Data Structures

Search Trees

Dimitrios Michail



Dept. of Informatics and Telematics Harokopio University of Athens

The problem

Search

We would like to maintain items with keys and except from add/remove to also be able to quickly search an item based on a key.

Example

Assume items are bank transactions we would like to search based on the date.

Search

We can implement a data structure for searching in multiple ways.

Our goal is to implement the search operations efficiently while keeping the remaining operations also efficient.

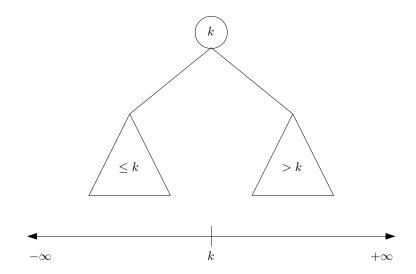
e.g. searching on a linked list is not efficient!

Binary Search Tree (BST)

Definition

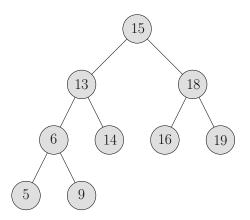
A **binary search tree** is a binary tree where each node is associated with a key with the additional property that the key of a node is larger (or equal) from the keys of all nodes in its left subtree, and smaller that all the keys of the nodes in its right subtree.

Binary Search Tree (BST)



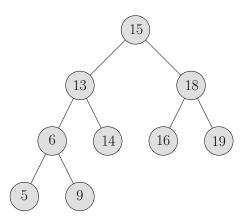
Example

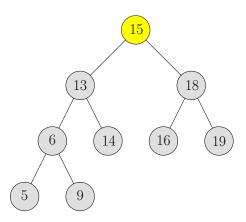
Binary Search Tree (BST)

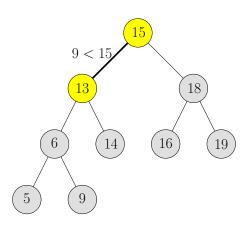


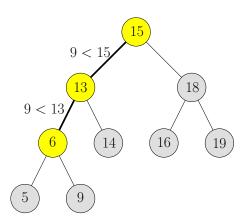
Algorithm

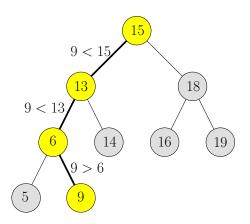
- start from the root
- ▶ if we are at a node which has the same key as the one we are looking for, then return the node
- otherwise go left or right depending on the comparison between the key we are looking for and the key of the node

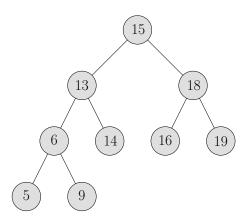


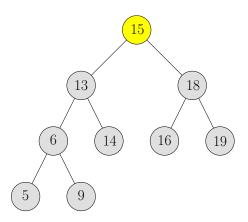


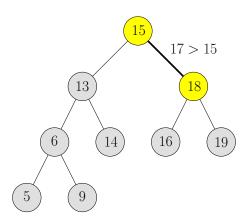


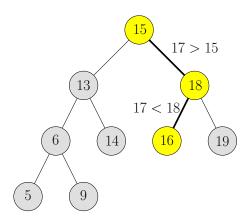


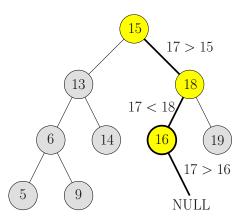












Inorder Traversal

Rule

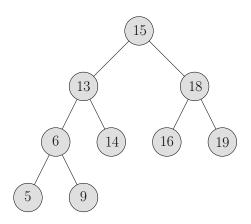
First visit the left subtree recursively, then the root then the right subtree recursively.

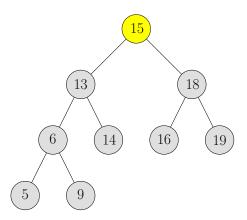
Returns nodes in non-decreasing key order.

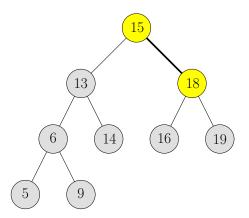
The traversal needs linear time $\mathcal{O}(n)$ where n are the number of items in the tree.

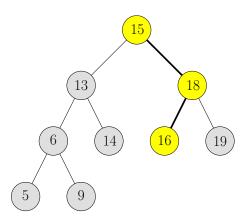
Algorithm

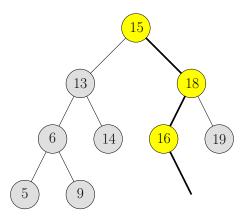
- ▶ first perform a search
- ▶ if the key is found, do nothing
- otherwise add a new node with the new key at the point where the search ended

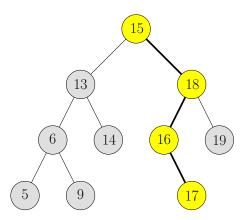


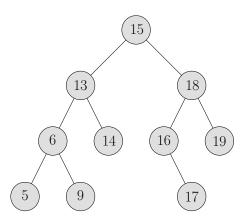






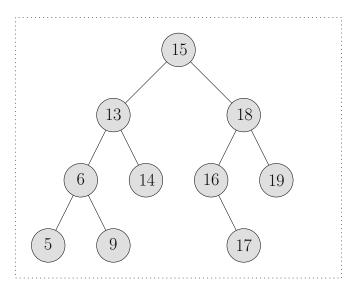


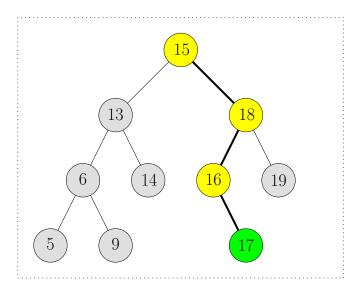


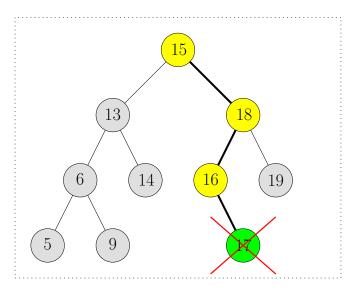


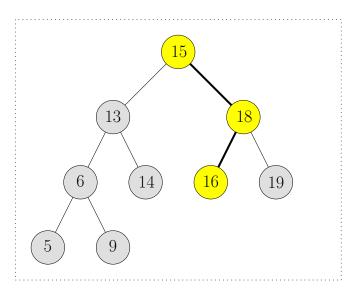
Algorithm

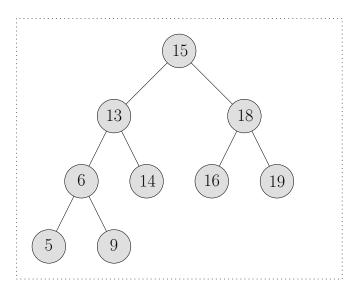
- ▶ first perform a search
- ▶ if the key is not found, do nothing
- otherwise there are 3 cases based on the number of children of the node we want to delete



