

# Algorithms and Data Structures

## (Search) Trees

Ulf Leser



Source: [whmsoft.net/](http://whmsoft.net/)

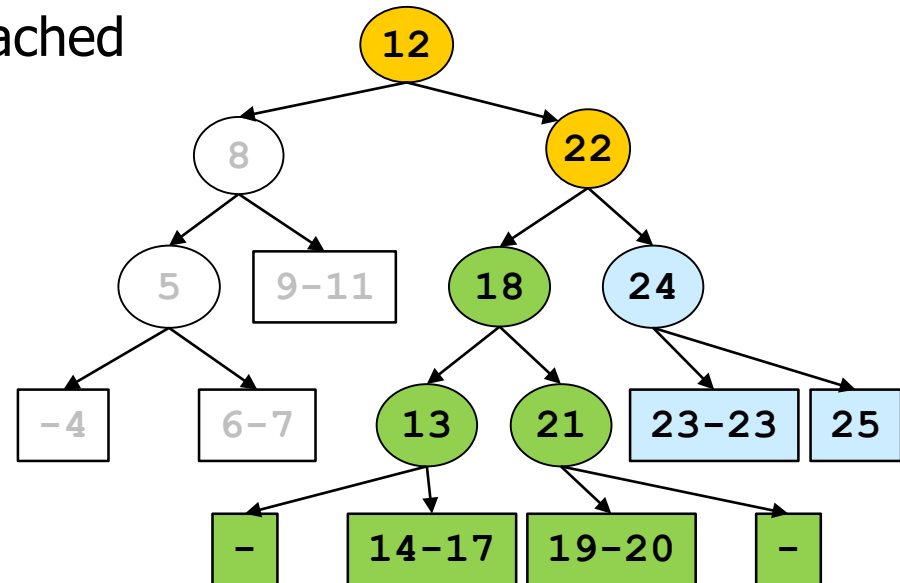
# Content of this Lecture

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- Trees
- Search Trees
- Natural Trees

# Motivation

- In a list, (almost) every element has one predecessor / successor
- In a tree, (almost) every element has one predecessor but **many successors**
- Elements create **partitions of the set of all elements**
  - Every node in a tree can be reached by **only one path** from root
    - I.e., path  $\sim$  element
  - Partitions: All nodes with **the same path prefix**
  - Prominent **semantic split** criterion: Order
    - Lower - left subtree,
    - Higher - right subtree



# Trees are everywhere in computer science

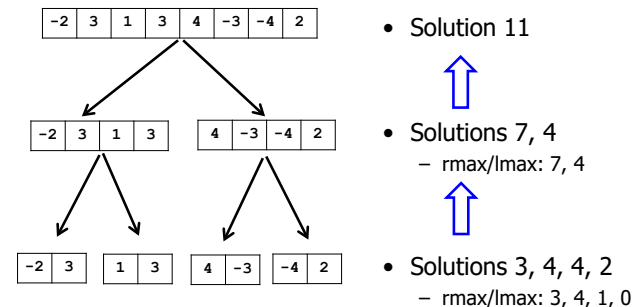
- **Divide-and-conquer** partitions

- Max-subarray
- Merge-Sort
- QuickSort
- ...

