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



Albert-Ludwigs-Universität Freiburg

Prof. Dr. Rolf Backofen

Bioinformatics Group / Department of Computer Science Algorithms and Datastructures, January 2017

Structure



Balanced Trees

Motivation

AVL-Trees

(a,b)-Trees

Introduction

Runtime Complexity

Red-Black Trees

Motivation



Binary search tree:

Motivation



Binary search tree:

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

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

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,...)

Motivation



Gnarley trees:

Gnarley trees:

■ http://people.ksp.sk/~kuko/bak





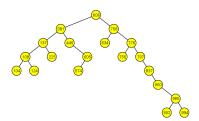


Figure: Binary search tree with random insert [Gna]



Gnarley trees:



■ http://people.ksp.sk/~kuko/bak

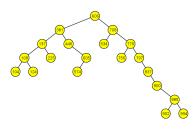


Figure: Binary search tree with random insert [Gna]

Figure: Binary search tree with descending insert [Gna]

Motivation



Balanced trees:

Motivation



Balanced trees:

■ We do not want to rely on certain properties of our key set

Motivation



Balanced trees:

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

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

Motivation

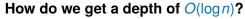


Motivation

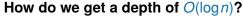


How do we get a depth of $O(\log n)$?

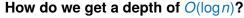




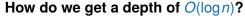
- AVL-Tree:
 - Binary tree with 2 children per node



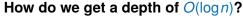
- AVL-Tree:
 - Binary tree with 2 children per node
 - Balancing via "rotation"



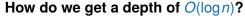
- AVL-Tree:
 - Binary tree with 2 children per node
 - Balancing via "rotation"
- (a,b)-Tree or B-Tree:



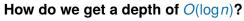
- AVL-Tree:
 - Binary tree with 2 children per node
 - Balancing via "rotation"
- (a,b)-Tree or B-Tree:
 - Node has between *a* and *b* children



- 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



- 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



- 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:

- 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

- 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"

- 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

- 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



Motivation

AVL-Trees

(a,b)-Trees
Introduction
Runtime Complexity

Red-Black Trees

Balanced Trees AVL-Tree



AVL-Tree



AVL-Tree:

Gregory Maximovich Adelson-Velskii, Yevgeniy Mikhailovlovich Landis (1963)

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

- 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

- 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

- 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 $O(\log n)$

- 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



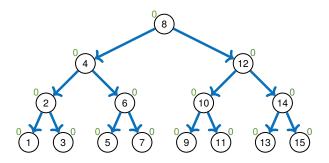


Figure: Example of an AVL-Tree

Balanced Trees AVL-Tree



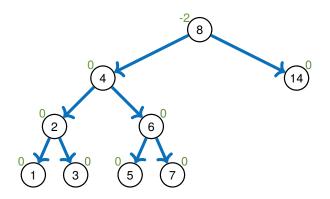


Figure: Not an AVL-Tree

Balanced Trees **AVL-Tree**



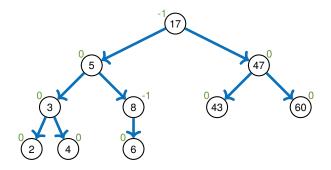


Figure: Another example of an AVL-Tree



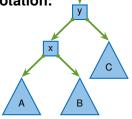


Figure: Before rotating

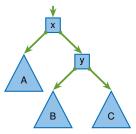


Figure: After rotating

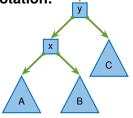


Figure: Before rotating

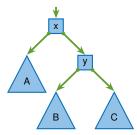


Figure: After rotating

■ Central operation of rebalancing

Rotation:

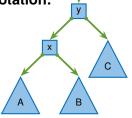


Figure: Before rotating

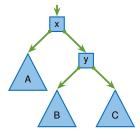


Figure: After rotating

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



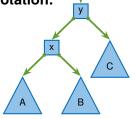


Figure: Before rotating

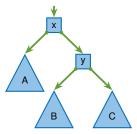


Figure: After rotating

- Central operation of rebalancing
- After rotation to the right:
 - Subtree *A* is a layer higher and subtree *C* a layer lower



