

Algorithms and Datastructures

Balanced Trees (AVL-Trees, (a,b)-Trees, Red-Black-Trees)

Albert-Ludwigs-Universität Freiburg



**UNI
FREIBURG**

Prof. Dr. Rolf Backofen

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

Balanced Trees

- Motivation

- AVL-Trees

- (a,b)-Trees

 - Introduction

 - Runtime Complexity

- Red-Black Trees

Balanced Trees

Motivation

AVL-Trees

(a,b)-Trees

Introduction

Runtime Complexity

Red-Black Trees

Balanced Trees

Motivation

AVL-Trees

(a,b)-Trees

Introduction

Runtime Complexity

Red-Black Trees



Binary search tree:

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: $d = O(\log n)$

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: $d = O(\log n)$
 - If the 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: $d = O(\log n)$
 - If the keys are inserted randomly
- Worst case: $d = O(n)$

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



Gnarley trees:



Gnarley trees:

- <http://people.ksp.sk/~kuko/bak>



Gnarley trees:

■ <http://people.ksp.sk/~kuko/bak>



Figure: Binary search tree with random insert [Gna]



Gnarley trees:

■ <http://people.ksp.sk/~kuko/bak>

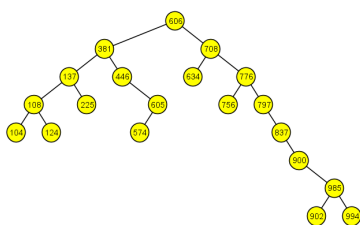


Figure: Binary search tree with random insert [Gna]

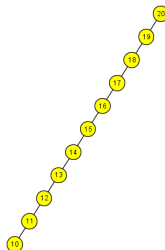


Figure: Binary search tree with descending insert [Gna]



Balanced trees:

Balanced trees:

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

Balanced trees:

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

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 $O(\log n)$?



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

- AVL-Tree:



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

- **AVL-Tree:**

- Binary tree with 2 children per node

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

- **AVL-Tree:**

- Binary tree with 2 children per node
- Balancing via “rotation”

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:**

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 have between a and b children

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 have between a and b children
- Balancing through **splitting** and **merging** nodes

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 have between a and b children
- Balancing through **splitting** and **merging** nodes
- Used in data bases and file systems

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 have between a and b children
- Balancing through **splitting** and **merging** nodes
- Used in data bases and file systems

- **Red-Black-Tree:**

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 have between a and b children
- Balancing through **splitting** and **merging** nodes
- Used in data bases and file systems

- **Red-Black-Tree:**

- Binary tree with “black” and “red” nodes

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 have between a and b children
- Balancing through **splitting** and **merging** nodes
- Used in data bases and file systems

- **Red-Black-Tree:**

- Binary tree with “black” and “red” nodes
- Balancing through “rotation” and “recoloring”

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 have between a and b children
- Balancing through **splitting** and **merging** nodes
- Used in data bases 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

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 have between a and b children
- Balancing through **splitting** and **merging** nodes
- Used in data bases 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`, Java `SortedMap`

Balanced Trees

Motivation

AVL-Trees

(a,b)-Trees

Introduction

Runtime Complexity

Red-Black Trees



AVL-Tree:

AVL-Tree:

- Gregory Maximovich **A**delson-**V**elskii, Yevgeniy Mikhailovlovich **L**andis (1963)

AVL-Tree:

- Gregory Maximovich **A**delson-**V**elskii, Yevgeniy Mikhailovlovich **L**andis (1963)
- Search tree with modified **insert** and **remove** operations while satisfying a **depth** condition

AVL-Tree:

- Gregory Maximovich **A**delson-**V**elskii, Yevgeniy Mikhailovlovich **L**andis (1963)
- Search tree with modified **insert** and **remove** operations while satisfying a **depth** condition
- Prevents degeneration of the search tree