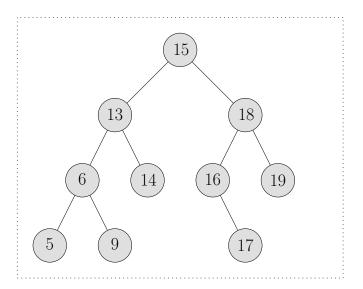




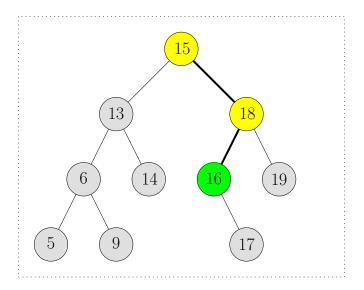




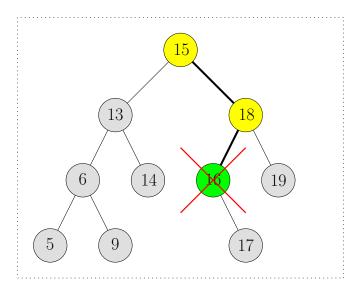
Case 2: Deletion of 16 (one child)

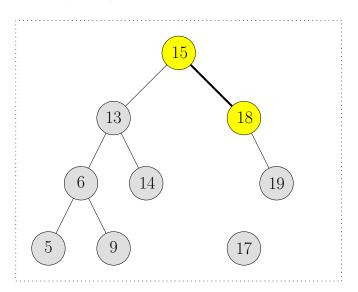


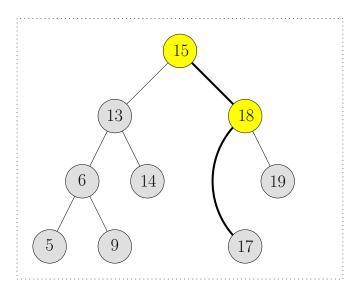
Case 2: Deletion of 16 (one child)

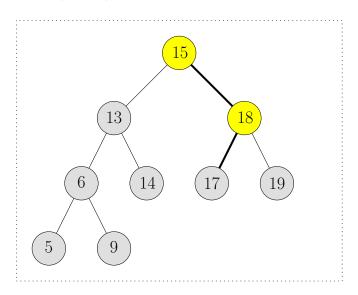


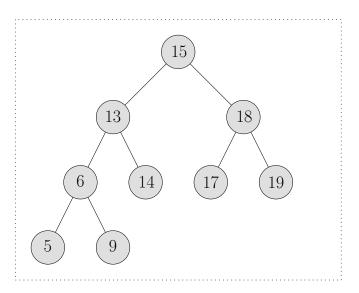
Case 2: Deletion of 16 (one child)

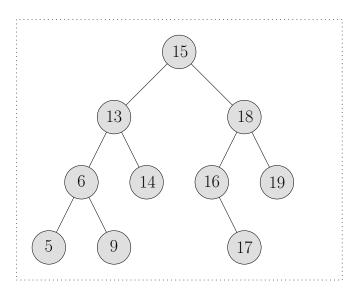










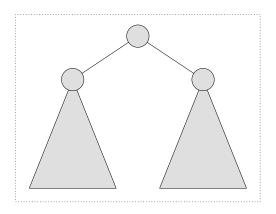


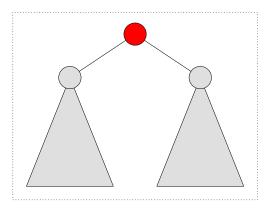
Calculate the successor

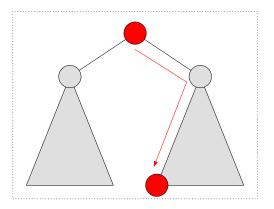
Useful routine for deletion

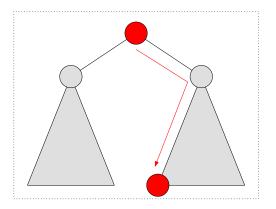
The problem

Given a subtree we would like to find the node with the successor key from the root of the subtree

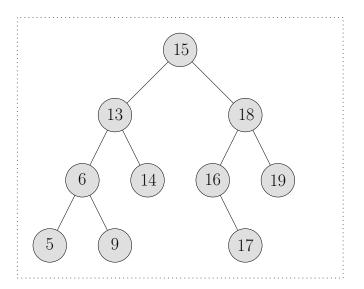


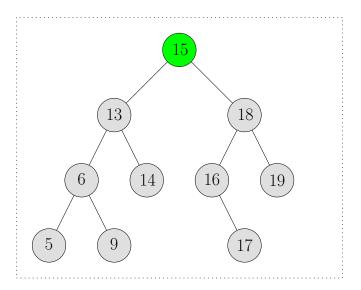


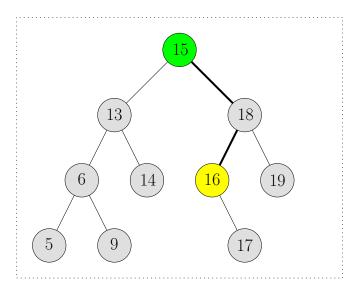


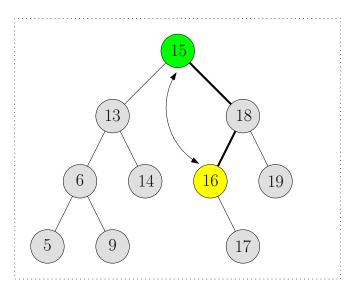


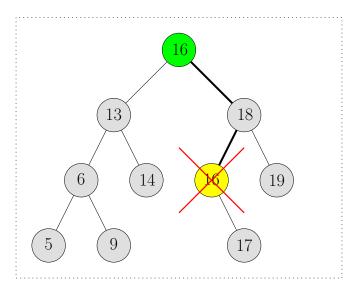
The successor of the root has at most one child. (why?)

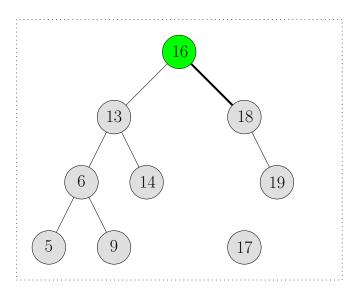


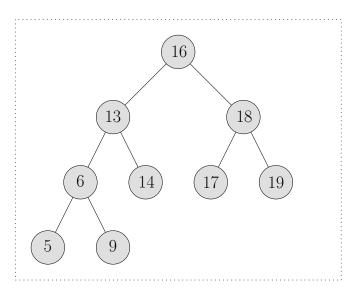








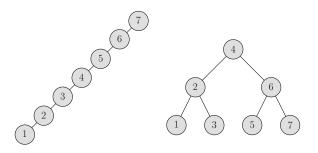




Is it fast?

All operations require time proportional to the height of the tree.

Problems



There are many trees for the same set of items

- worst case on the left: height $h = \mathcal{O}(n)$
- **b** best case on the right: height $h = \mathcal{O}(\log n)$

The tree depends on the order that the insertions and deletions are performed, something that we cannot know beforehand



Balanced BSTs

Search trees which are balanced at any point in time independently of the order that the insertions and/or deletions of items are performed.

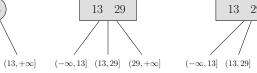
$$h = \mathcal{O}(\log n)$$

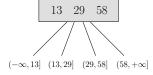
2-3-4 Search Trees

top-down

We allow nodes which have 2, 3 or 4 children.







2-3-4 Search Trees

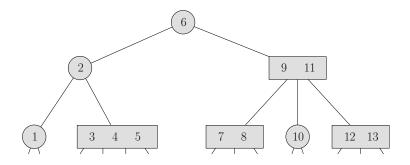
top-down

Definition

A balanced 2-3-4 search tree is a tree which:

- ▶ is either empty
- or is comprised of tree different node types: 2-nodes, 3-nodes4-nodes
- ▶ all the external nodes (nulls null links) are at the exact same distance from the root

Example of a Balanced 2-3-4 Search Tree



Theorem

A 2-3-4 search tree with n keys has height $\mathcal{O}(\log n)$.

