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



Prof. Dr. Rolf Backofen

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

Binary search tree:

- With BinarySearchTree we could perform an lookup or insert in O(d), with d being the depth of the tree
- 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, ...)

Gnarley trees:



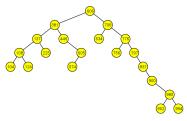
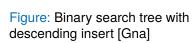


Figure: Binary search tree with random insert [Gna]





Balanced Trees

Motivation



Balanced trees:

- 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

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

- AVL-Tree:
 - Binary tree with 2 children per node
 - Balancing via "rotation"
- (a,b)-Tree or B-Tree:
 - Node has between a and b children
 - Balancing through splitting and merging nodes
 - Used in databases and file systems
- Red-Black-Tree:
 - Binary tree with "black" and "red" nodes
 - Balancing through "rotation" and "recoloring"
 - Can be interpreted as (2, 4)-tree
 - Used in C++ std::map and Java SortedMap

AVL-Tree:

- Gregory Maximovich Adelson-Velskii, Yevgeniy Mikhailovlovich Landis (1963)
- Search tree with modified insert and remove operations while satisfying a depth condition
- Prevents degeneration of the search tree
- 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



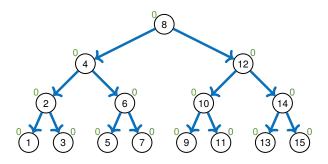


Figure: Example of an AVL-Tree

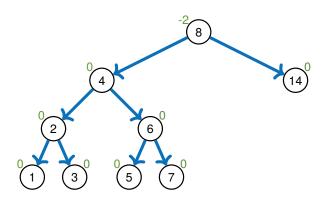


Figure: Not an AVL-Tree

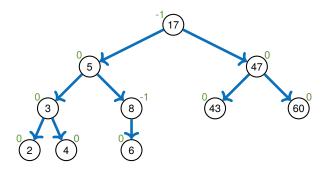
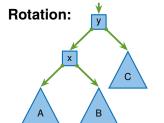


Figure: Another example of an AVL-Tree



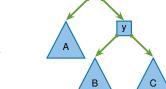


Figure: Before rotating

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

AVL-Tree:

- If a height difference of ± 2 occurs on an insert or remove operation the tree is rebalanced
- Many different cases of rebalancing
- **Example:** insert of 1,2,3,...

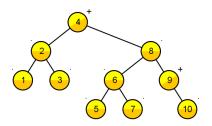


Figure: Inserting 1,...,10 into an AVL-tree [Gna]

Summary:

- Historical the first search tree providing guaranteed insert, remove and lookup in O(log n)
- 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

(a,b)-Tree:

- Also known as b-tree (b for "balanced")
- Used in databases and file systems

Idea:

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

(a,b)-Tree:

Introduction

- All leaves have the same depth
- Each inner node has $\geq a$ and $\leq b$ nodes (Only the root node may have less nodes)



- Each node with n children is called "node of degree n" and holds n - 1 sorted elements
- Subtrees are located "between" the elements
- We require: $a \ge 2$ and $b \ge 2a 1$

(2,4)-Tree:

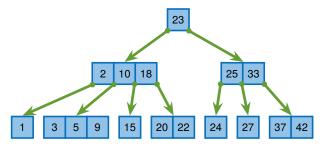


Figure: Example of an (2,4)-tree

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

Not an (2,4)-Tree:

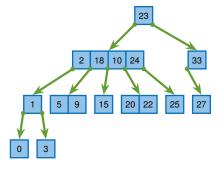


Figure: Not an (2,4)-tree

- Invalid sorting
- Degree of node too large / too small
- Leaves on different levels

Searching an element: (lookup)

- The same algorithm as in BinarySearchTree
- Searching from the root downwards
- The keys at each node set the path

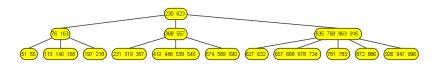


Figure: (3,5)-Tree [Gna]

- Search the position to insert the key into
- This position will always be an leaf
- Insert the element into the tree
- Attention: As a result node can overflow by one element (Degree b + 1)
- Then we **split** the node

Inserting an element: (insert)



Figure: Splitting a node

- If the degree is higher than b+1 we split the node
- 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 $b \ge 2a 1$

Inserting an element: (insert)

- If the degree is higher than b+1 we split the node
- Now the parent node can be of a higher degree than b + 1
- We split the parent nodes the same way
- 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
- Case 1: The element is contained by a leaf
 - Remove element
- Case 2: The element is contained by an inner node
 - Search the successor in the right subtree
 - The successor is always contained by a leaf
 - Replace the element with its successor and delete the successor from the leaf
- **Attention:** The leaf might be too small (degree of a-1)
 - ⇒ We rebalance the tree

- **Attention:** The leaf might be too small (degree of a-1) \Rightarrow We rebalance the tree
 - Case a: If the left or right neighbour node has a degree greater than a we borrow one element from this node



Figure: Borrow an element

- **Attention:** The leaf might be too small (degree of a-1) \Rightarrow We rebalance the tree
 - Case b: We merge the node with its right or left neighbour

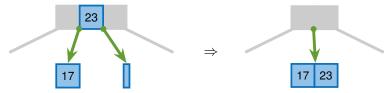


Figure: Merge two nodes

- 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

Runtime complexity of lookup, insert and remove:

- \blacksquare All operations in O(d) with d being the depth of the tree
- Each node (except the root) has more than a children $\Rightarrow n > a^{d-1}$ and $d < 1 + \log_a n = O(\log_a n)$

In detail:

- \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

N FIRING

Counter example (2,3)-Tree:

■ Before executing delete(11)

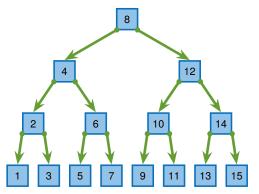


Figure: Normal (2,3)-Tree

■ Executing delete(11)

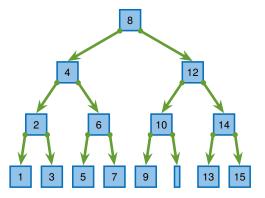


Figure: (2,3)-Tree - Delete step 1

■ Executing delete(11)

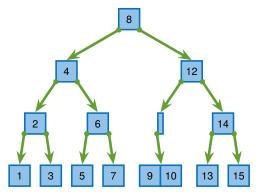


Figure: (2,3)-Tree - Delete step 2

NIREIBURG

Counter example (2,3)-Tree:

■ Executing delete(11)

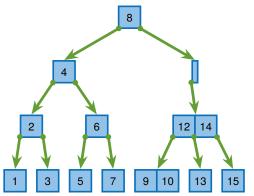


Figure: (2,3)-Tree - Delete step 3

■ Executed delete(11)

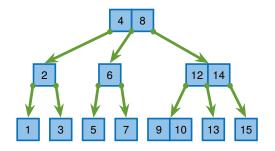


Figure: (2,3)-Tree - Delete step 4

■ Executing insert(11)

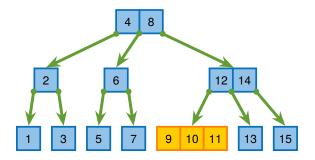


Figure: (2,3)-Tree - Insert step 1

■ Executing insert(11)

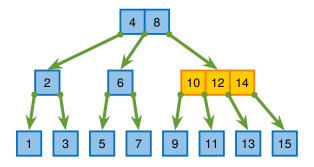


Figure: (2,3)-Tree - Insert step 2

■ Executing insert(11)

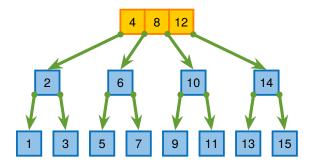


Figure: (2,3)-Tree - Insert step 3

■ Executed insert(11)

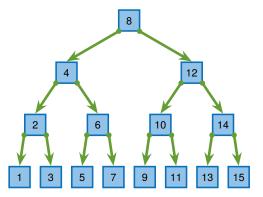


Figure: (2,3)-Tree - Insert step 4

- 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

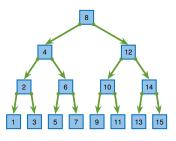


Figure: (2,3)-Tree

Runtime Complexity - (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
- 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

Runtime Complexity - (2,4)-Tree

(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)

Terminology:

- We analyze a sequence of n operations
- Let Φ_i be the potential of the tree after the *i-th* operation
- \blacksquare Φ_i = the number of stable nodes with degree 3
- Empty tree has 0 nodes: $\Phi = 0$

Example:

Nodes of degree 3 are highlighted

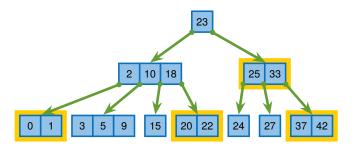


Figure: Tree with potential $\Phi = 4$

- Let c_i be the costs = runtime of the i-th operation
- We will show:
 - Each operation can at most destroy one stable node
 - For each cost incurring step the operation creates an additional stable node
- The costs for operation i are coupled to the difference of the potential levels

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

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

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

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



Figure: Splitting a node on insert

- 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

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

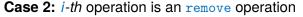
- Let *m* be the number of nodes split
- The potential rises by m
- If the "stop-node" is of degree 3 then the potential goes down by one

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

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

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 2.1: Inner node
 - Searching the successor in a tree is $O(d) = O(\log n)$
 - Normally the tree is coupled with a doubly linked list
 - \Rightarrow We can find the successor in O(1)

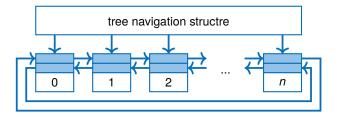
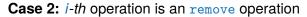


Figure: Tree with doubly linked list



- Case 2.1: Borrow a node
 - Creates no additional operations
 - Case 2.1.1: Potential rises by one



Figure: Case 2.1.1: Borrow an element

- Case 2.1: Borrow a node
 - Creates no additional operations
 - Case 2.1.2: Potential is lowered by one



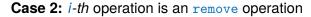
Figure: Case 2.1.2: Borrow an element

■ Case 2.2: Merging two node



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



Case 2.2: Merging two node



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:

We know:

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

■ With that we can conclude:

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

Proof:

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

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

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

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

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

Red-Black Tree:

- Binary tree with red and black nodes
- Number of black nodes on path to leaves is equal
- 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



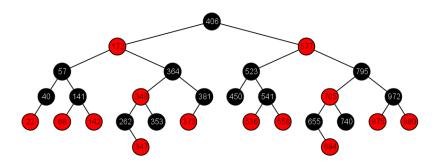


Figure: Example of an red-black-tree [Gna]

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

https://people.ksp.sk/~kuko/gnarley-trees/

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

■ Red-Black-Tree

[Wik] Red-black tree

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