

Binary Search Trees

Advanced Algorithms and Data Structures - Lecture 3A

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Dynamic Sets

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Dictionaries

In practice elements of a dynamic sets will be pairs:

A **key** used for searching, a **value** to be returned

Such a dynamic set is also called a **dictionary**

Example: In a database of students, the key could be the ID number, the value the name of the student (and all other relevant data)

$$\{(7, \text{Monica}), (5, \text{Richard}), (1, \text{Fang}), (10, \text{Wei}), \\ (4, \text{Jan}), (6, \text{Femke}), (9, \text{Clara})\}$$

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For the study of the algorithms just the keys are relevant

I will describe the algorithms just using a set of keys

Exercise: Extend them to include the values

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Fast search in $O(\log n)$ time

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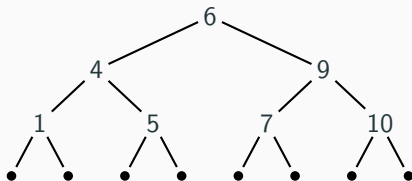
Fast search in $O(\log n)$ time

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Can we find a data structure for which all three operations are efficient?

BINARY SEARCH TREES



The operations of **search**, **insert**, **delete**

can be done in $O(k)$ time

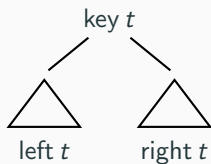
where k is the **depth of the tree**

But tricky to keep k small: $k \sim O(\log n)$

There are more advanced variants that guarantee this: **Red-Black Trees**

The BST Property

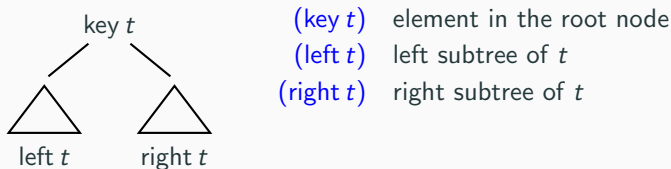
For every tree t , we use the notation:



(key t) element in the root node
(left t) left subtree of t
(right t) right subtree of t

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The **defining property of Binary Search Trees** is:

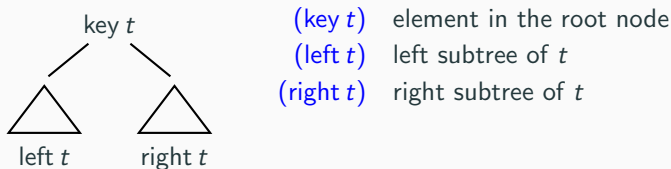
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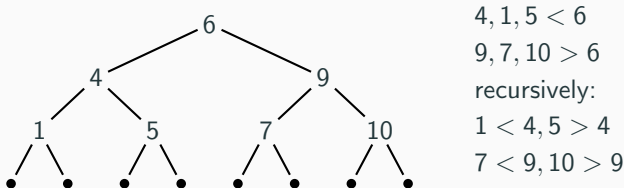


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Trees with key-value pairs

In practical applications:

nodes will contain **pairs** $\langle k, v \rangle$ of
a **key** k and a **value** v

The key is used for searching

The value is the information we want to store

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For example

- In a dictionary: **keys** = words; **values** = definitions
- In an address book: **keys** = names; **values** = addresses

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For the sake of the definition of the algorithms, we only use keys

Exercise: modify the algorithms with key-value pairs

The type of BSTs

BST is a type with two constructors:

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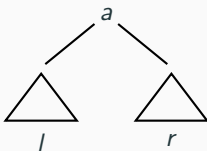
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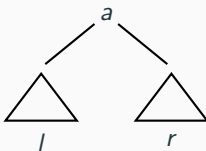
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- Node for internal nodes
containing elements from an ordered type Key
(Node $a \ / \ r$) is the tree



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In functional programming it is defined as this inductive data type

```
data BST = Nil | Node Key BST BST
```


Component Functions

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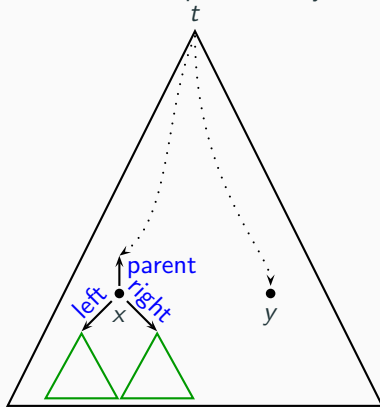
Imperative Implementation

In imperative programming, we use pointers to nodes, with methods giving the two children and the parent (See Ch.12 of IA)

(We can do it in functional programming too: using paths or defining them directly with those fields (Exercise))

Global Objects, Local Pointers

Work with a global tree t and with pointers x , y to subtrees/nodes



We can move around the tree with operations on pointers

- $(\text{parent } x)$ the immediate precursor of x ; Nil if x is the root
- $(\text{left } x)$ and $(\text{right } x)$ the children of x ; Nil if they are leaves

Searching a tree t for a key k is done by following a single path

At each node x :

- If $k = \text{key } x$, we have found it!
- If $k < \text{key } x$, go to the left child of x ;
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Searching

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Functional/Recursive Version:

```
search :: Key → BST → Bool
search k Nil = False
search k (Node x l r) =
    (k == x) || if (k < x) then search k l
                else search k r
```