- XML

- depth-first vs breadth-first traversal

## Example

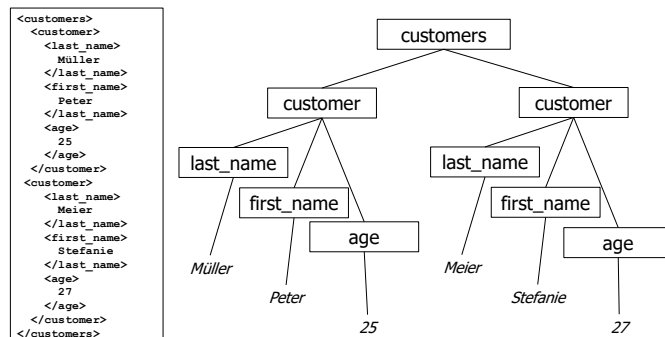


Ulf Leser: Alg&DS, Summer semester 2011

22

## Data – A Tree

- The data items of an XML database form a tree



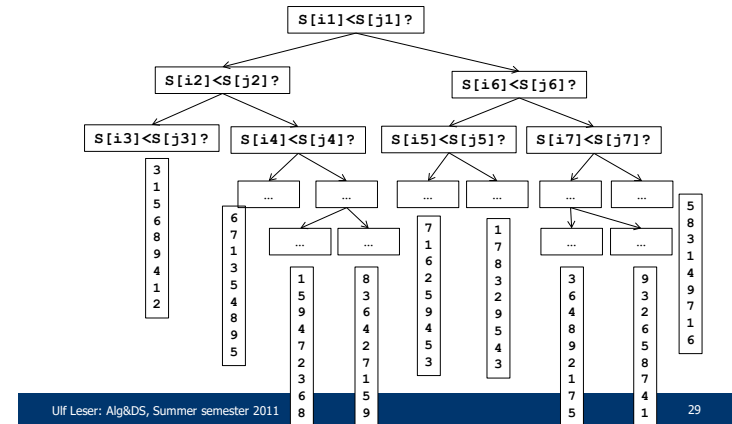
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10

# Already Seen

- Decision trees for proving the lower bound for sorting

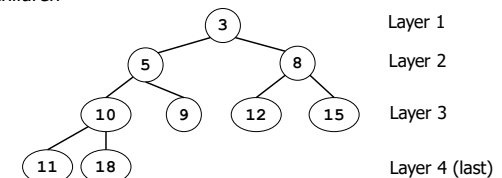
## Full Decision Tree



- Heaps for priority queues

## Heaps

- Definition  
A *heap* is a labeled binary tree for which the following holds
  - Form-constraint (FC): The tree is complete except the last layer
    - I.e.: Every node has exactly two children
  - Heap-constraint (HC): The value of any node is smaller than that of its children



