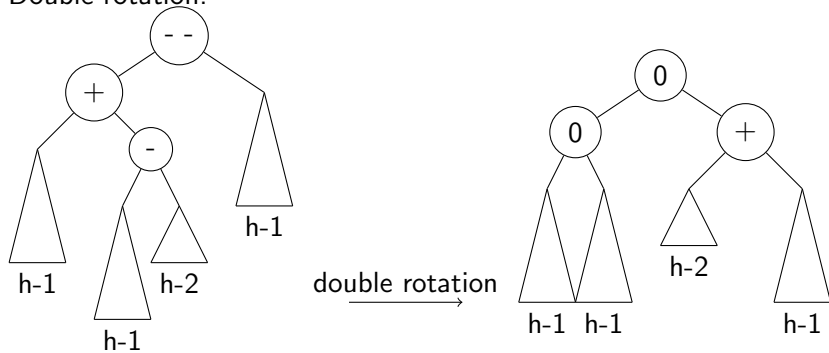
$$\begin{aligned} h &\leq \frac{\log n}{\log\left(\frac{1+\sqrt{5}}{2}\right)} - \log\left(\frac{1}{\sqrt{5}}\right) - 2 \\ &\approx 1.44 \log n + 1.1 \end{aligned}$$



Single rotation:



Double rotation:

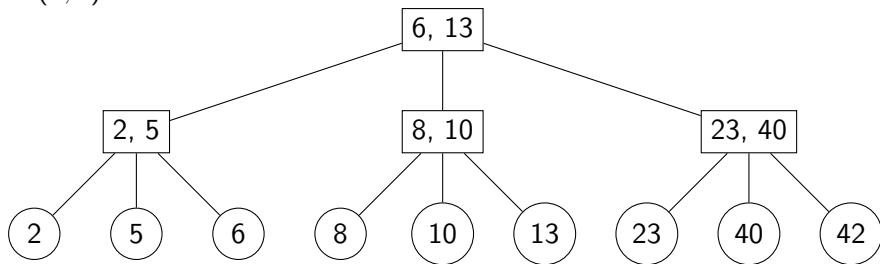


## Definition

An external search tree is an  $(a, b)$ -tree if it applies to the following conditions:

- ▶ All leaves appear on the same level.
- ▶ Every node, except of the root, has  $\geq a$  children.
- ▶ The root has at least two children.
- ▶ Every node has at most  $b$  children.
- ▶ Every node with  $k$  children contains  $k - 1$  keys.
- ▶  $b \geq 2a - 1$

A (2, 4)-tree:



## Theorem

Every  $(a, b)$ -Tree with height  $h$  has

$$2a^{h-1} \leq n \leq b^h$$

leaves.

## Proof.

1. In an  $(a, b)$ -tree which branching factor is as small as possible, the root has two children and every other node has  $a$  children.
2. If we choose the branching factor as high as possible, every node has  $b$  children.



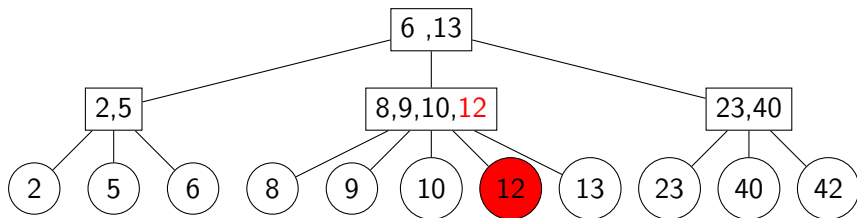
$$\log_b n \leq h \leq \log_a \frac{n}{2} + 1$$