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

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 $O(\log n)$

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 $O(\log n)$



Figure: Example of an AVL-Tree

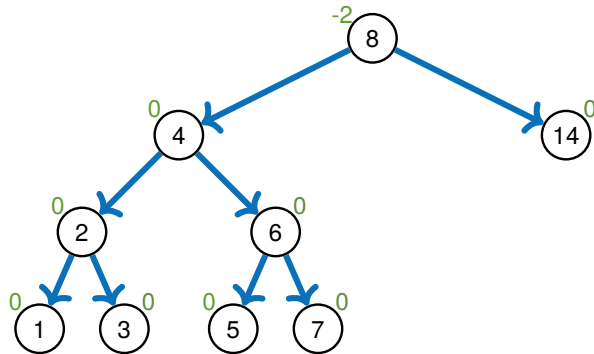


Figure: **Not** an AVL-Tree



Figure: Another example of an AVL-Tree

Rotation:

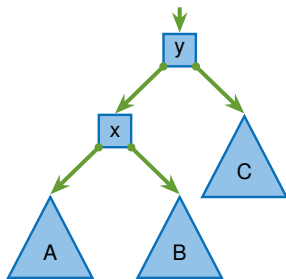


Figure: Before rotating



Figure: After rotating

Rotation:

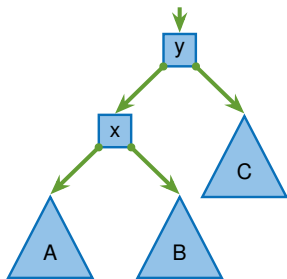


Figure: Before rotating

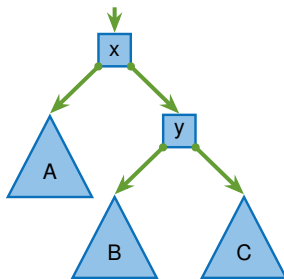


Figure: After rotating

- Central operation of **rebalancing**

Rotation:

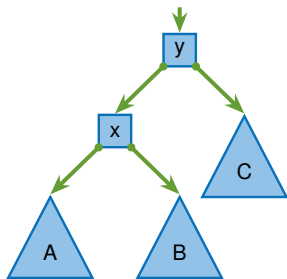


Figure: Before rotating

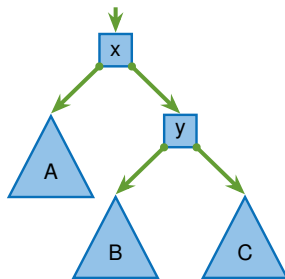


Figure: After rotating

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

Rotation:



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

Rotation:



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:

AVL-Tree:

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

AVL-Tree:

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

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,...



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,...
- <http://people.ksp.sk/~kuko/bak>

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,...
- <http://people.ksp.sk/~kuko/bak>



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



Summary:

Summary:

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

Summary:

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

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

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

Balanced Trees

Motivation

AVL-Trees

(a,b)-Trees

Introduction

Runtime Complexity

Red-Black Trees



(a,b) -Tree:

(a,b)-Tree:

- Also known as **b-tree** (b for “balanced”)

(a,b)-Tree:

- Also known as **b-tree** (b for “balanced”)
- Used in data bases and file systems

(a,b)-Tree:

- Also known as **b-tree** (b for “balanced”)
- Used in data bases and file systems

Idea:

(a,b)-Tree:

- Also known as **b-tree** (b for “balanced”)
- Used in data bases and file systems

Idea:

- Save a varying number of elements per node

(a,b)-Tree:

- Also known as **b-tree** (b for “balanced”)
- Used in data bases 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:



(a,b) -Tree:

- All leaves have the same depth

(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)

(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)



(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)



- Each node with n children is called “node of degree n ” and holds $n - 1$ sorted elements

(a,b)-Tree:

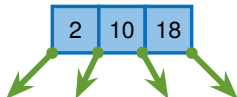
- All leaves have the same depth
- Each inner node has $\geq a$ and $\leq b$ nodes
(Only the root node may have less nodes)



- Each node with n children is called “node of degree n ” and holds $n - 1$ sorted elements
- Subtrees are located “between” the elements