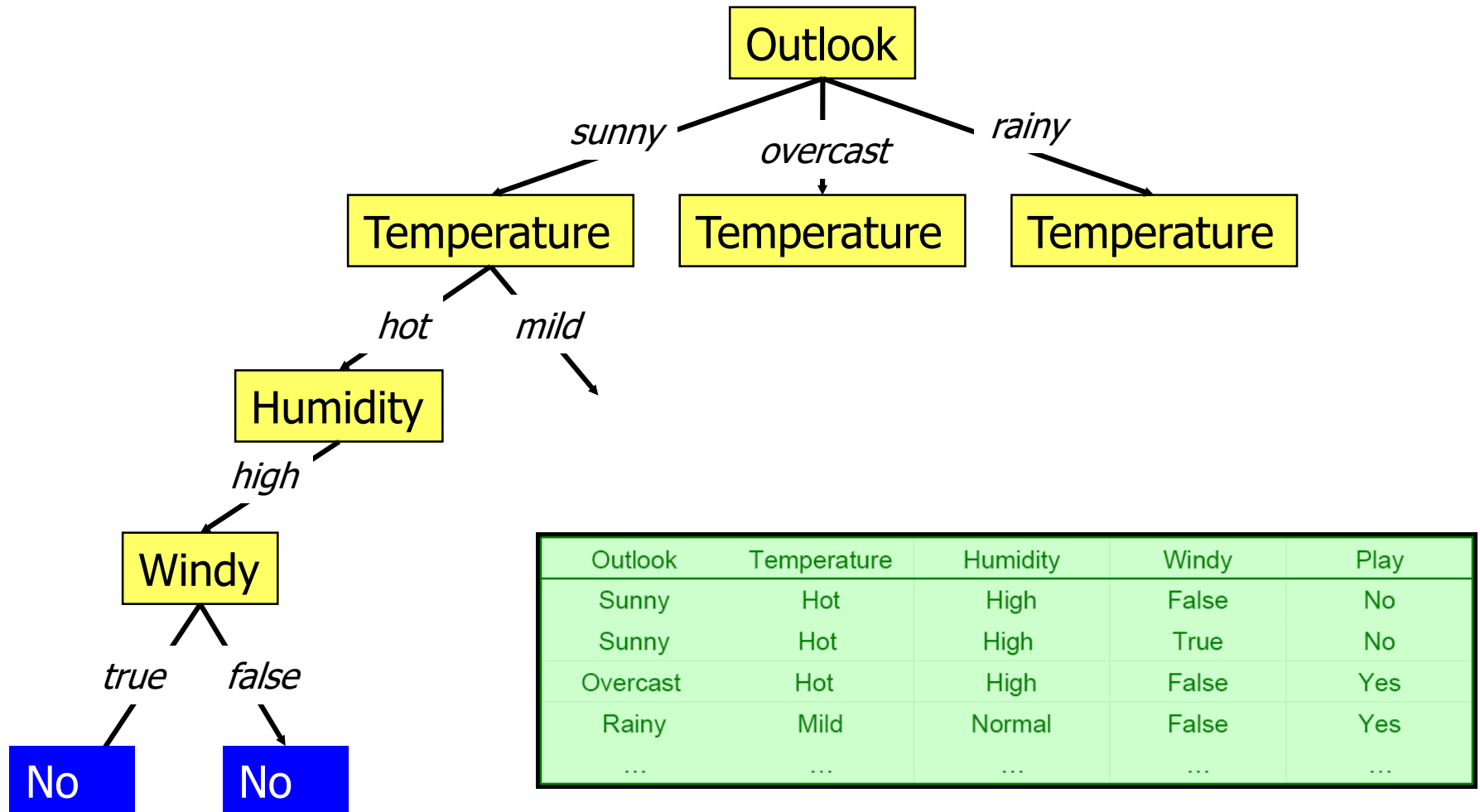
# Machine Learning

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- Want to go to a football game?
- Might be canceled – depends on the whether
- Let's **learn from examples**

Outlook	Temperature	Humidity	Windy	Play
Sunny	Hot	High	False	No
Sunny	Hot	High	True	No
Overcast	Hot	High	False	Yes
Rainy	Mild	Normal	False	Yes
...	...	...	...	...

# Decision Trees



# Many Applications

The decision tree partitions the set of all possible situations based on predefined characteristics (attributes)

Challenge: Which tree leads to the best decisions as soon as possible?