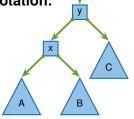


Figure: Before rotating

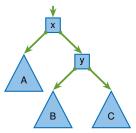


Figure: After rotating

- 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





Balanced Trees

AVL-Tree - Rebalancing



AVL-Tree:

■ If a height difference of ± 2 occurs on an insert or remove operation the tree is rebalanced

- If a height difference of ± 2 occurs on an insert or remove operation the tree is rebalanced
- Many different cases of rebalancing

- 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,...

- 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,...

- 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,...

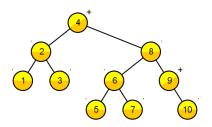


Figure: Inserting 1,...,10 into an AVL-tree [Gna]

Balanced Trees

AVL-Tree - Summary



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

- Historical the first search tree providing guaranteed insert, remove and lookup in O(log n)
- However not amortized update costs of O(1)

- 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

- 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

Structure



Balanced Trees

Motivation AVL-Trees

(a,b)-Trees

Introduction
Runtime Complexity

Red-Black Trees



■ Also known as **b-tree** (b for "balanced")

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



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

Idea:

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

Idea:

Save a varying number of elements per node

- 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)-Trees Introduction





(a,b)-Tree:

All leaves have the same depth

(a,b)-Trees Introduction





(a,b)-Tree:

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

Introduction



(a,b)-Tree:

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





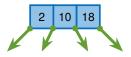
- All leaves have the same depth
- Each inner node has $\geq a$ and $\leq b$ nodes (Only the root node may have less nodes)



■ Each node with n children is called "node of degree n" and holds n-1 sorted elements

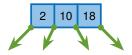
Introduction

- 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

- 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:

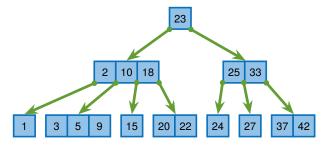


Figure: Example of an (2,4)-tree

(2,4)-Tree:

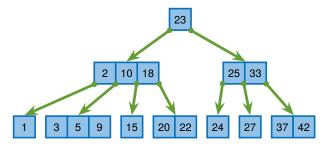


Figure: Example of an (2,4)-tree

■ (2,4)-tree with depth of 3

(2,4)-Tree:

Introduction

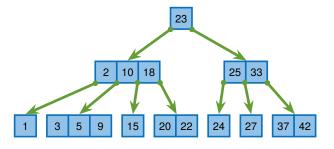


Figure: Example of an (2,4)-tree

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

Not an (2,4)-Tree:

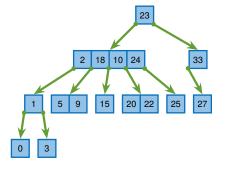


Figure: Not an (2,4)-tree

Introduction

Not an (2,4)-Tree:

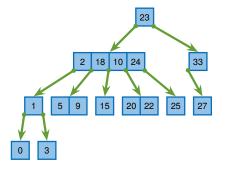


Figure: Not an (2,4)-tree

Invalid sorting

Introduction

Not an (2,4)-Tree:

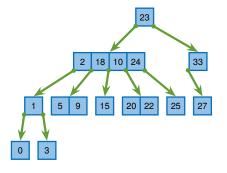


Figure: Not an (2,4)-tree

- Invalid sorting
- Degree of node too large / too small

Introduction

Not an (2,4)-Tree:

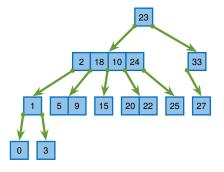


Figure: **Not** an (2,4)-tree

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



Searching an element: (lookup)



Searching an element: (lookup)

The same algorithm as in BinarySearchTree



- The same algorithm as in BinarySearchTree
- Searching from the root downwards



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

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

- Search the position to insert the key into
- This position will always be an leaf