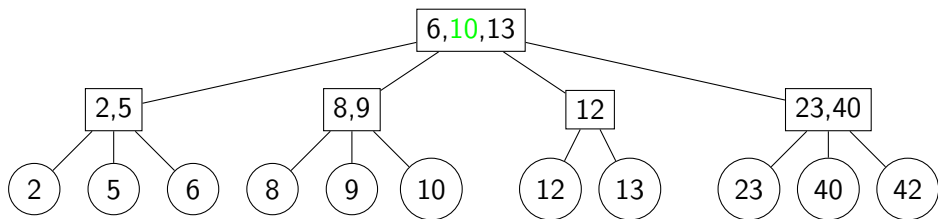
```
IsElement(T,k)
{
  v=T.root
  while(not v.leaf)
  {
    i=min{s; 1 ≤ s ≤ v.children+1 and k ≤ key no. s}
    // define key no. v.children+1 = ∞
    v=child no. i
  }
  return v
}
```

```
Insert(T, k)
{
  w=IsElement(T, k)
  v=parent(w)
  if(w.key!=k)
  {
    if(k< max_key(v) )
      insert k left of w
    else
      insert k right of w
    if( v.children > b )
      rebalance(v)
  }
}
```

```
rebalance(T, l)
{
  w=parent_n(l) // returns an new root , if w==T.root
  r=new node with nodes ( $\lceil \frac{m}{2} \rceil \dots m$ )
  w.add_node( $K_{\frac{m}{2}}$ , r)
  if (w.children > b)
    rebalance(w)
}
```



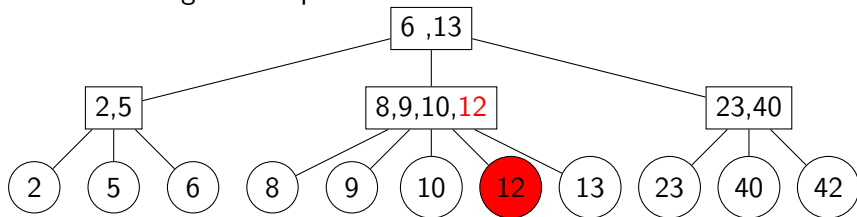


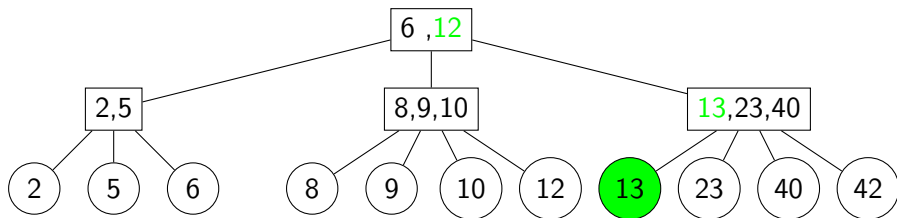


```
Delete(T, k)
{
  w=IsElement(T, k)
  v=parent(w)
  if(k=w.key)
    remove(w)
  if( v.children < a)
    rebalance_delete(T, v)
}
```

```
rebalance_delete(T, v)
{
  w=previous/next_sibling(v)
  r=join(v,w)
  if(r.children >b)
  {
    rebalance_delete(r)
  }
}
```

An alternative way for rebalancing is the idea of overflow.  
Test if a sibling can adopt a child of an overfull node.





## Definition

A  $B^*$ -tree with order  $b$  is defined as follows:

- ▶ All leaves appear on the same level
- ▶ Every node except when the root has at most  $b$  children
- ▶ Every node except when the root has at least  $(2b - 1)/3$  children
- ▶ The root has at least two and at most  $2\lfloor(2m - 2)/3\rfloor + 1$
- ▶ Every internal node with  $k$  children contains  $k - 1$  keys

Splay trees are self-organizing internal binary search trees.

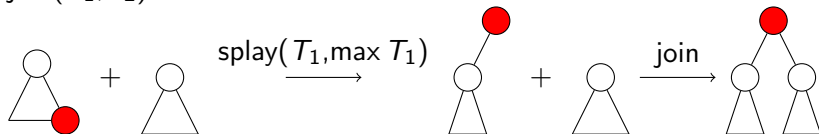
Basic idea: Self-adjusting linear list with the move to front rule.

- ▶ Simple algorithm
- ▶ Good run time in an amortized sense

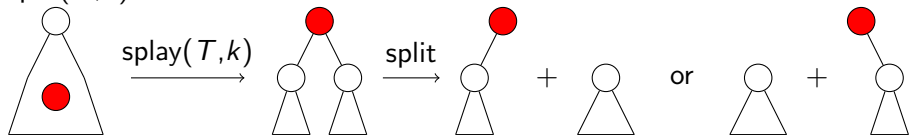
The splay operation moves a node  $x$  with respect to the properties of a search tree to the root of a binary tree  $T$ .