Source: Am J Transplant © 2004 Blackwell

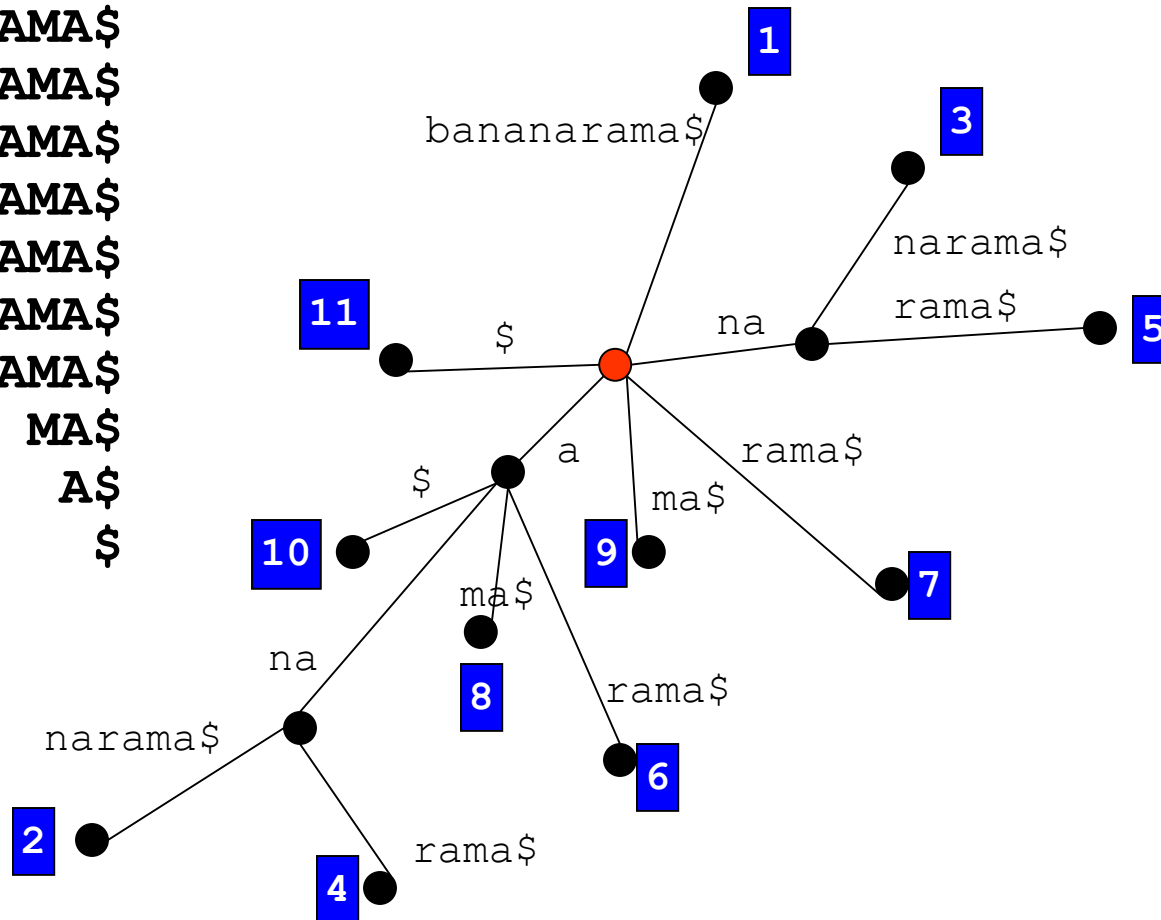
# Suffix-Trees

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- Recall the problem to find all occurrences of a (short) string  $P$  in a (long) string  $T$
- Fastest way ( $O(|P|)$ ): Suffix Trees
  - Look at all suffixes of  $T$  (there are  $|T|$  many)
  - Construct a tree
    - Every edge is labeled with a letter from  $T$
    - All edges emitting from a node are labeled differently
    - Every path from root to a leaf is uniquely labeled
    - All suffixes of  $T$  are represented as leaves
- Every occurrence of  $P$  must be the prefix of a suffix of  $T$
- Thus, every occurrence of  $P$  must map to a path starting at the root of the suffix tree

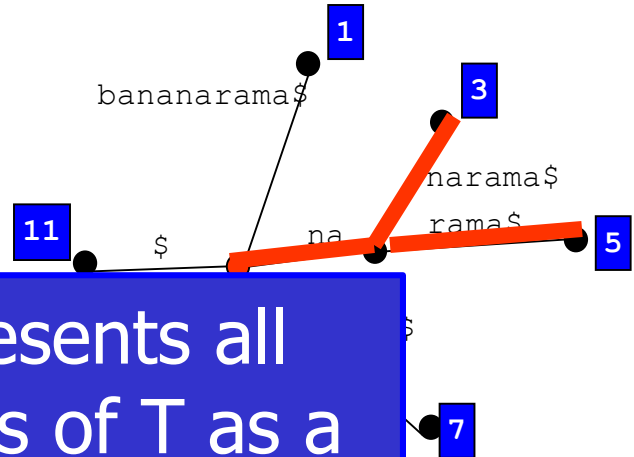
# Example

12345678901  
BANANARAMA\$  
ANANARAMA\$  
NANARAMA\$  
ANARAMA\$  
NARAMA\$  
ARAMA\$  
RAMA\$  
AMA\$  
MA\$  
A\$  
\$



