Algorithms and Data Structures Balanced Trees (AVL-Trees, (a,b)-Trees, Red-Black-Trees)



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Bioinformatics Group / Department of Computer Science Algorithms and Data Structures, January 2019

Structure



Balanced Trees

Motivation

AVL-Trees

(a,b)-Trees

Introduction

Runtime Complexity

Red-Black Trees

Motivation



Binary search tree:

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- Best case: $d \in O(\log n)$, keys are inserted randomly
- Worst case: $d \in O(n)$, keys are inserted in ascending / descending order (20, 19, 18, ...)

Motivation



Gnarley trees:

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■ http://people.ksp.sk/~kuko/bak



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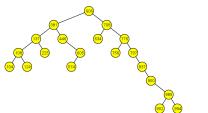


Figure: Binary search tree with random insert [Gna]



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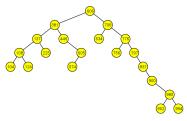
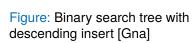


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- We do not want to rely on certain properties of our key set
- We explicitly want a depth of $O(\log n)$
- We rebalance the tree from time to time

Motivation

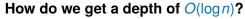


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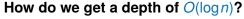


How do we get a depth of $O(\log n)$?

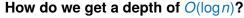
AVL-Tree:



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 - Binary tree with 2 children per node

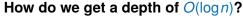


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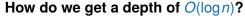


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 - Used in C++ std::map and Java SortedMap



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- Search tree with modified insert and remove operations while satisfying a depth condition
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- Height difference of left and right subtree is at maximum one
- With that the height of the search tree is always $O(\log n)$
- We can perform all basic operations in $O(\log n)$

Balanced Trees AVL-Tree



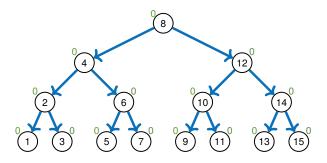


Figure: Example of an AVL-Tree

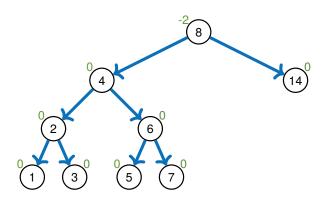


Figure: Not an AVL-Tree

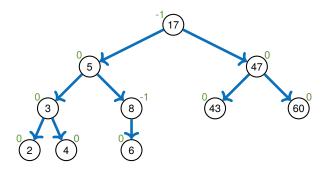
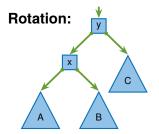
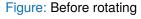


Figure: Another example of an AVL-Tree





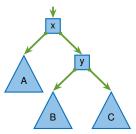
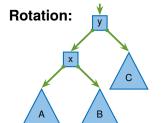
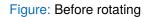


Figure: After rotating





A B C

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■ Central operation of rebalancing

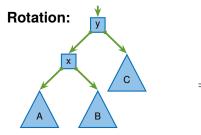


Figure: Before rotating

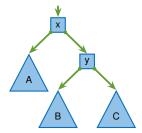
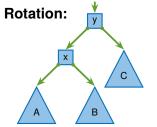


Figure: After rotating

- Central operation of rebalancing
- After rotation to the right:





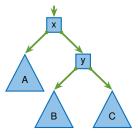
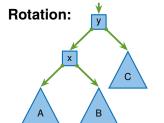


Figure: After rotating

- Central operation of rebalancing
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 - lacksquare Subtree A is a layer higher and subtree C a layer lower



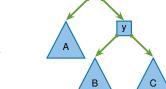


Figure: Before rotating

Figure: After rotating

- Central operation of rebalancing
- After rotation to the right:
 - Subtree *A* is a layer higher and subtree *C* a layer lower
 - The parent child relations between nodes *x* and *y* have been swapped

Balanced Trees

AVL-Tree - Rebalancing





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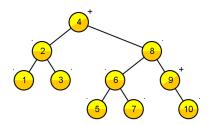


Figure: Inserting 1,...,10 into an AVL-tree [Gna]

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- However not amortized update costs of O(1)
- Additional memory costs: We have to save a height difference for every node
- Better (and easier) to implement are (a,b)-trees



Balanced Trees

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(a,b)-Trees

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Red-Black Trees

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Idea:

- Save a varying number of elements per node
- So we have space for elements on an insert and balance operation

(a,b)-Trees Introduction



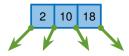


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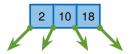
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- Subtrees are located "between" the elements
- We require: $a \ge 2$ and $b \ge 2a 1$

(2,4)-Tree:

Introduction

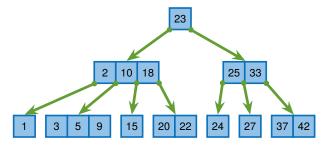


Figure: Example of an (2,4)-tree

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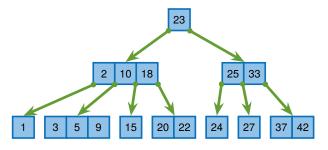