(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)



- 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: $a \geq 2$ and $b \geq 2a - 1$

(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:



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

Not an (2,4)-Tree:

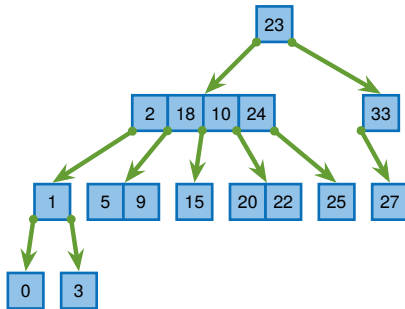


Figure: **Not** an (2,4)-tree

- Invalid sorting

Not an (2,4)-Tree:

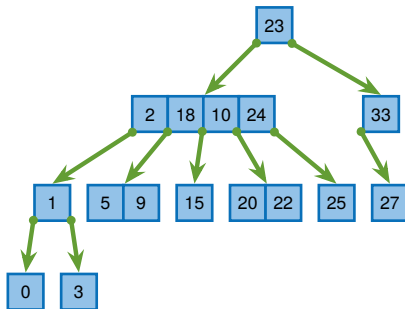


Figure: **Not** an (2,4)-tree

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

Not an (2,4)-Tree:

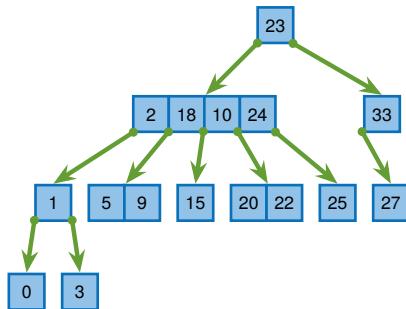


Figure: **Not** an (2,4)-tree

- 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](#)

Searching an element: (lookup)

- The same algorithm as in [BinarySearchTree](#)
- Searching from the root downwards

Searching an element: (lookup)

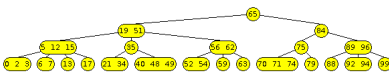
- The same algorithm as in [BinarySearchTree](#)
- Searching from the root downwards
- The keys at each node set the path

Searching an element: (lookup)

- The same algorithm as in [BinarySearchTree](#)
- Searching from the root downwards
- The keys at each node set the path

BST AVL tree B tree Red-black tree AA tree Skiplist Max Heap Min Heap Treap Scapegoat tree Splay tree

Display



Control

50 Insert Find Delete Next

☐ Pause ☐ Small 4

#Nodes: 22 #Keys: 37 = 56% full Height: 3

Text

Search Found.

Inserting an element: (`insert`)

Inserting an element: (`insert`)

- Search the position to insert the key into

Inserting an element: (`insert`)

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

Inserting an element: (`insert`)

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

Inserting an element: (`insert`)

- Search the position to insert the key into
- This position will always be an leaf
- Insert the element into the tree
- **Attention:** Nodes can have one element too many (Degree $b + 1$)

Inserting an element: (`insert`)

- Search the position to insert the key into
- This position will always be an leaf
- Insert the element into the tree
- **Attention:** Nodes can have one element too many (Degree $b + 1$)
- Then we **split** the node

Inserting an element: (`insert`)



Figure: Splitting a node

Inserting an element: (`insert`)



Figure: Splitting a node

- If the degree is higher than $b + 1$ 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 $\text{ceil}\left(\frac{b-1}{2}\right)$ elements, a element for the parent node, and a node with $\text{floor}\left(\frac{b-1}{2}\right)$ elements

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 $\lceil \frac{b-1}{2} \rceil$ elements, a element for the parent node, and a node with $\lfloor \frac{b-1}{2} \rfloor$ elements
 - That's why we have the limit $b \geq 2a - 1$

Inserting an element: (`insert`)

Inserting an element: (`insert`)

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

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 $b + 1$

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

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 the node to split is the root we split it and create a new root node
(The tree is now one level deeper)



Removing an element: (`remove`)