Proof.

Let *h* be the height of the tree.

Theorem

A 2-3-4 search tree with n keys has height $\mathcal{O}(\log n)$.

Proof.

Let *h* be the height of the tree.

Every level i has at least 2^i nodes.

Theorem

A 2-3-4 search tree with n keys has height $O(\log n)$.

Proof.

Let *h* be the height of the tree.

Every level i has at least 2^i nodes.

Thus
$$n \ge 1 + 2 + 4 + \dots 2^{h-1} = 2^h - 1$$
.

Theorem

A 2-3-4 search tree with n keys has height $O(\log n)$.

Proof.

Let *h* be the height of the tree.

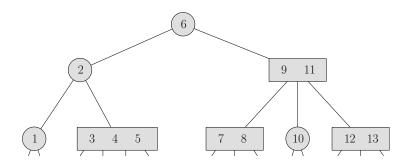
Every level i has at least 2^i nodes.

Thus
$$n \ge 1 + 2 + 4 + \dots 2^{h-1} = 2^h - 1$$
.

Taking logarithms we get that $log_2(n+1) \ge h$.



Search on a Balanced 2-3-4 Search Tree

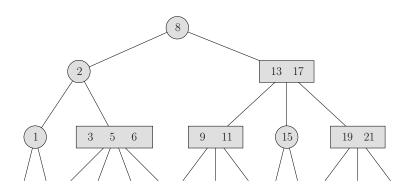


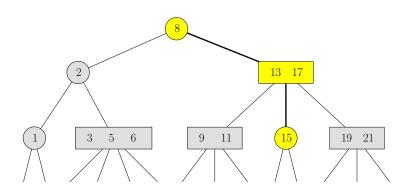
Search on a Balanced 2-3-4 Search Tree: Generalization of the BST algorithm

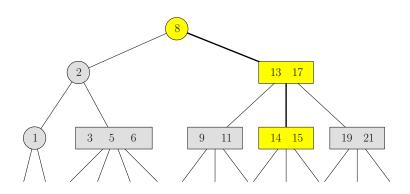
While maintaining the balance!

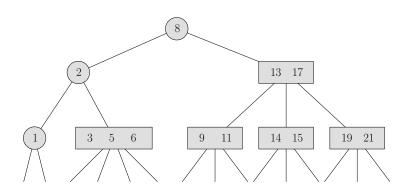
We search for the new key in order to find where we are supposed to add it.

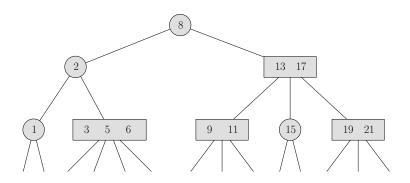
- ▶ if the search ends on a 2-node we simply make it a 3-node
- ▶ if the search ends on a 3-node we make it a 4-node
- ▶ if the search ends on a 4-node
 - break the 4-node into two 2-nodes by sending the middle key upwards, and then
 - add the new key in one of the two 2-nodes (we might need to do the same procedure at the parent node, possibly multiple times all the way up to the root)

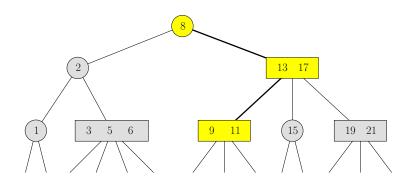


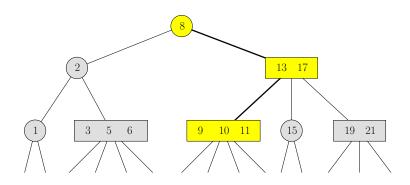


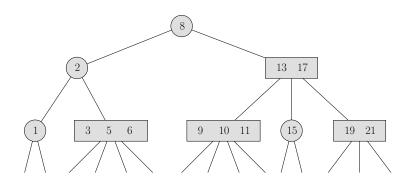


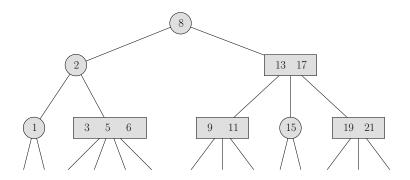


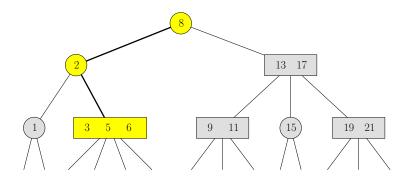


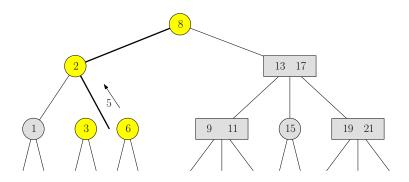


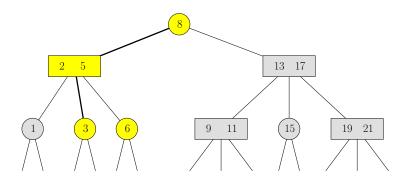


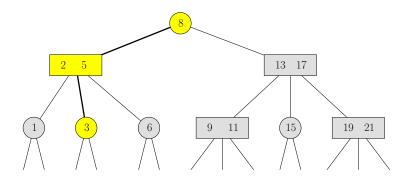


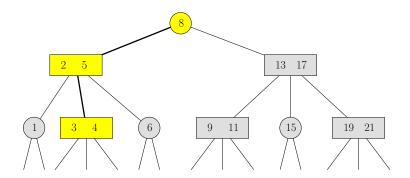


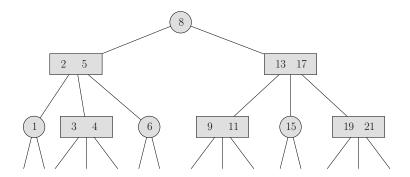




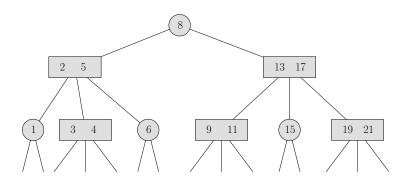








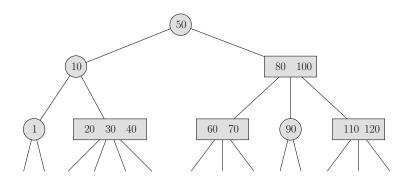
Ends at a 4-node (insertion of 4)

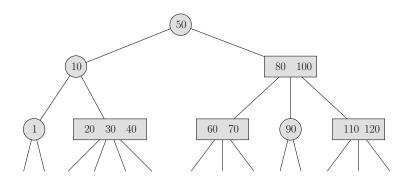


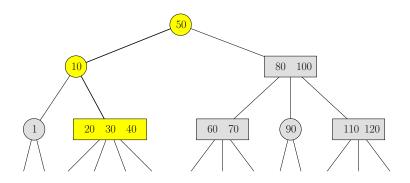
There is also the case where the father is also a 4-node. In this case we need to perform the split again, possible multiple times all the way to the root

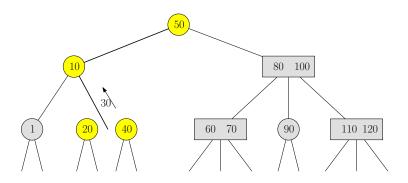
In order to simplify insertion and to avoid such cascades, we use the following technique:

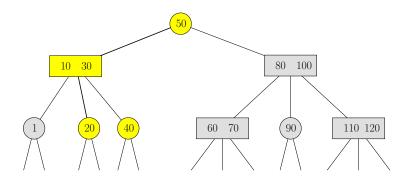
we make sure that the search path does not contain 4-nodes, by spliting any 4-node during our **descend** on the tree

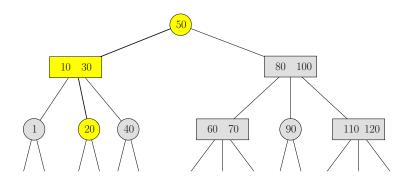


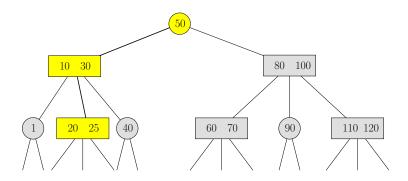


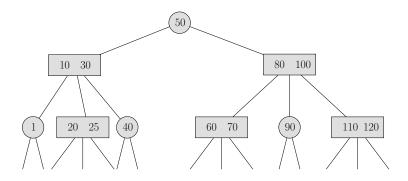


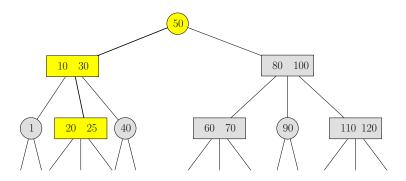


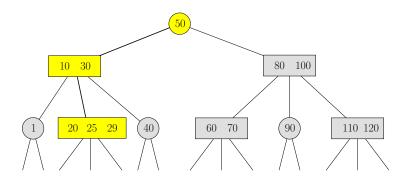


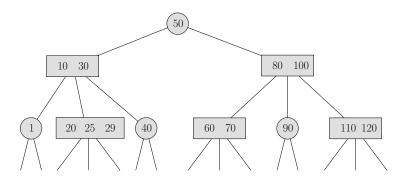


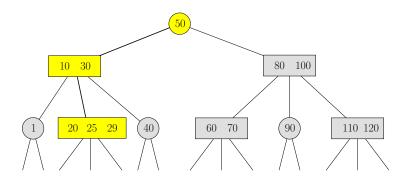


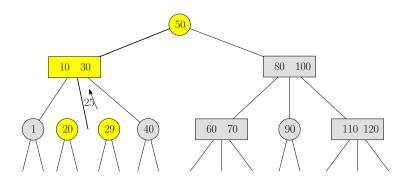


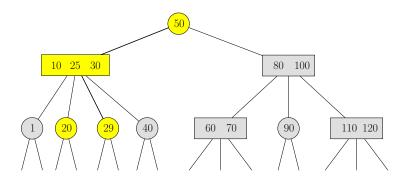


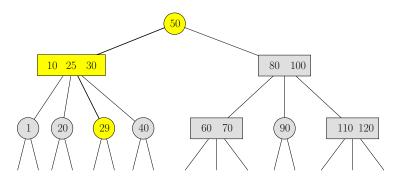


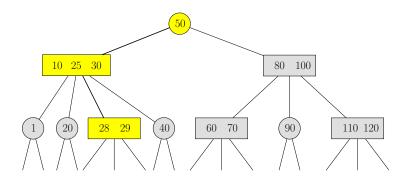


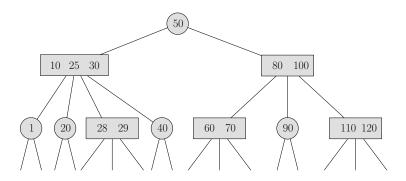


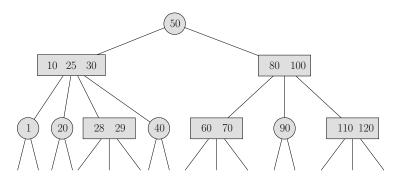


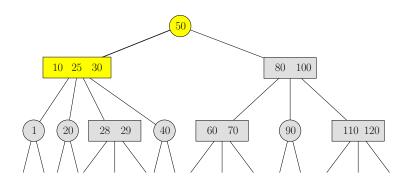


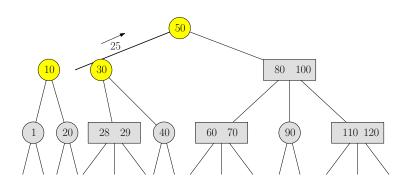


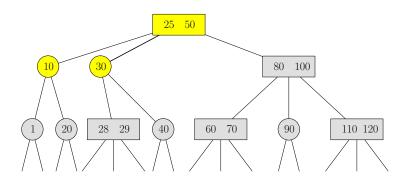


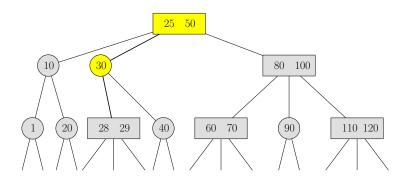


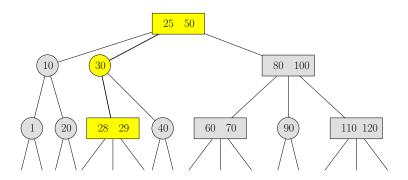


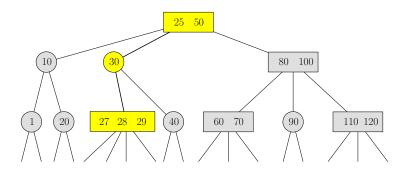


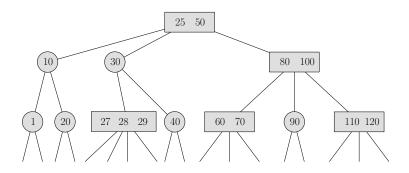


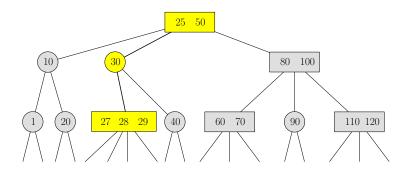


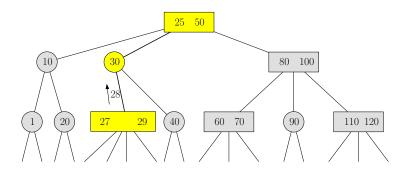


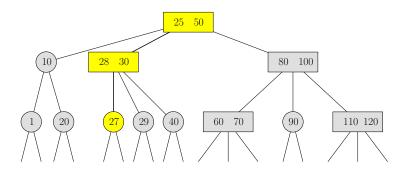


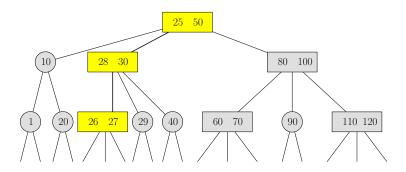


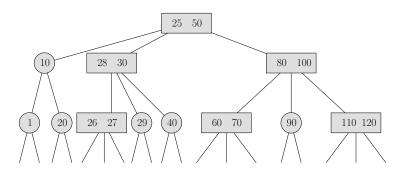








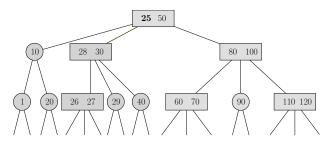




Deletion from a non-leaf

- ▶ We reduce the problem to the deletion of a leaf
- ► Same technique as in the BSTs

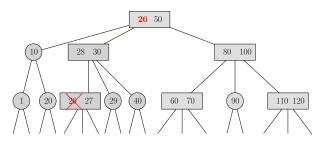
e.g. to remove 25, we put in its place the *inorder successor* (26) or the *inorder predecessor* (20).