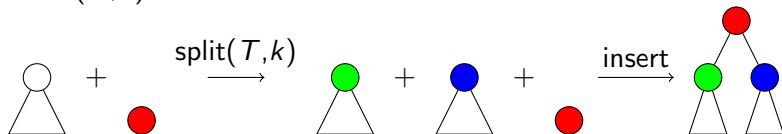
$\text{join}(T_1, T_2)$ :



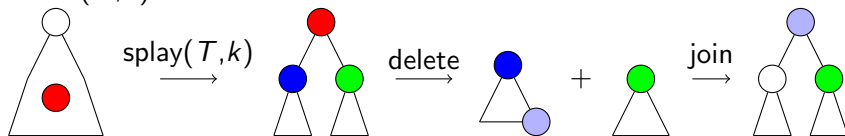
$\text{split}(T, k)$ :



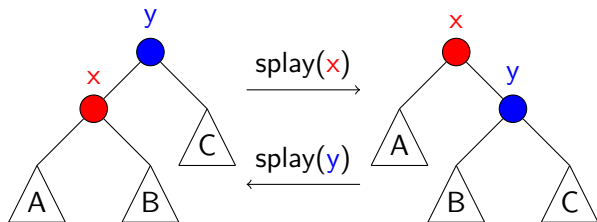
insert( $T, k$ ):

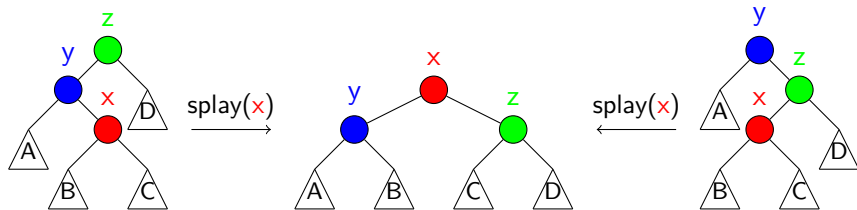
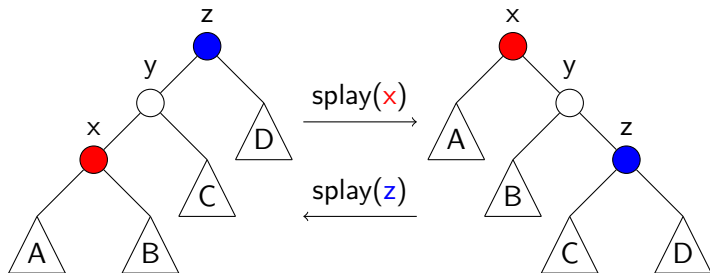


delete( $T, k$ ):

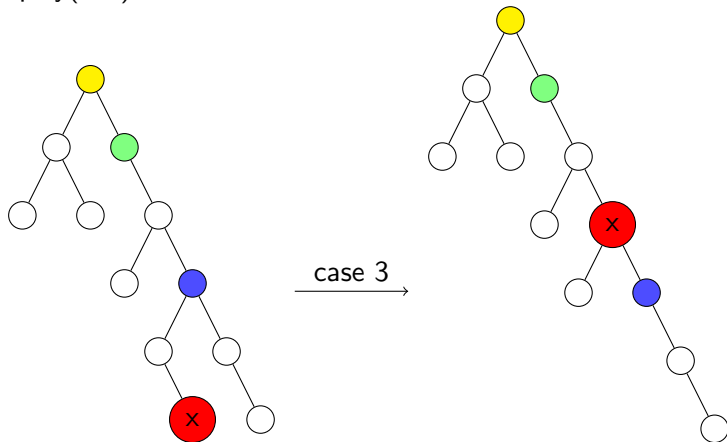


$\text{Splay}(T, x)$  uses single and double rotations for transporting node  $x$  to the root of a splay tree  $T$ .