- Search the position to insert the key into
- This position will always be an leaf
- Insert the element into the tree

- 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)

- 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



Figure: Splitting a node



Figure: Splitting a node

■ If the degree is higher than b+1 we split the node



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



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 $b \ge 2a 1$



■ If the degree is higher than b + 1 we split the node

- 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



- 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 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



- Search the element in $O(\log n)$ time
- Case 1: The element is contained by a leaf
 - Remove element

- \blacksquare 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

- \blacksquare 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



- \blacksquare 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

- \blacksquare 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

- \blacksquare 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



Implementation - Remove



- **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

- **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

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

- **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

- Now the parent node can be of degree a-1
- We merge parent nodes the same way

- 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



 \blacksquare All operations in O(d) with d being the depth of the tree

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

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

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

In detail:

■ lookup always takes $\Theta(d)$

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

- lookup always takes $\Theta(d)$
- \blacksquare insert and remove often require only O(1) time

- \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 $d < 1 + \log_a n = O(\log_a n)$

- 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

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

- lookup always takes $\Theta(d)$
- \blacksquare 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 $b \ge 2a 1$ we need $b \ge 2a$

(a,b)-Trees

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

Counter example (2,3)-Tree:



(a,b)-Trees

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



Counter example (2,3)-Tree:

Before executing delete(11)



VI EIBURG

Counter example (2,3)-Tree:

■ Before executing delete(11)

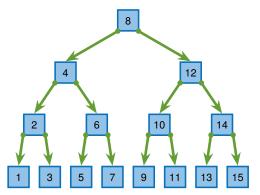


Figure: Normal (2,3)-Tree

NI

Counter example (2,3)-Tree:

■ Executing delete(11)

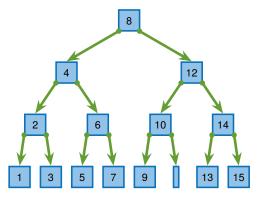


Figure: (2,3)-Tree - Delete step 1

VI

Counter example (2,3)-Tree:

■ Executing delete(11)

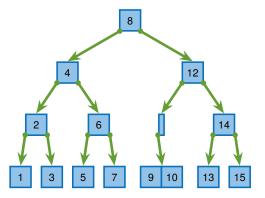


Figure: (2,3)-Tree - Delete step 2

NI

Counter example (2,3)-Tree:

■ Executing delete(11)

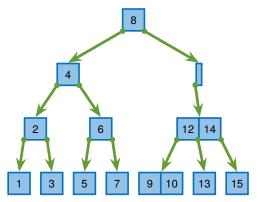


Figure: (2,3)-Tree - Delete step 3

■ Executed delete(11)

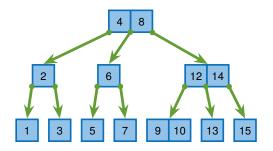


Figure: (2,3)-Tree - Delete step 4

(a,b)-Trees

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



Counter example (2,3)-Tree:

(a,b)-Trees

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



Counter example (2,3)-Tree:

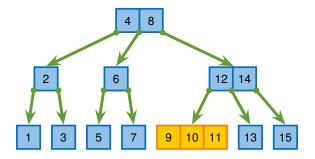


Figure: (2,3)-Tree - Insert step 1

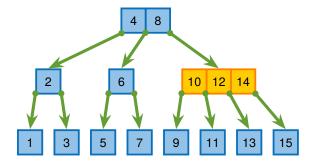


Figure: (2,3)-Tree - Insert step 2

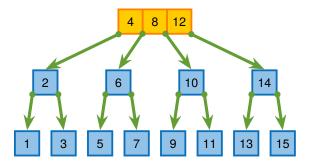


Figure: (2,3)-Tree - Insert step 3

NI

Counter example (2,3)-Tree:

■ Executed insert(11)