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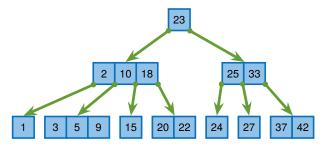


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- (2,4)-tree with depth of 3
- Each node has between 2 and 4 children (1 to 3 elements)

Introduction

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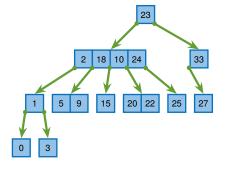


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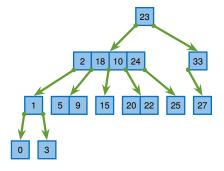


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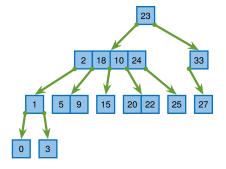


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- Invalid sorting
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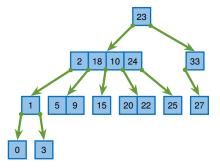


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- Invalid sorting
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- Leaves on different levels



NE NE

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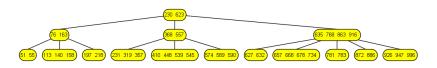


Figure: (3,5)-Tree [Gna]



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- Then we **split** the node



Figure: Splitting a node



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- This results in a node with $\operatorname{ceil}\left(\frac{b-1}{2}\right)$ elements, a node with $\operatorname{floor}\left(\frac{b-1}{2}\right)$ elements and one element for the parent node



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- This results in a node with $\operatorname{ceil}\left(\frac{b-1}{2}\right)$ elements, a node with $\operatorname{floor}\left(\frac{b-1}{2}\right)$ elements and one element for the parent node
- Thats why we have the limit $b \ge 2a 1$



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- If we split the root node we create a new parent root node (The tree is now one level deeper)



Search the element in $O(\log n)$ time



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 - Remove element



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Figure: Borrow an element



NE NE

Implementation - Remove

Removing an element: (remove)

Attention: The leaf might be too small (degree of a-1) ⇒ We rebalance the tree

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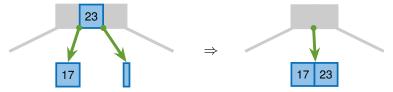


Figure: Merge two nodes



■ Now the parent node can be of degree a – 1

- Now the parent node can be of degree a-1
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- Now the parent node can be of degree a-1
- We merge parent nodes the same way
- If the root has only a single child
 - Remove the root
 - Define sole child as new root
 - The tree shrinks by one level

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- \blacksquare All operations in O(d) with d being the depth of the tree
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In detail:

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- \blacksquare lookup always takes $\Theta(d)$
- insert and remove often require only O(1) time
- Worst case: split or merge all nodes on path up to the root
- Therefore instead of b > 2a 1 we need b > 2a

(a,b)-Trees

Runtime Complexity - Counter-example for (2,3)-Tree



Counter example (2,3)-Tree:

■ Before executing delete(11)

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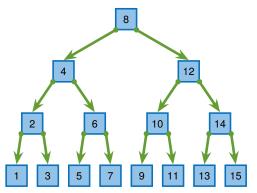


Figure: Normal (2,3)-Tree

■ Executing delete(11)

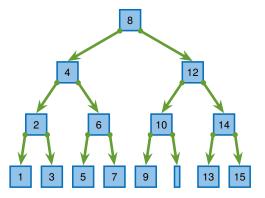


Figure: (2,3)-Tree - Delete step 1

■ Executing delete(11)

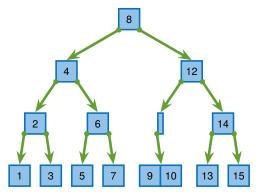


Figure: (2,3)-Tree - Delete step 2

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Counter example (2,3)-Tree:

■ Executing delete(11)

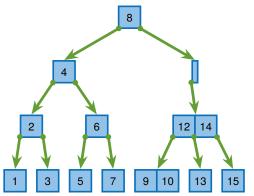


Figure: (2,3)-Tree - Delete step 3

■ Executed delete(11)

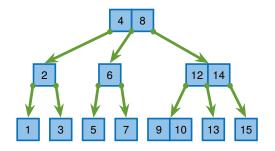


Figure: (2,3)-Tree - Delete step 4

Runtime Complexity - Counter example for (2,3)-Tree



Counter example (2,3)-Tree:



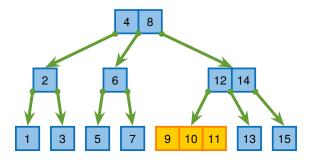


Figure: (2,3)-Tree - Insert step 1

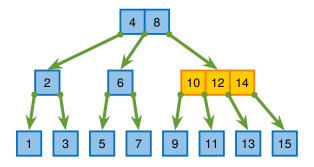


Figure: (2,3)-Tree - Insert step 2

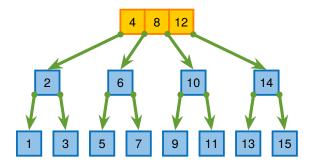


Figure: (2,3)-Tree - Insert step 3

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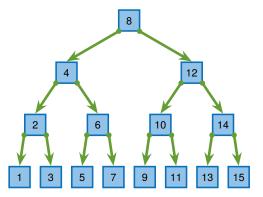


Figure: (2,3)-Tree - Insert step 4

We are exactly where we started

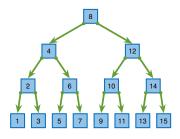


Figure: (2,3)-Tree

- We are exactly where we started
- If b = 2a 1 then we can create a sequence of insert and remove operations where each operation costs O(log n)

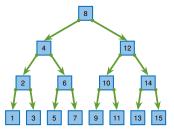


Figure: (2,3)-Tree

- We are exactly where we started
- If b = 2a 1 then we can create a sequence of insert and remove operations where each operation costs O(log n)
- We need $b \ge 2a$ instead of b > 2a 1

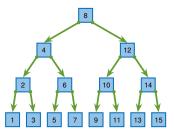


Figure: (2,3)-Tree

(2,4)-Tree:



■ If all nodes have 2 children we have to merge the nodes up to the root on a remove operation

Runtime Complexity - (2,4)-Tree

(2,4)-Tree:

- If all nodes have 2 children we have to merge the nodes up to the root on a remove operation
- If all nodes have 4 children we have to split the nodes up to the root on a insert operation

Runtime Complexity - (2,4)-Tree



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Runtime Complexity - (2,4)-Tree

(2,4)-Tree:

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- If all nodes have 4 children we have to split the nodes up to the root on a insert operation
- If all nodes have 3 children it takes some time to reach one of the previous two states
- ⇒ Nodes of degree 3 are stable Neither an insert nor a remove operation trigger rebalancing operations

■ Idea:

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Runtime Complexity - (2,4)-Tree

- Idea:
 - After an expensive operation the tree is in a stable state
 - It takes some time until the next expensive operation occurs
- Like with dynamic arrays:
 - Reallocation is expensive but it takes some time until the next expensive operation occurs
 - If we overallocate clever we have an amortized runtime of O(1)

■ We analyze a sequence of *n* operations

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- Empty tree has 0 nodes: $\Phi = 0$

Example:



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■ Nodes of degree 3 are highlighted



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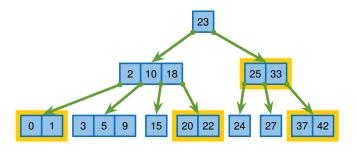


Figure: Tree with potential $\Phi = 4$

Runtime Complexity - (2,4)-Tree



Terminology:

■ Let c_i be the costs = runtime of the i-th operation

Runtime Complexity - (2,4)-Tree



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$$c_i \le A \cdot (\Phi_i - \Phi_{i-1}) + B, \quad A > 0 \text{ and } B > A$$

Number of gained stable nodes (degree 3) ≥ -1

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Number of gained stable nodes (degree 3) ≥ -1

■ Each operation has an amortitzed cost of O(1) summing up to O(n) in total





Figure: Splitting a node on insert



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■ Each splitted node creates a node of degree 3



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- Each splitted node creates a node of degree 3
- The parent node receives an element from the splitted node



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- Each splitted node creates a node of degree 3
- The parent node receives an element from the splitted node
- If the parent node is also full we have to split it too

Runtime Complexity - (2,4)-Tree



Runtime Complexity - (2,4)-Tree



Case 1: *i-th* operation is an insert operation on a full node

■ Let *m* be the number of nodes split

Runtime Complexity - (2,4)-Tree



- \blacksquare Let m be the number of nodes split
- The potential rises by m



- Let *m* be the number of nodes split
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$$\Rightarrow m < \Phi_i - \Phi_{i-1} + 1$$

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- The potential rises by m
- If the "stop-node" is of degree 3 then the potential goes down by one

$$\Phi_i \ge \Phi_{i-1} + m - 1$$

$$\Rightarrow m \le \Phi_i - \Phi_{i-1} + 1$$

Costs: $c_i \leq A \cdot m + B$

$$\Rightarrow c_i \leq A \cdot (\Phi_i - \Phi_{i-1} + 1) + B$$
$$c_i \leq A \cdot (\Phi_i - \Phi_{i-1}) + \underbrace{A + B}_{B'}$$