Removing an element: (`remove`)

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

Removing an element: (`remove`)

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

Removing an element: (`remove`)

- Search the element in $O(\log n)$ time
- **Case 1:** The element is contained by a leaf, remove it
- **Case 2:** The element is contained by an inner node

Removing an element: (`remove`)

- Search the element in $O(\log n)$ time
- **Case 1:** The element is contained by a leaf, remove it
- **Case 2:** The element is contained by an inner node
 - Search the `successor` in the right subtree

Removing an element: (`remove`)

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

Removing an element: (`remove`)

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

Removing an element: (`remove`)

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



Removing an element: (`remove`)

Removing an element: (remove)

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

Removing an element: (remove)

- **Attention:** The leaf might be too small (degree of $a - 1$)
⇒ 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

Removing an element: (`remove`)

- **Attention:** The leaf might be too small (degree of $a - 1$)
⇒ 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: Borrowing an element



Removing an element: (`remove`)

Removing an element: (remove)

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

Removing an element: (remove)

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

Removing an element: (remove)

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



Figure: Combining two nodes



Removing an element: (`remove`)

Removing an element: (remove)

- Now the parent node can be of degree $a - 1$

Removing an element: (remove)

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

Removing an element: (remove)

- Now the parent node can be of degree $a - 1$
- We combine parent nodes the same way
- If the root has only one child left we take the child as new root
(The tree shrinks one level)



Runtime complexity of `lookup`, `insert` and `remove`:

Runtime complexity of **lookup**, **insert** and **remove**:

- All operations in $O(d)$ with d being the depth of the tree

Runtime complexity of **lookup**, **insert** and **remove**:

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

Runtime complexity of **lookup**, **insert** and **remove**:

- All operations in $O(d)$ with d being the depth of the tree
- Each node (except the root) has more than a children
 $\Rightarrow n \geq a^{d-1}$ and $d \leq 1 + \log_a n = O(\log_a n)$
- If we look closer:

Runtime complexity of **lookup**, **insert** and **remove**:

- All operations in $O(d)$ with d being the depth of the tree
- Each node (except the root) has more than a children
 $\Rightarrow n \geq a^{d-1}$ and $d \leq 1 + \log_a n = O(\log_a n)$
- If we look closer:
 - **lookup** always takes $\Theta(d)$

Runtime complexity of **lookup**, **insert** and **remove**:

- All operations in $O(d)$ with d being the depth of the tree
- Each node (except the root) has more than a children
 $\Rightarrow n \geq a^{d-1}$ and $d \leq 1 + \log_a n = O(\log_a n)$
- If we look closer:
 - **lookup** always takes $\Theta(d)$
 - **insert** and **remove** often require only $O(1)$ time

Runtime complexity of **lookup**, **insert** and **remove**:

- All operations in $O(d)$ with d being the depth of the tree
- Each node (except the root) has more than a children
 $\Rightarrow n \geq a^{d-1}$ and $d \leq 1 + \log_a n = O(\log_a n)$
- If we look closer:
 - **lookup** always takes $\Theta(d)$
 - **insert** and **remove** often require only $O(1)$ time
 - Only in the **worst case** we have to **split** or **combine** all nodes on a path up to the root

Runtime complexity of **lookup**, **insert** and **remove**:

- All operations in $O(d)$ with d being the depth of the tree
- Each node (except the root) has more than a children
 $\Rightarrow n \geq a^{d-1}$ and $d \leq 1 + \log_a n = O(\log_a n)$
- If we look closer:
 - **lookup** always takes $\Theta(d)$
 - **insert** and **remove** often require only $O(1)$ time
 - Only in the **worst case** we have to **split** or **combine** all nodes on a path up to the root
 - We want to analyse in detail

Runtime complexity of **lookup**, **insert** and **remove**:

- All operations in $O(d)$ with d being the depth of the tree
- Each node (except the root) has more than a children
 $\Rightarrow n \geq a^{d-1}$ and $d \leq 1 + \log_a n = O(\log_a n)$
- If we look closer:
 - **lookup** always takes $\Theta(d)$
 - **insert** and **remove** often require only $O(1)$ time
 - Only in the **worst case** we have to **split** or **combine** all nodes on a path up to the root
 - We want to analyse in detail
 - Therefore instead of $b \geq 2a - 1$ we need $b \geq 2a$.