Imperative/Iterative Version

```
search (t,k):  
  x := t  
  while x /= Nil and k /= (key x)  
    if k < (key x) then x := (left x)  
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  return k == (key x)
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This algorithm just returns a Boolean value:

true if k is present in the tree, **false** otherwise

Exercise: Modify the algorithm for trees containing key-values pairs;
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Exercise: Modify the algorithm for trees containing key-values pairs;
if the key is found, it must return the corresponding value.

Complexity: The search algorithm starts at the root and follows a specific path, until it reaches a node that matches the search key or a leaf. The time complexity is $O(h)$ where h is the height of the tree.

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Exercise: Implement the insert operation
Hint: Insert in a leaf in the correct position

Minimum and Maximum

The **minimum element** in a binary search tree is the **leftmost** one, the **maximum is the rightmost**

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minimum :: BST → Maybe Key
minimum Nil = Nothing
minimum (Node x Nil _) = Just x
minimum (Node x l _) = minimum l
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In iterative style:

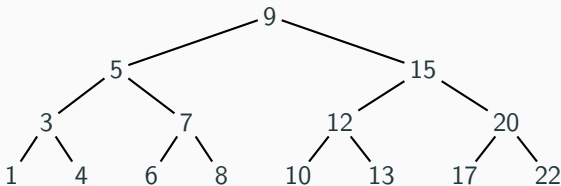
```
minimum (x):
  while (left x) /= Nil
    x := (left x)
  return (key x)
```

The maximum is defined similarly, going right instead of left.

Complexity: $O(h)$ where h is the height of the tree.

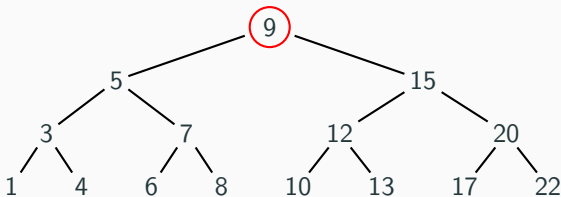
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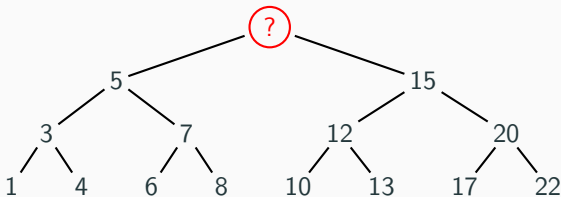
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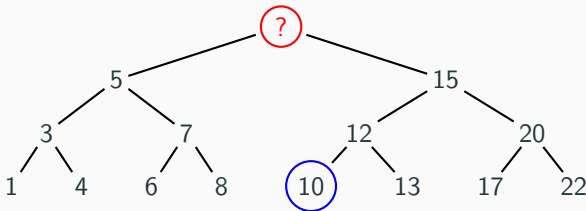


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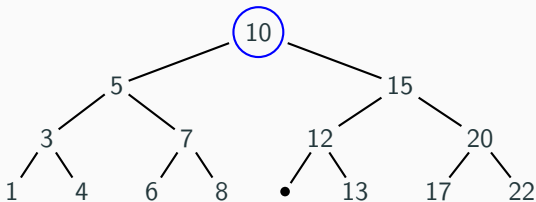
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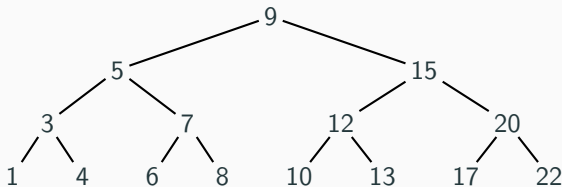
- Remove the root
- Find the minimum of the right child
- Place it at the root

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Predecessor and Successor

Extra Operations:

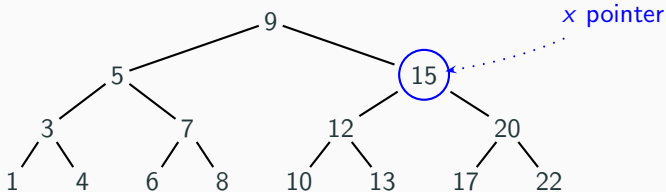
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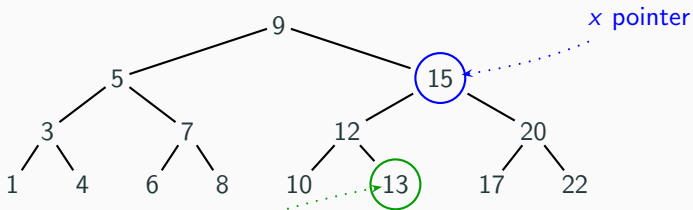
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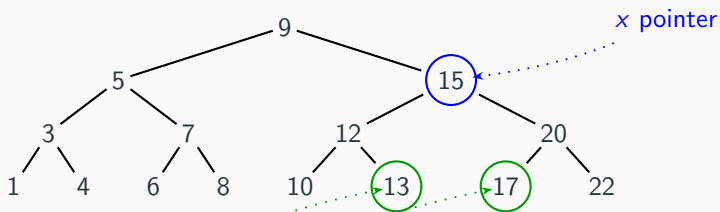


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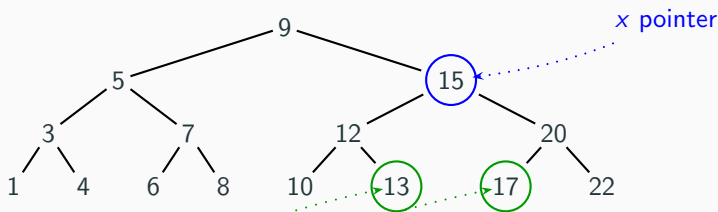


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Exercise: What if the node doesn't have a left child (or a right child)?