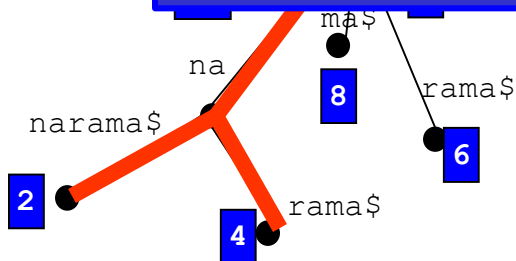
# Searching in the Suffix Tree

$P = \text{„na“}$



The suffix tree for T represents all common prefixes of suffixes of T as a unique path from root.

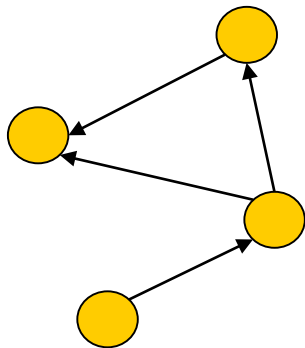
Challenge: Construction of a suffix tree in linear time.



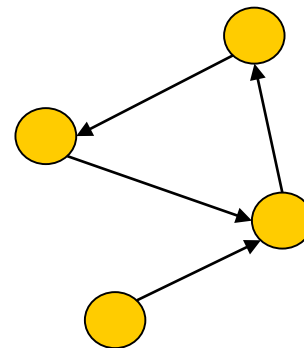
$P = \text{„an“}$

# Not Trees

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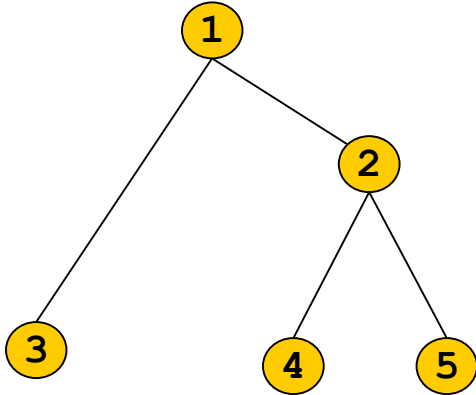
DAG: Directed,  
acyclic graph



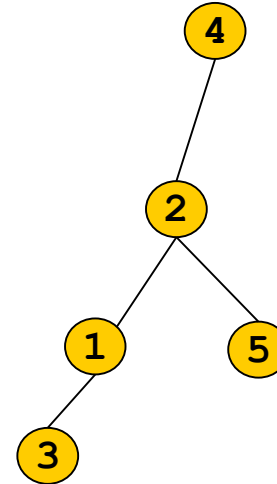
General  
(directed) graph

# Directed?

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We sometimes draw undirected edges with root at the top and **assume directed edges** from root to leaves



This visual aid is necessary!  
Otherwise, root and leaves are **indistinguishable**

# Graphs

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- Definition

A *graph*  $G=(V, E)$  consists of a set  $V$  of vertices (nodes) and a set  $E$  of edges ( $E \subseteq V \times V$ ).