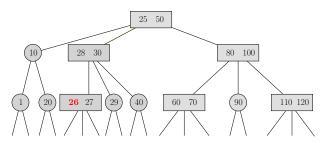
Deletion from a non-leaf

- ▶ We reduce the problem to the deletion of a leaf
- ► Same technique as in the BSTs

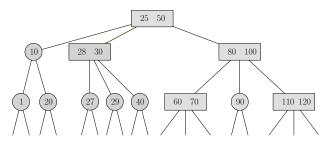
e.g. to remove 25, we put in its place the *inorder successor* (26) or the *inorder predecessor* (20).



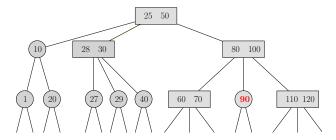
- ▶ The leaf has \geq 2 keys (e.g. delete 26 elow), which means that it is a 3-node or a 4-node.
- Easy case, just delete by switching node types.



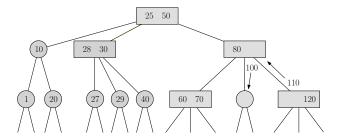
- ▶ The leaf has \geq 2 keys (e.g. delete 26 elow), which means that it is a 3-node or a 4-node.
- Easy case, just delete by switching node types.



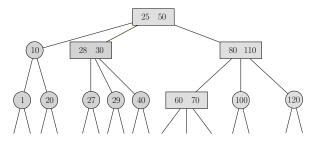
- ▶ The leaf has 1 key (underflow), which means it is a 2-node.
- ▶ some sibling node has ≥ 2 keys
- **balance**: e.g. delete 90



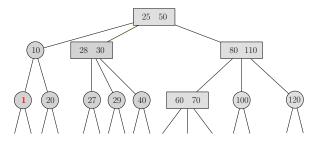
- ▶ The leaf has 1 key (underflow), which means it is a 2-node.
- ▶ some sibling node has ≥ 2 keys
- **balance**: e.g. delete 90



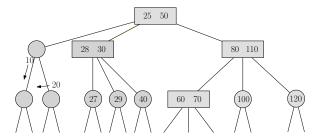
- ▶ The leaf has 1 key (underflow), which means it is a 2-node.
- ▶ some sibling node has ≥ 2 keys
- **balance**: e.g. delete 90



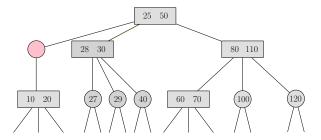
- ▶ The leaf has 1 key (underflow), which means it is a 2-node.
- ▶ all sibling nodes have 1 key (they are 2-nodes)
- ▶ fusion: e.g. delete of 1



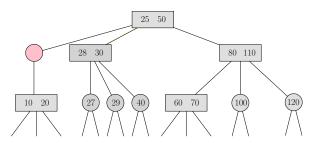
- ▶ The leaf has 1 key (underflow), which means it is a 2-node.
- ▶ all sibling nodes have 1 key (they are 2-nodes)
- ▶ fusion: e.g. delete of 1



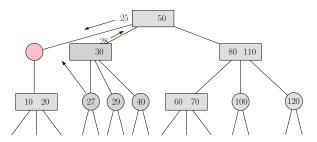
- ▶ The leaf has 1 key (underflow), which means it is a 2-node.
- ▶ all sibling nodes have 1 key (they are 2-nodes)
- **fusion**: e.g. delete of 1



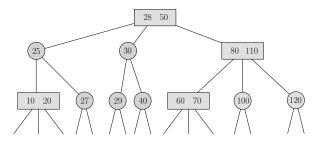
- Now the parent has a problem (underflow)
- repeat the same sequence, either balance or fusion based on the sibling nodes



- Now the parent has a problem (underflow)
- repeat the same sequence, either balance or fusion based on the sibling nodes



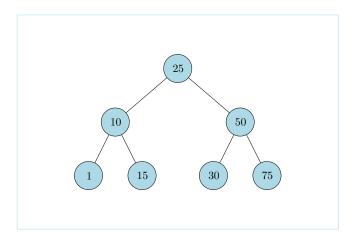
- Now the parent has a problem (underflow)
- repeat the same sequence, either balance or fusion based on the sibling nodes



- Fusion can cascade up until the root.
- In case it reaches the root, the tree's height is reduced by one.

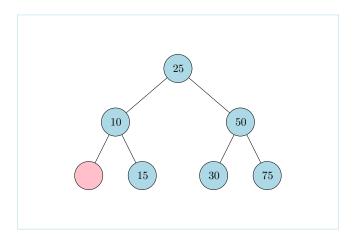
Deletion costs $\mathcal{O}(\log n)$ time.

Example of cascade up to the root



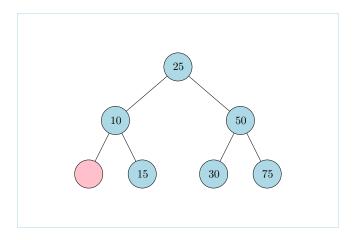
Delete key 1

Example of cascade up to the root



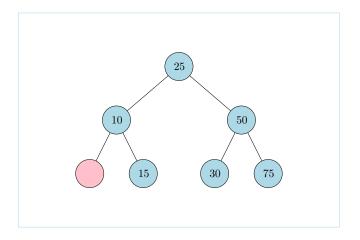
Delete key 1

Example of cascade up to the root



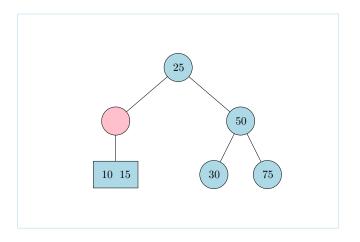
Node with underflow - test first balance, then fusion

Example of cascade up to the root



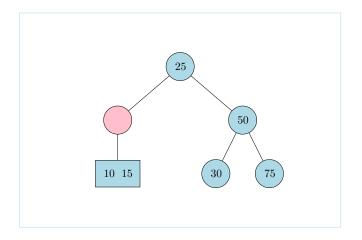
Balance not possible - siblings do not have spare keys

Example of cascade up to the root



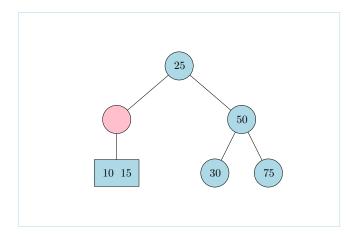
Fusion - use both keys from brother and parent

Example of cascade up to the root



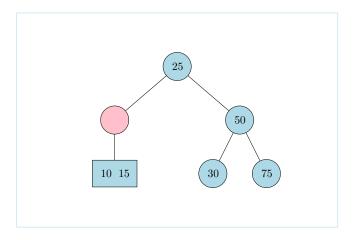
Node with underflow - one level up

Example of cascade up to the root



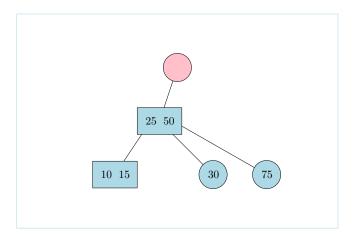
Node with underflow - test first balance, then fusion