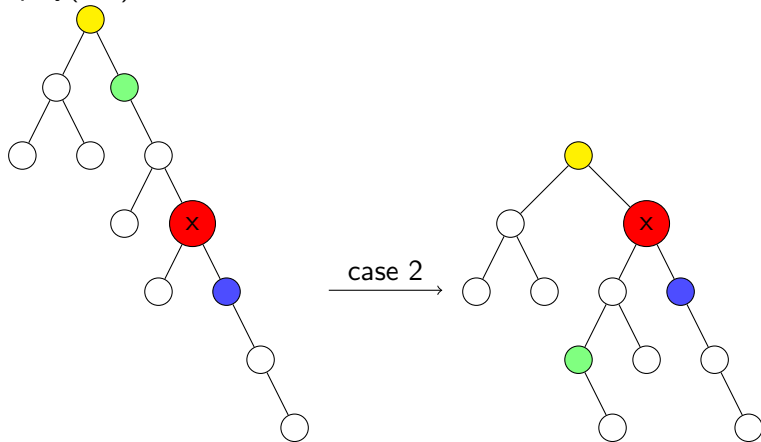




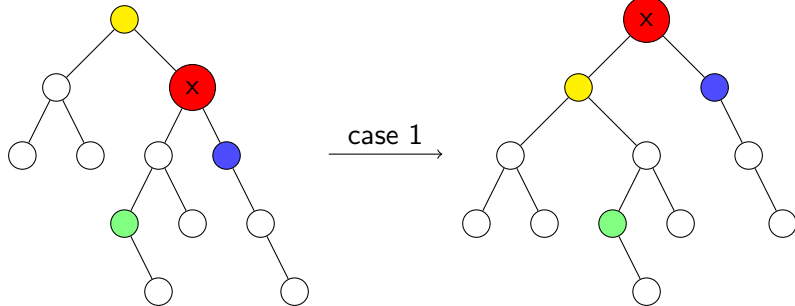
$\text{splay}(T,x)$ :



splay(T,x):



splay(T,x):



In amortized analysis of algorithms we investigate the costs of  $m$  operations.

$$a_i = t_i + \Phi_i - \Phi_{i-1}$$

$$\sum_{i=1}^m t_i = \sum_{i=1}^m (a_i + \Phi_{i-1} - \Phi_i) = \sum_{i=1}^m a_i + \Phi_0 - \Phi_m$$

For the following analysis, we define:

- ▶ A weight  $w(i)$  for each node  $i$
- ▶ The size of node  $x$ :  $s(x) = \sum_{i \in \sigma(x)} w(i)$
- ▶ The rank of node  $x$ :  $r(x) = \log s(x)$
- ▶ The potential of a tree  $T$ :  $\Phi = \sum_{i \in T} r(i)$



## Theorem

*Splay*( $T,x$ ) needs at most

$$3(r(v) - r(x)) + 1 = O\left(\log\left(\frac{s(v)}{s(x)}\right)\right)$$

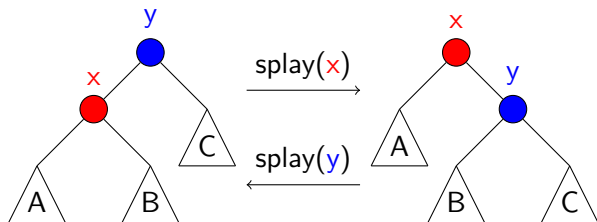
*amortized time*, where  $v$  is the root of  $T$ .

We can divide the splay operation in the rotations which are the influential operations in splay. Thus we consider the number of the rotations.

Just one more notation:

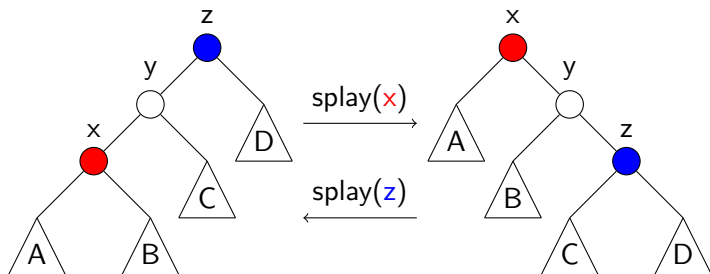
Let  $r(x)$  be the rank of  $x$  before the rotation and  $R(x)$  the rank after the rotation. Let  $s(x)$  be the size of  $x$  before the rotation and  $S(x)$  the size after the rotation.