- A sequence of edges  $e_1, e_2, \dots, e_n$  is called a *path* iff  $\forall 1 \leq i < n-1: e_i = (v_i, v_{i+1})$  and  $e_{i+1} = (v_{i+1}, v_{i+2})$
- The *length of a path*  $e_1, e_2, \dots, e_n$  is  $n$
- A path  $(v_1, v_2), (v_2, v_3), \dots, (v_{n-1}, v_n)$  is *acyclic* iff all  $v_i$  are different
- $G$  is *undirected*, if  $\forall (v, v') \in E \Rightarrow (v', v) \in E$ . Otherwise  $G$  is *directed*
- $G$  is *connected* if every pair  $v_i, v_j$  is connected by at least one path
- $G$  is *acyclic* if it contains no cyclic path

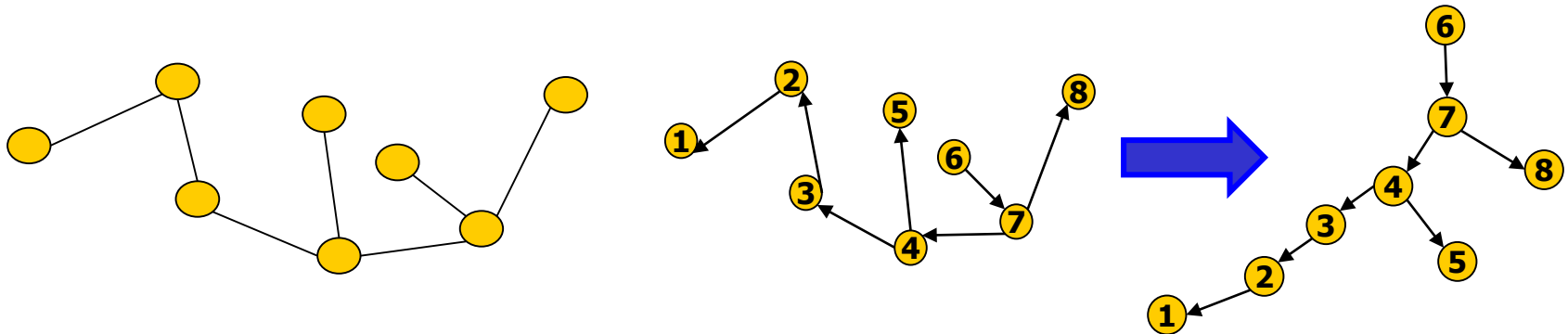
Let  $G=(V, E)$  be a directed graph and let  $v, v' \in V$ .

- Every edge  $(v, v') \in E$  is called *outgoing for*  $v$
- Every edge  $(v', v) \in E$  is called *incoming for*  $v$

# Trees as Connected Graphs

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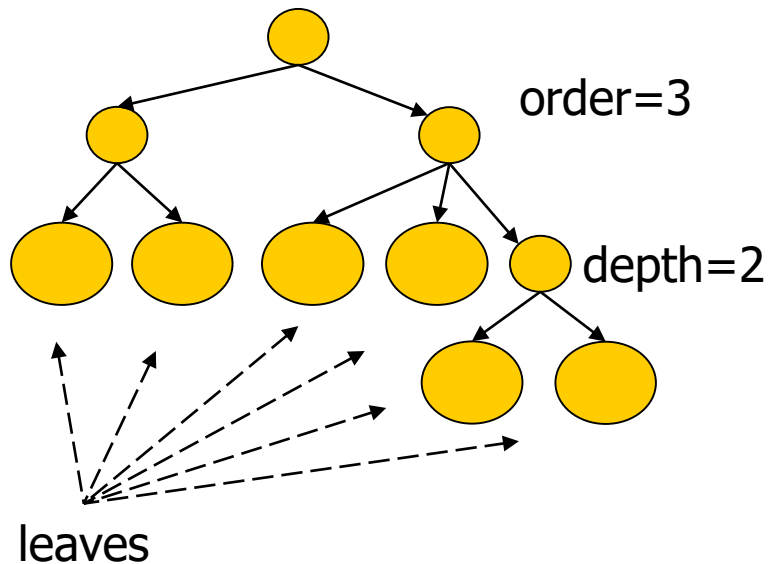
- Definition
  - A undirected connected acyclic graph is called a *undirected tree*
  - A directed acyclic graph in which all but one vertex have in-degree 1 and one vertex has in-degree 0 (the root) and there is a path from this node to every other node is *called a directed rooted tree*
- From now on: “Tree” means “rooted directed tree”
- Lemma
  - In a tree, there exists exactly one path between root and any other node