Example of cascade up to the root



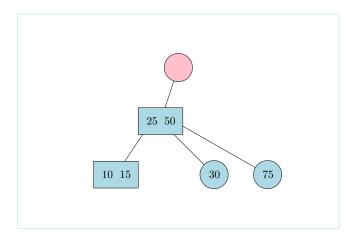
Balance not possible - siblings do not have spare keys

Example of cascade up to the root



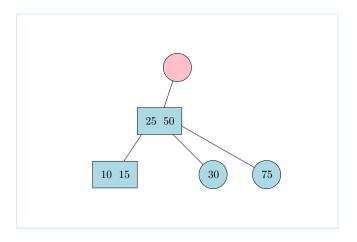
Fusion - use both keys from brother and parent

Example of cascade up to the root



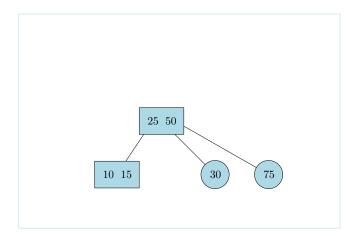
Node with underflow - one level up

Example of cascade up to the root



If the root has underflow - simply delete it

Example of cascade up to the root



If the root has underflow - simply delete it

Balanced 2-3-4 Search Trees

Positive:

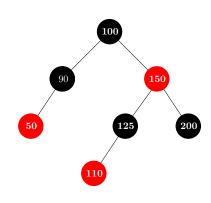
- ▶ simple insertion algorithm
- ightharpoonup complexity $\mathcal{O}(\log n)$ for insertion, deletion and search

negative:

- ▶ 3 different node types
- complex procedured due to multiple links, copying of links, etc.

A Red-Black tree is a binary search tree (BST) with the following additional properties:

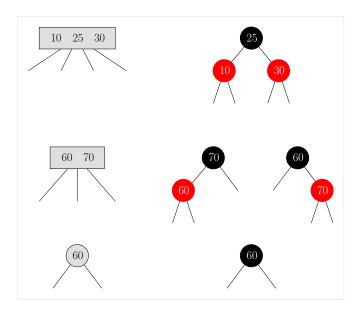
- Every node is either red or black
- ► The root is black
- Every red node does not have any red child
- Every path from an external node (null) to the root has to pass through the same number of black nodes

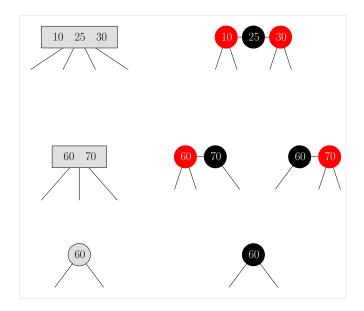


Theorem

A red-black tree with n nodes has height $O(\log n)$.

- ▶ Instead of a proof we will see a correspondence between read-black trees and 2-3-4 trees.
- ▶ We can view red-black trees as a representation of 2-3-4 trees.
- This correspondence allows us to easily show the properties of red-black trees.





Search, insert and delete

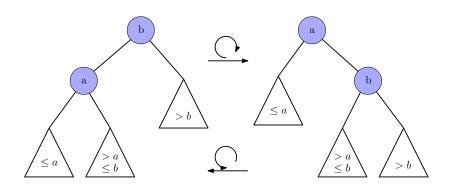
Search

Red-Black keys are binary search trees (BSTs)!

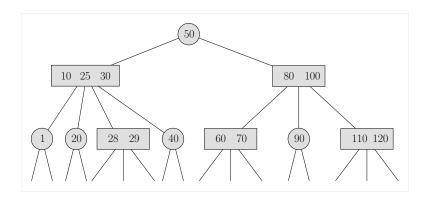
Insertion and Deletion

In correspondence with the 2-3-4 trees, describe the moves by changing colors and rotations.

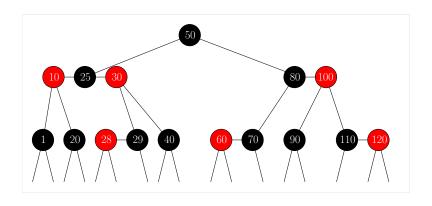
Rotations (on any BST)



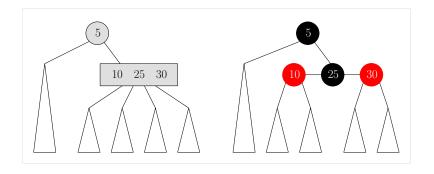
Example



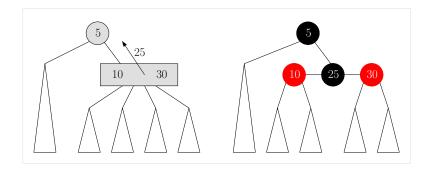
Example



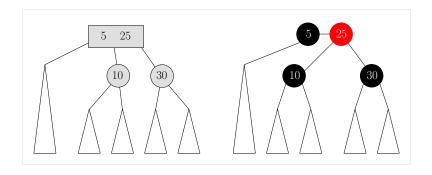
Rotation (first)

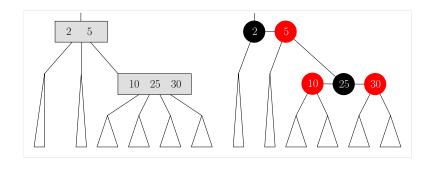


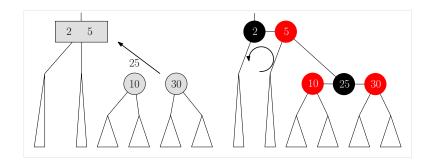
Rotation (first)

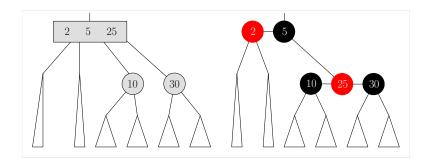


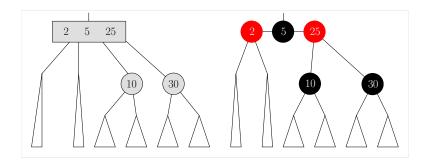
Rotation (first)











Rotations

There are more cases...

Insertion and Deletion Complexity

- ► There are a lot of complicated rules! All of them make sense if we view the corresponding 2-3-4 tree.
- ▶ Simulating the changes of a node of a 2-3-4 tree with the corresponding nodes of a red-black tree takes $\mathcal{O}(1)$ time.
- ▶ Since 2-3-4 trees support insertion in $\mathcal{O}(\log n)$ time, so do red-black trees.
- ▶ The same idea also works for deletions.

Height

Theorem

A red-black tree with n nodes has height $O(\log n)$.

Proof.

We can collect all red nodes to their parents in order to convert the tree to a 2-3-4 tree.

The height of the tree can at most lose half of its size.

The resulting 2-3-4 tree has height $\mathcal{O}(\log n)$, thus the original red-black tree has height at most $2 \cdot \mathcal{O}(\log n) = \mathcal{O}(\log n)$.

Other Balanced Trees

- ▶ AVL trees: $\mathcal{O}(\log n)$ height using rotations
- splay trees
- scapegoat trees
- etc.

The first balanced tree!

G.M. Adelson-Velskii and E.M. Landis. An algorithm for the organization of information, Proceedings of the USSR Academy of Sciences 146: 263–266, 1962.