Runtime complexity of **lookup**, **insert** and **remove**:

- All operations in $O(d)$ with d being the depth of the tree
- Each node (except the root) has more than a children
 $\Rightarrow n \geq a^{d-1}$ and $d \leq 1 + \log_a n = O(\log_a n)$
- If we look closer:
 - **lookup** always takes $\Theta(d)$
 - **insert** and **remove** often require only $O(1)$ time
 - Only in the **worst case** we have to **split** or **combine** all nodes on a path up to the root
 - We want to analyse in detail
 - Therefore instead of $b \geq 2a - 1$ we need $b \geq 2a$.
 - Here is a counter-example for (2,3)-trees, analysis of (2,4)-trees

(a,b) -Trees

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



$(2,3)$ -Tree:



(2,3)-Tree:

- Before executing `delete(11)`

(2,3)-Tree:

- Before executing `delete(11)`

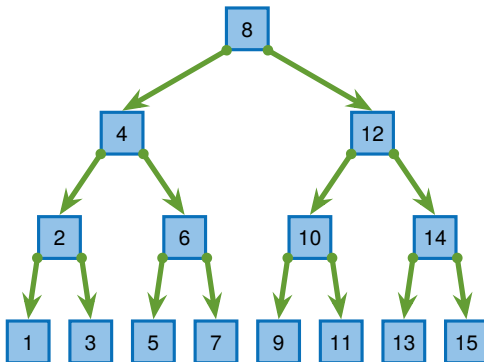


Figure: Normal (2,3)-Tree

(2,3)-Tree:

- Executing `delete(11)`

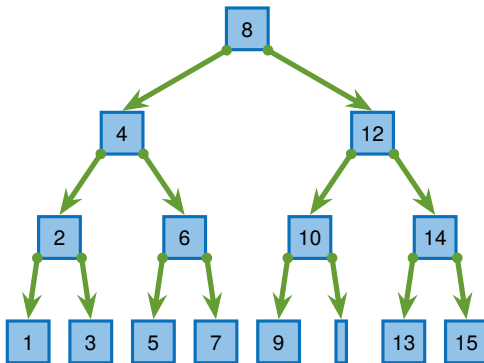


Figure: (2,3)-Tree - Delete step 1

(2,3)-Tree:

- Executing `delete(11)`

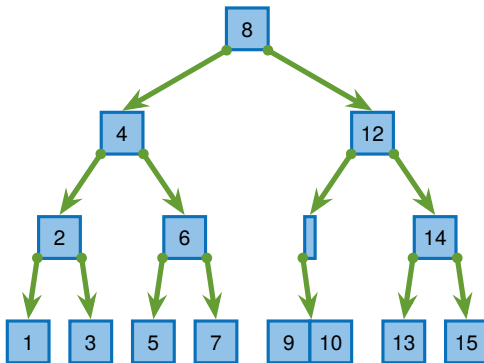


Figure: (2,3)-Tree - Delete step 2

(2,3)-Tree:

- Executing `delete(11)`

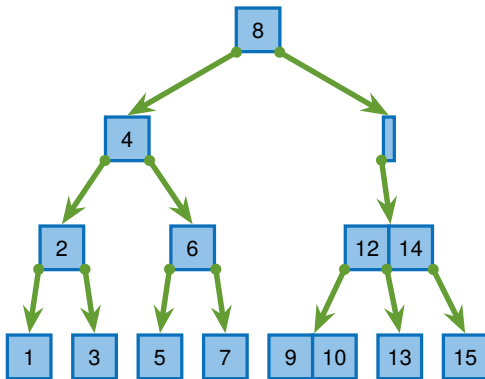


Figure: (2,3)-Tree - Delete step 3

(2,3)-Tree:

- Executed `delete(11)`

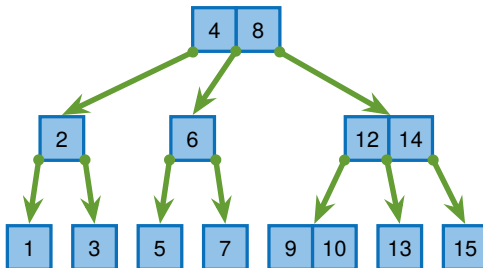


Figure: (2,3)-Tree - Delete step 4

(a,b) -Trees

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



$(2,3)$ -Tree:



(2,3)-Tree:

- Executing `insert(11)`

(2,3)-Tree:

- Executing `insert(11)`

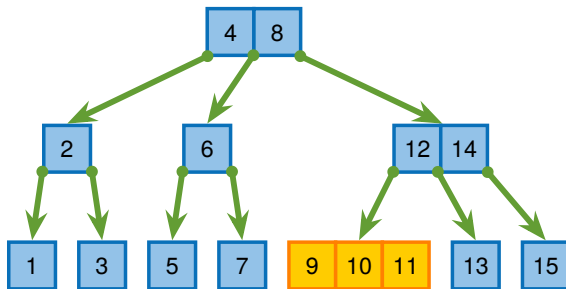


Figure: (2,3)-Tree - Insert step 1

(2,3)-Tree:

- Executing `insert(11)`

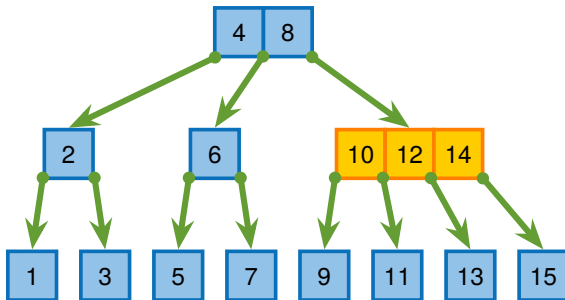


Figure: (2,3)-Tree - Insert step 2

(2,3)-Tree:

- Executing `insert(11)`

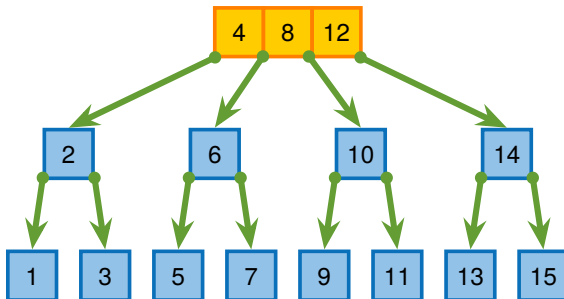


Figure: (2,3)-Tree - Insert step 3

(2,3)-Tree:

- Executed `insert(11)`



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

(2,3)-Tree:



Figure: (2,3)-Tree

(2,3)-Tree:

- We are exactly where we started

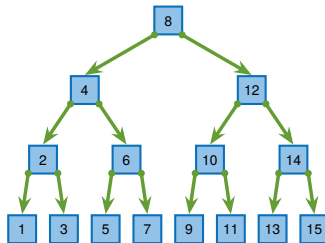


Figure: (2,3)-Tree

(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 $O(\log n)$

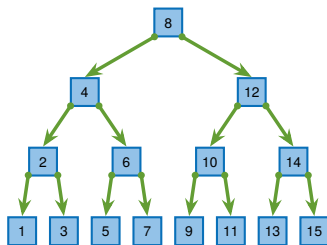


Figure: (2,3)-Tree

(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 $b \geq 2a$ instead of $b \geq 2a - 1$

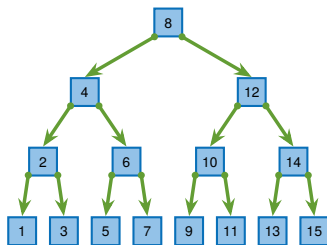


Figure: (2,3)-Tree



$(2,4)$ -Tree:

(2,4)-Tree:

- If all nodes have 2 children we have to combine the nodes up to the root on a remove operation

(2,4)-Tree:

- If all nodes have 2 children we have to combine 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

(2,4)-Tree:

- If all nodes have 2 children we have to combine 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

(2,4)-Tree:

- If all nodes have **2 children** we have to **combine** 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 harmless**

Neither an **insert** nor a **remove** operation trigger rebalancing operations



$(2,4)$ -Tree:



$(2,4)$ -Tree:

- Idea:

(2,4)-Tree:

- **Idea:**

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

(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

(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:

(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

(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:



Terminology:

- We analyze a sequence of n operations

Terminology:

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

Terminology:

- We analyze a sequence of n operations
- Let ϕ_i be the potential of the tree after the i -th operation
- ϕ_i = is the number of nodes with degree 3



Example:



Example:

- Nodes of degree 3 are highlighted

Example:

- Nodes of degree 3 are highlighted



Figure: Tree with potential $\phi = 4$



Terminology:



Terminology:

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

Terminology:

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

Terminology:

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

Terminology:

- Let c_i be the costs = runtime of the i -th operation
- We will show:
 - Each operation can maximally destroy one harmless node
 - For each further step, that incurs cost, the operation creates a further harmless node

Terminology:

- Let c_i be the costs = runtime of the i -th operation
- We will show:
 - Each operation can maximally destroy one harmless node
 - For each further step, that incurs cost, the operation creates a further harmless node
- The costs for operation i are coupled to the difference of the potential levels

$$c_i \leq A \cdot (\underbrace{\phi_i - \phi_{i-1}}_{\text{difference of potential levels}}) + B, \quad A > 0 \text{ and } B > A$$

Number of harmless (degree 3) nodes at operation i . Can be -1 , but not smaller than -1

Terminology:

- Let c_i be the costs = runtime of the i -th operation
- We will show:
 - Each operation can maximally destroy one harmless node
 - For each further step, that incurs cost, the operation creates a further harmless node
- The costs for operation i are coupled to the difference of the potential levels

$$c_i \leq A \cdot \underbrace{(\phi_i - \phi_{i-1})} + B, \quad A > 0 \text{ and } B > A$$

Number of harmless (degree 3) nodes at operation i . Can be -1 , but not smaller than -1

- With that each operation has an amortized cost of $O(1)$

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

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



Figure: Splitting a node on `insert`

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**

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

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



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

- Let m be the number of nodes split

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

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

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

$$\begin{aligned}\phi_i &\geq \phi_{i-1} + m - 1 \\ \Rightarrow m &\leq \phi_i - \phi_{i-1} + 1\end{aligned}$$

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

$$\begin{aligned}\phi_i &\geq \phi_{i-1} + m - 1 \\ \Rightarrow m &\leq \phi_i - \phi_{i-1} + 1\end{aligned}$$

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

$$\begin{aligned}\Rightarrow c_i &\leq A \cdot (\phi_i - \phi_{i-1} + 1) + B \\ c_i &\leq A \cdot (\phi_i - \phi_{i-1}) + \underbrace{A + B}_{B'}\end{aligned}$$



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



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

- **Case 2.1:** Inner node



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

- **Case 2.1:** Inner node

- Searching the successor in a tree is $O(d) = O(\log n)$

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
⇒ We can find the successor in $O(1)$

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
⇒ We can find the successor in $O(1)$



Figure: Tree with doubly linked list



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



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

- **Case 2.1:** Borrowing a node

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

- **Case 2.1:** Borrowing a node
 - Creates no additional operations

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

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

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

■ **Case 2.1:** Borrowing a node

- Creates no additional operations
- Case 2.1.1: Potential rises by one



Figure: Borrowing an element case 2.1.1



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



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

- **Case 2.1:** Borrowing a node



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

- **Case 2.1:** Borrowing a node
 - Creates no additional operations

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

- **Case 2.1:** Borrowing a node
 - Creates no additional operations
 - Case 2.1.2: Potential lowers by one