Case 1:



$$\begin{aligned}
 & 1 + R(x) + R(y) - r(x) - r(y) \\
 \leq & \quad 1 + R(x) - r(x) && \text{since } R(y) \leq r(y) \\
 \leq & \quad 1 + 3(R(x) - r(x)) && \text{since } r(x) \leq R(x)
 \end{aligned}$$

Case 2:



$$\begin{aligned}
 & 2 + R(x) + R(y) + R(z) - r(x) - r(y) - r(z) \\
 = & \quad 2 + R(y) + R(z) - r(x) - r(y) \\
 \leq & \quad 2 + R(x) + R(z) - 2r(x)
 \end{aligned}$$

since  $R(x) = r(z)$ since  $R(y) \leq R(x)$ and  $r(x) \leq r(y)$

Claim:

$$\begin{aligned}2 + R(x) + R(z) - 2r(x) &\leq 3(R(x) - r(x)) \\2 &\leq 2R(x) - r(x) - R(z) \\-2 &\geq \log\left(\frac{s(x)}{S(x)}\right) + \log\left(\frac{S(z)}{S(x)}\right)\end{aligned}$$

$$\begin{aligned}s(x) + S(z) &\leq S(x) \\ \frac{s(x)}{S(x)} + \frac{S(z)}{S(x)} &\leq 1\end{aligned}$$

The log-function is strictly increasing. Thus the maximum of  $f(x, y) = \log x + \log y$  is given by  $x, y$  with  $y = 1 - x$ . For maximization we receive the function  $g(x) = \log_a x + \log_a(1 - x)$ .

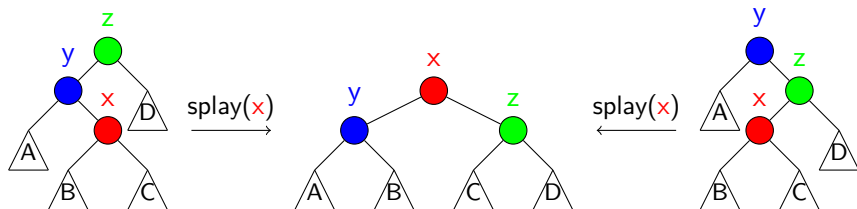
$$g'(x) = \frac{1}{\ln a} \left( \frac{1}{x} - \frac{1}{1-x} \right)$$
$$g''(x) = \frac{1}{\ln a} \left( \frac{1}{x^2} + \frac{1}{(1-x)^2} \right)$$

This leads us to  $x = \frac{1}{2}$ . Since  $g''(\frac{1}{2})$  is negative we can be sure that  $x = \frac{1}{2}$  is a local maximum. Because  $g(\frac{1}{2}) = -2$  equation

$$-2 \geq \log\left(\frac{s(x)}{S(x)}\right) + \log\left(\frac{S(z)}{S(x)}\right)$$

holds.

Case 3:



$$\begin{aligned}
 & 2 + R(x) + R(y) + R(z) - r(x) - r(y) - r(z) \\
 = & \quad 2 + R(y) + R(z) - r(x) - r(y) \\
 \leq & \quad 2 + R(y) + R(z) - 2r(x)
 \end{aligned}$$

since  $R(x) = r(z)$ since  $r(x) - r(y)$

### Proof.

By adding all rotations used for  $splay(T, x)$  we receive a telescope sum, which yields us the amortized time

$$\leq 3(R(x) - r(x)) + 1 = 3(r(t) - r(x)) + 1.$$



If the weights  $w(i)$  are constant,  $-\Phi_m(x)$  for a sequence of  $m$  splay has the upper bound:

$$\sum_{i=1}^n \log W - \log w(i) = \sum_{i=1}^n \frac{W}{w(i)}$$

with

$$W = \sum_{i=1}^n w(i)$$



## Theorem

*The costs of  $m$  access operations in a splay tree are*

$$O((m + n) \log n + m)$$

## Proof.

Choose  $w(i) = \frac{1}{n}$ .

Because  $W = 1$  it follows,  $a_i \leq 1 + 3 \log n$ .

$$-\Phi_m = \sum_{i=1}^n \log \frac{W}{w(i)} = \sum_{i=1}^n \log n = n \log n$$

$$\text{Thus } t = a - \Phi_m = m(1 + 3 \log n) + n \log n$$



# Summary

- ▶ Graph theory
- ▶ Binary search trees
- ▶ AVL-trees
- ▶  $(a, b)$ -trees
- ▶ Splay trees

# End

Thank you for your attention