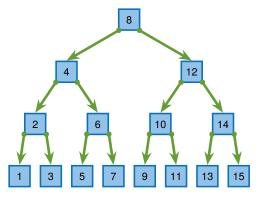


Figure: (2,3)-Tree - Insert step 4

We are exactly where we started

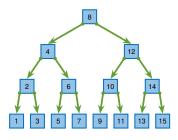


Figure: (2,3)-Tree

- 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)

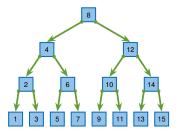


Figure: (2,3)-Tree

- 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

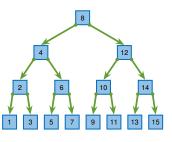


Figure: (2,3)-Tree



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

- 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

- 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

- 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



■ Idea:

- Idea:
 - After an expensive operation the tree is in a stable state

- Idea:
 - After an expensive operation the tree is in a stable state
 - It takes some time until the next expensive operation occurs

- 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:

- 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

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

- We analyze a sequence of n operations
- Let Φ_i be the potential of the tree after the *i-th* operation

- 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

- \blacksquare 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:



Example:

■ Nodes of degree 3 are highlighted



Example:

■ Nodes of degree 3 are highlighted

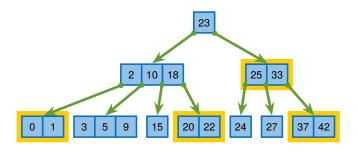


Figure: Tree with potential $\Phi = 4$

Runtime Complexity - (2,4)-Tree



Terminology:

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

Runtime Complexity - (2,4)-Tree



- Let c_i be the costs = runtime of the *i*-th operation
- We will show:

Runtime Complexity - (2,4)-Tree



- Let c_i be the costs = runtime of the i-th operation
- We will show:
 - Each operation can at most destroy one stable node

- 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

- 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

- 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

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





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

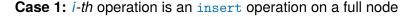




Figure: Splitting a node on insert



Figure: Splitting a node on insert

Each splitted node creates a node of degree 3



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

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

Runtime Complexity - (2,4)-Tree

BURG

FREIB

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

Runtime Complexity - (2,4)-Tree



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

 \blacksquare Let m be the number of nodes split

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
- The potential rises by m



- 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



- 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

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

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

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

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

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



FREE

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
 - Case 2.1: Inner node
 - Searching the successor in a tree is $O(d) = O(\log n)$

Runtime Complexity - (2,4)-Tree



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

- 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)

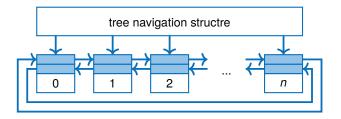
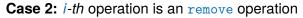


Figure: Tree with doubly linked list







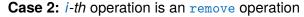


■ Case 2.1: Borrow a node

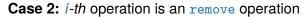


- Case 2: *i-th* operation is an remove operation
 - Case 2.1: Borrow a node
 - Creates no additional operations

Runtime Complexity - (2,4)-Tree



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



- 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

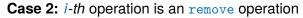


- Case 2.1: Borrow a node
 - Creates no additional operations

Runtime Complexity - (2,4)-Tree



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



- 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





(a,b)-Trees

Runtime Complexity - (2,4)-Tree



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



■ Case 2.2: Merging two node



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

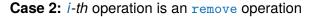




Figure: Merging two nodes



Figure: Merging two nodes

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

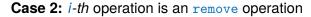




Figure: Merging two nodes

- 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:

Lemma:

■ We know:

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

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:

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

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

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

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

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



Balanced Trees

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

Introduction



Introduction

Red-Black Tree:

■ Binary tree with red and black nodes

Introduction



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

- 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)

- 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

Introduction



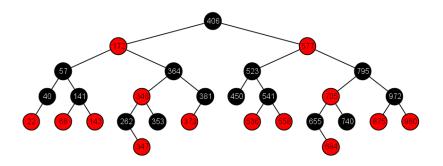


Figure: Example of an red-black-tree [Gna]

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

[Wik] Red-black tree

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