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

■ **Case 2.1:** Borrowing a node

- Creates no additional operations
- Case 2.1.2: Potential lowers by one



Figure: Borrowing an element case 2.1.2



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



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

- **Case 2.2:** Merging a node

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

■ **Case 2.2:** Merging a node



Figure: Merging two nodes

■ Potential rises by one

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

■ **Case 2.2:** Merging a node

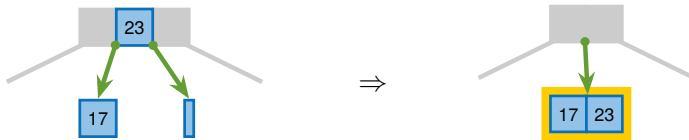


Figure: Merging two nodes

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

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

■ **Case 2.2:** Merging a 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 degree 2 node, which can borrow from a neighbour

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

■ **Case 2.2:** Merging a node

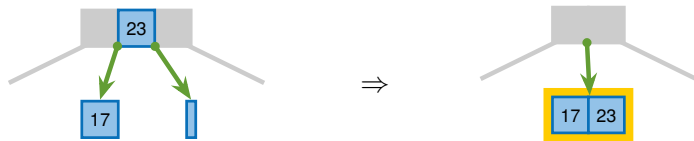


Figure: Merging two nodes

- 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 degree 2 node, which can borrow from a neighbour
- The potential rises by m

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

■ **Case 2.2:** Merging a node

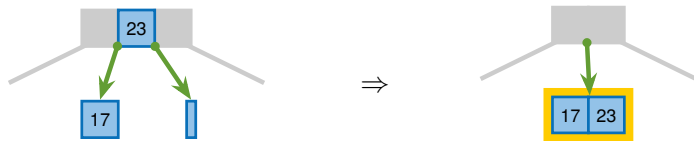


Figure: Merging two nodes

- 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 degree 2 node, which can borrow from a neighbour
- The potential rises by m
- If the “stop-node” is of **degree 2** then the potential eventually goes down by one

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

■ **Case 2.2:** Merging a node

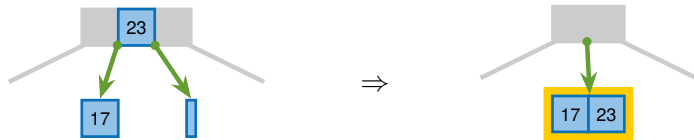


Figure: Merging two nodes

- 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 degree 2 node, which can borrow from a neighbour
- 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:

$$c_i \leq A \cdot (\phi_i - \phi_{i-1}) + B, \quad A > 0 \text{ and } B > A$$

Lemma:

- We know:

$$c_i \leq A \cdot (\phi_i - \phi_{i-1}) + B, \quad A > 0 \text{ and } B > A$$

- With that we can conclude:

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

Proof:

$$\begin{aligned}\sum_{i=0}^n c_i &\leq \underbrace{A \cdot (\phi_1 - \phi_0) + B}_{\leq c_1} + \underbrace{A \cdot (\phi_2 - \phi_1) + B}_{\leq c_1} + \dots + \underbrace{A \cdot (\phi_n - \phi_{n-1}) + B}_{\leq c_n} \\ &= A \cdot (\phi_n - \phi_0) + B \cdot n && | \text{ telescope sum} \\ &= A \cdot \phi_n + B \cdot n && | \text{ we start with an empty tree} \\ &< A \cdot n + B \cdot n = O(n) && | \text{ number of degree 3 nodes} \\ &&& < \text{ number of nodes}\end{aligned}$$

Balanced Trees

Motivation

AVL-Trees

(a,b)-Trees

Introduction

Runtime Complexity

Red-Black Trees



Red-Black Tree:

Red-Black Tree:

- Binary tree with red and black nodes

Red-Black Tree:

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

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)

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



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)

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