# Terminology

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height=3



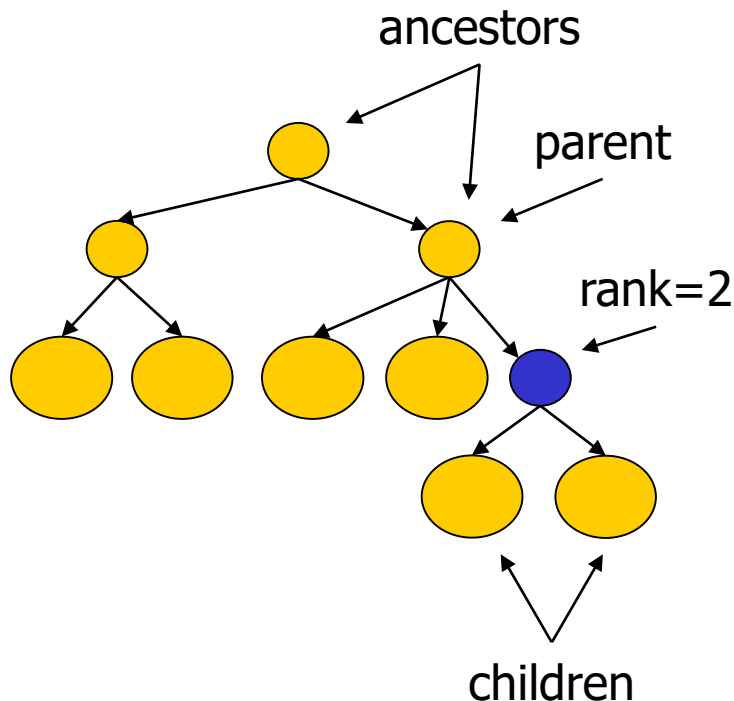
- Definition

*Let  $T$  be a tree. Then ...*

- *A node with no outgoing edge is a **leaf**; other nodes are **inner nodes***
- *The **depth of a node**  $p$  is the length of the path from root to  $p$*
- *The **height of  $T$**  is the depth of its deepest leaf*
- *The **order of  $T$**  is the maximal number of children of its nodes*
- *"Level  $i$ " are all nodes at depth  $i$*
- ***$T$  is ordered** if the children of inner nodes are ordered*

# More Terminology

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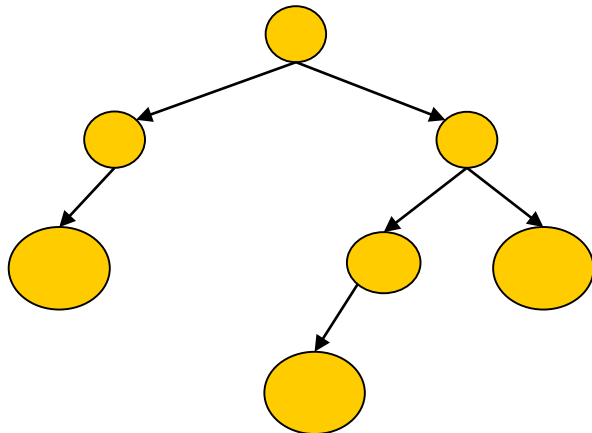
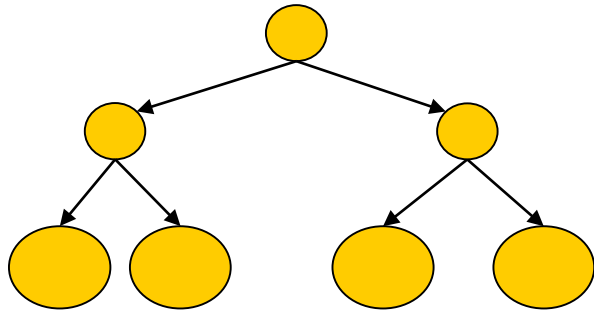
- Definition