Case 2: *i-th* operation is an remove operation

Runtime Complexity - (2,4)-Tree



Case 2: *i-th* operation is an remove operation

■ Case 2.1: Inner node



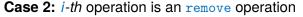
Runtime Complexity - (2,4)-Tree



Case 2: *i-th* operation is an remove operation

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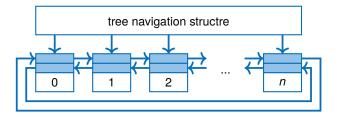


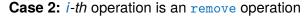
Figure: Tree with doubly linked list



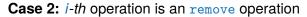
Case 2.1: Borrow a node



- Case 2: *i-th* operation is an remove operation
 - Case 2.1: Borrow a node
 - Creates no additional operations



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 - Case 2.1.1: Potential rises by one



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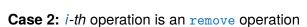


Figure: Case 2.1.1: Borrow an element

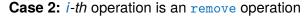




■ Case 2.1: Borrow a node



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Figure: Case 2.1.2: Borrow an element

(a,b)-Trees Runtime Complexity - (2,4)-Tree



Case 2: *i-th* operation is an remove operation





Figure: Merging two nodes

Potential rises by one



Figure: Merging two nodes

- Potential rises by one
- Parent node has one element less after the operation

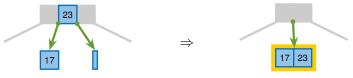


Figure: Merging two nodes

- Potential rises by one
- Parent node has one element less after the operation
- This operation propagates upwards until a node of degree
 2 or a node of degree 2, which can borrow from a neighbour

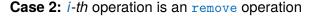




Figure: Merging two nodes



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■ The potential rises by m



Figure: Merging two nodes

- The potential rises by m
- If the "stop-node" is of degree 2 then the potential eventually goes down by one

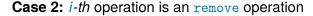




Figure: Merging two nodes

- The potential rises by m
- If the "stop-node" is of degree 2 then the potential eventually goes down by one
- Same costs as insert

Lemma:

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We know:

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$$c_i \le A \cdot (\Phi_i - \Phi_{i-1}) + B$$
, $A > 0$ and $B > A$

■ With that we can conclude:

$$\sum_{i=0}^n c_i \in O(n)$$

Proof:

$$\sum_{i=0}^{n} c_{i} \leq \underbrace{A \cdot (\Phi_{1} - \Phi_{0}) + B}_{\leq c_{1}} + \underbrace{A \cdot (\Phi_{2} - \Phi_{1}) + B}_{\leq c_{2}} + \cdots + \underbrace{A \cdot (\Phi_{n} - \Phi_{n-1}) + B}_{\leq c_{n}}$$

$$= A \cdot (\Phi_{n} - \Phi_{0}) + B \cdot n \qquad | \text{ telescope sum}$$

$$= A \cdot \Phi_{n} + B \cdot n \qquad | \text{ we start with an empty tree}$$

$$< A \cdot n + B \cdot n \in O(n) \qquad | \text{ number of degree 3 nodes}$$

$$< \text{ number of nodes}$$



Balanced Trees

Motivation
AVL-Trees
(a,b)-Trees
Introduction
Buntime Complexity

Red-Black-Trees

Introduction



Red-Black-Trees

Introduction

Red-Black Tree:

■ Binary tree with red and black nodes

Introduction

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- Can be interpreted as (2,4)-tree (also named 2-3-4-tree)
- Each (2,4)-tree-node is a small red-black-tree with a black root node

Red-Black-Trees

Introduction



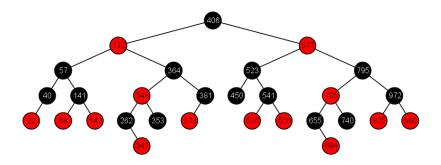


Figure: Example of an red-black-tree [Gna]

General

- [CRL01] Thomas H. Cormen, Ronald L. Rivest, and Charles E. Leiserson. Introduction to Algorithms. MIT Press, Cambridge, Mass, 2001.
- [MS08] Kurt Mehlhorn and Peter Sanders. Algorithms and data structures, 2008. https://people.mpi-inf.mpg.de/~mehlhorn/ftp/Mehlhorn-Sanders-Toolbox.pdf.

■ Gnarley Trees

[Gna] Gnarley Trees

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AVL-Tree

[Wik] AVL tree https://en.wikipedia.org/wiki/AVL_tree

■ (a,b)-Tree

[Wika] 2-3-4 tree

https://en.wikipedia.org/wiki/2%E2%80%933% E2%80%934_tree

[Wikb] (a,b)-tree

https://en.wikipedia.org/wiki/(a,b)-tree

[Wik] Red-black tree

https://en.wikipedia.org/wiki/Red%E2%80%93black_tree