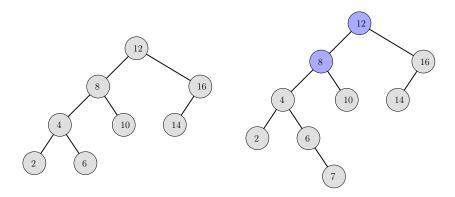
Named from the initials of its inventors.

- ► Binary Search Tree
- Additional balance property

The additional balance property needs to be maintained easily and to be able to preserve the height of the tree to $\mathcal{O}(\log n)$.

Definition

An AVL tree is a binary search tree T where for each node $v \in T$ the height of its two subtrees differ by at most 1. The height of an empty tree is defined as -1.



Left AVL tree, right non-AVL tree.

Height

Theorem

An AVL tree with n keys has height $\mathcal{O}(\log n)$.

Height

Theorem

An AVL tree with n keys has height $O(\log n)$.

Proof.

Let n(h) be the smallest number of nodes of an AVL tree with height h.



Theorem

An AVL tree with n keys has height $\mathcal{O}(\log n)$.

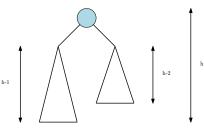
Proof.

Let n(h) be the smallest number of nodes of an AVL tree with height h.

We can easily show that n(1) = 1 and n(2) = 2.

For h>2 an AVL tree has a root and two subtrees, one AVL subtree with height h-1 and one AVL subtree with height h-2.

Thus
$$n(h) = 1 + n(h-1) + n(h-2)$$
.



Height

Theorem

An AVL tree with n keys has height $\mathcal{O}(\log n)$.

Proof.

Let n(h) be the smallest number of nodes of an AVL tree with height h.

Since n(h-1) > n(h-2) we get that

$$n(h) = 1 + n(h-1) + n(h-2) > 2n(h-2) > 4n(h-4) > 8n(h-6) > \dots$$

using induction

$$n(h) > 2^i n(h-2i)$$

Since n(1) = 1 we have that $n(h) > 2^{(h-1)/2}$ and by taking logarithms we have $h < 2 \log n(h) + 1$.

Insertion on an AVL tree

Insertion is performed like in any other BST but afterwards we might need to fix the tree.

We make sure we perform $\mathcal{O}(d)$ operations where d is the depth where insertion is performed.

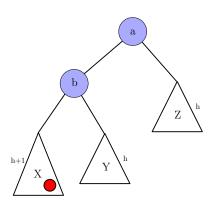
Insertion on an AVL tree

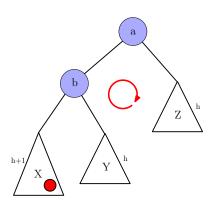
We fix the tree with rotations. Let *x* be the node with the largest depth that is non-balanced.

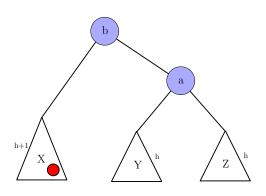
There are 4 cases based on whether the insertion happened at

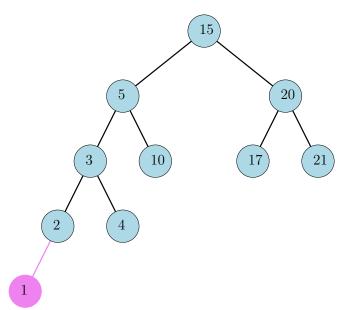
- 1. the left subtree of the left child of *x*
- 2. the right subtree of the left child of x
- 3. the left subtree of the right child of x
- 4. the right subtree of the right child of *x*

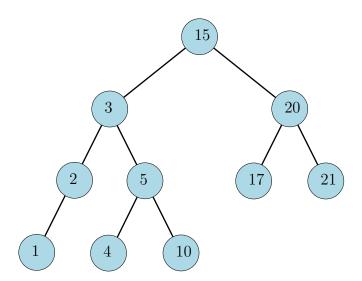
Cases 1&4 are resolved with a *single rotation* while cases 2&3 needs a *double rotation*.





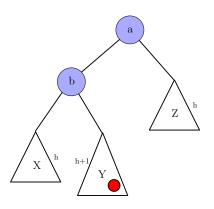






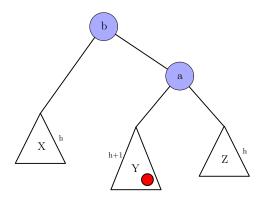
Other cases

In cases 2&3 the single rotation does not work.

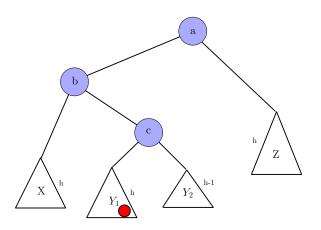


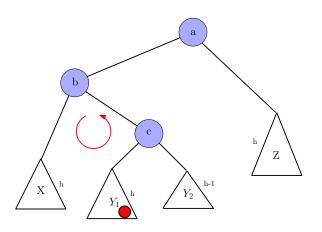
Other cases

In cases 2&3 the single rotation does not work.

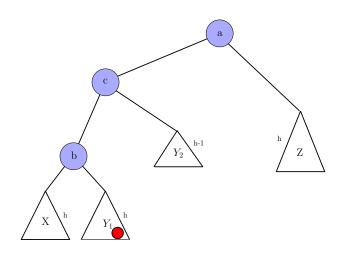




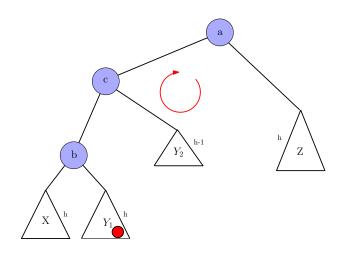




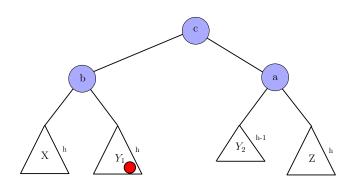
$$X < b < Y_1 < c < Y_2 < a < Z$$



$$X < b < Y_1 < c < Y_2 < a < Z$$



$$X < b < Y_1 < c < Y_2 < a < Z$$



$$X < b < Y_1 < c < Y_2 < a < Z$$

Deletion

Deleting elements from an AVL tree happens in the same way as in BSTs with additional rotations.

 $\mathcal{O}(\log n)$ in the worst case

Memory Hierarchy

(Year 2010)

A big dilemma exists between speed and size when it comes to memory.

SRAM (Static Random Access Memory)

- Registers are made using SRAM
- ▶ Very fast $(1ns = 10^{-9}sec$ handles GHz clock speeds)
- ▶ Very expensive $(1GB \approx 5000\$)$

DRAM (Dynamic Random Access Memory)

- Memory is made using this technology
- ► Fast (25ns)
- Less expensive $(16GB \approx 100\$)$

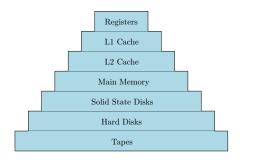
Hard Disk

- Very slow $(5ms = 5 \cdot 10^{-3}sec)$
- Very cheap

Memory Hierarchy

(Year 2010)

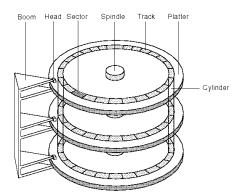
Idea: Use many different types of memory for better performance.



- 0.25 1ns
- 1 5ns
- 5 25ns
- 25 -100ns
- 100 150ns
- 2 $20\mathrm{ms}$
- sequential access

External Memory





- seek time: time in order for the heads to move to the right location (track)
- ► rotational delay: waiting time in order for the cylinder to come to the right sector that we are looking for

External Memory

- ▶ computer memory (DRAM) has access time of nanoseconds: 10^{-9} sec, e.g. 25 100 nsec
- ▶ hard disks have access time of milliseconds: 10^{-3} , e.g. 2 20 ms

Conclusion

It is not a good idea to read small amounts of data from the disk. For this reason disks always read/write a block which is in the order of Kilobytes, e.g. 512K.

B-Tree (Beta Tree)

- search tree which is efficient for storing to disk
- **Proof** generalizes the 2-3-4 trees using nodes with number of keys between t and 2t for any $t \ge 2$.

A *B-tree T* is a tree with a root and the following properties:

- 1. every node x has the following fields:
 - (a) n[x], the number of keys that x has
 - (b) the n[x] keys in order: $key_1[x] \le key_2[x] \le \ldots \le key_{n[x]}[x]$
 - (c) a value leaf[x] which is true iff node x is a leaf

A B-tree T is a tree with a root and the following properties:

- 1. every node x has the following fields:
 - (a) n[x], the number of keys that x has
 - (b) the n[x] keys in order: $key_1[x] \le key_2[x] \le ... \le key_{n[x]}[x]$
 - (c) a value leaf[x] which is true iff node x is a leaf
- 2. if x is an internal node it also contains n[x]+1 pointers $c_1[x], c_2[x], \ldots, c_{n[x]+1}[x]$ to its children. Leafs do not have children in which case these pointers are not defined.

A *B-tree T* is a tree with a root and the following properties:

- 1. every node x has the following fields:
 - (a) n[x], the number of keys that x has
 - (b) the n[x] keys in order: $key_1[x] \le key_2[x] \le \ldots \le key_{n[x]}[x]$
 - (c) a value leaf[x] which is true iff node x is a leaf
- 2. if x is an internal node it also contains n[x] + 1 pointers $c_1[x], c_2[x], \ldots, c_{n[x]+1}[x]$ to its children. Leafs do not have children in which case these pointers are not defined.
- 3. The keys $key_i[x]$ split the range of the keys which are stored in each subtree: if k_i is the key stored in subtree with root $c_i[x]$

$$k_1 \leq key_1[x] \leq k_2 \leq key_2[x] \leq \cdots \leq key_{n[x]}[x] \leq k_{n[x]+1}$$
.



4. Every leaf has the same depth, the height of the tree h.

- 4. Every leaf has the same depth, the height of the tree h.
- 5. Let $t \ge 2$ be the **minimum order** of the tree:
 - 5.1 Every node besides the root must have at least t-1 keys (which translates to at least t children). If the tree is not null, the root must have at least one key.
 - 5.2 Every node can contain at most 2t-1 keys, e.g. at most 2t children. A node is called **full** if it contains exactly 2t-1 keys.

- 4. Every leaf has the same depth, the height of the tree h.
- 5. Let t > 2 be the **minimum order** of the tree:
 - 5.1 Every node besides the root must have at least t-1 keys (which translates to at least t children). If the tree is not null, the root must have at least one key.
 - 5.2 Every node can contain at most 2t-1 keys, e.g. at most 2t children. A node is called **full** if it contains exactly 2t-1 keys.

The simplest B-tree is for t = 2. Every internal node has 2, 3 or 4 children, thus, is the 2-3-4 tree.

In practice we use much larger values for t.

B-tree Height

Theorem

A B-tree with $n \ge 1$ keys and minimum order $t \ge 2$ has height

$$h \leq log_t \frac{n+1}{2}$$
.

Proof.

A B-tree with height h has the least number of nodes if the root has one key and the rest have t-1 keys.

At lever 0 there is one node. At level 1 there are 2 nodes. At level 2 there are 2t nodes, at level $32t^2$ nodes, etc.



B-tree Height

Theorem

A B-tree with $n \ge 1$ keys and minimum order $t \ge 2$ has height

$$h \leq log_t \frac{n+1}{2}.$$

Proof.

The number of keys are:

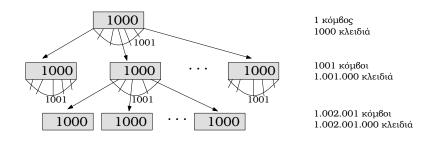
$$n \ge 1 + (t-1) \sum_{i=1}^{h} 2t^{i-1}$$

$$= 1 + 2(t-1) \left(\frac{t^h - 1}{t-1}\right)$$

$$= 2t^h - 1$$

Disk Accesses

Making sure that each node fits exactly on a disk block, we can use disk accesses equal to the height of the tree.



The tree has height $\mathcal{O}(\log_t n)$.

Basic Principles when Implementing B-trees

- (i) The root of a b-tree always remains in memory
- (ii) We need two functions which read and write nodes to disk
 - ► DISK-READ(x)
 - ► DISK-WRITE(x)

Search on a B-tree

Search works just like in BSTs except now we have more choices per node.

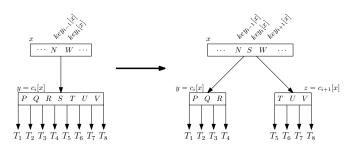
```
B - TREE - SEARCH(x, k)
i = 1
while i \le n[x] and k > key_i[x] do
   i = i + 1
end
if i \le n[x] and k = key_i[x] then
   return (x, i)
end
if leaf[x] then
   return nil
else
   DISK - READ(c_i[x])
end
return B - TREE - SEARCH(c_i[x], k)
```

Insertion in a B-tree

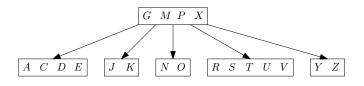
- While searching for the insertion location, we split full nodes (having 2t-1 keys) into two nodes
- ▶ When a node splits, the middle key goes up the tree.
- This makes enough room for possible splits at lower levels of the tree.

Node split

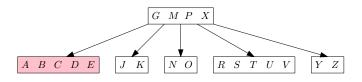
t = 4



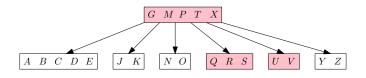
t = 3



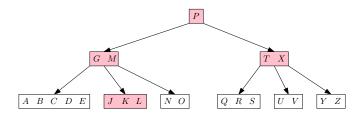
t = 3, insertion of B



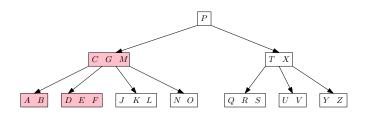
t = 3, insertion of Q



t = 3, insertion of L



t = 3, insertion of F



Deletion from a B-tree

Works the same as in 2-3-4 trees.

We may need to either

- balance, or
- ▶ fuse

Reading and Sources

- ➤ Sections 7.1, 7.2, 7.3

 Kurt Mehlhorn Peter Sanders. Algorithms and Data Structures,
 The basic toolbox, 1/e, 2014.
- ➤ Sections 12.5, 12.6, 12.8, 12.9, 13.3, 13.4, 16.3

 Robert Sedgewick. Algorithms in C: Fundamentals, Data Structures, Sorting, Searching. 3/e, Addison-Wesley Professional, 1997.
- ➤ Sections 12.1, 12.2, 12.3, 13.1, 13.2, 13.3, 13.4 Cormen, Leiserson, Rivest and Stein. Introduction to Algorithms. 3/e, MIT Press. 2009.