*Let  $T$  be a tree and  $v$  a node.*

- *All nodes adjacent to an outgoing edge of  $v$  are  $v$ 's **children***
- *$v$  is called the **parent** of all its children*
- *All nodes on the path from root to  $v$  without  $v$  are the **ancestors of  $v$***
- *All nodes reachable from  $v$  are **its successors***
- *The **rank of a node  $v$**  is the number of its children*

# Two More Concepts

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- Definition  
*Let  $T$  be a directed tree of order  $k$ .  $T$  is **complete** if all its inner nodes have rank  $k$  and all leaves have the same depth*
- In this lecture, we will mostly consider rooted ordered trees of order two (**binary trees**)

# Recursive Definition of Trees

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- We often traverse trees using recursive functions

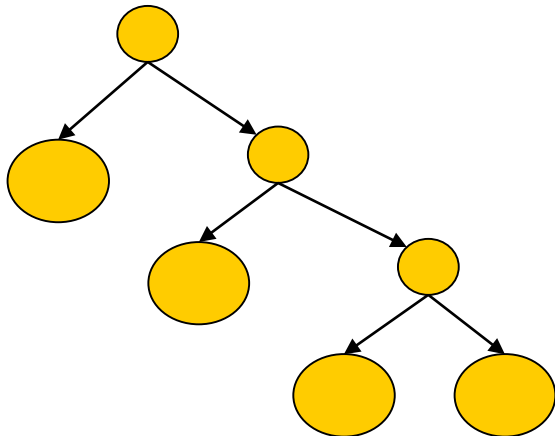
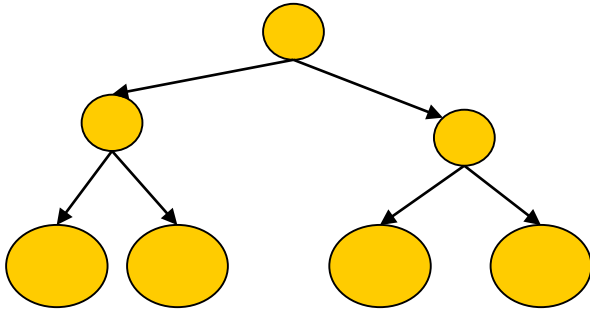
- Definition

*A (binary) tree is a structure defined as follows:*

- *A **single node** is a tree with height 0*
- *If  $T_1$  and  $T_2$  are trees, then the structure formed by a **new node  $v$**  and edges from  $v$  to the root of  $T_1$  and from  $v$  to the root of  $T_2$  is a tree*
  - *$v$  is its root*
  - *The height of this tree is  $\max(\text{height}(T_1), \text{height}(T_2))+1$ ;*
- *If  $T_1$  is a tree, then the structure formed by a **new node  $v$  and an edge from  $v$  to the root of  $T_1$**  is a tree*
  - *$v$  is its root*
  - *The height of this tree is  $\text{height}(T_1)+1$ ;*

# Some Properties (without proofs)

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- Lemma  
*Let  $T=(V, E)$  be a tree of order  $k$ .  
Then*
  - $|V|=|E|+1$
  - *If  $T$  is complete,  $T$  has  $k^{\text{height}(T)}$  leaves*
  - *If  $T$  is a complete binary tree,  $T$  has  $2^{\text{height}(T)+1}-1$  nodes*
  - *If  $T$  is a binary tree with  $n$  leaves,  $\text{height}(T) \in [\text{floor}(\log(n)), n-1]$*

# Content of this Lecture

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- Trees
- Search Trees
  - Definition
  - Searching
  - Inserting
  - Deleting
- Natural Trees

# Search Trees

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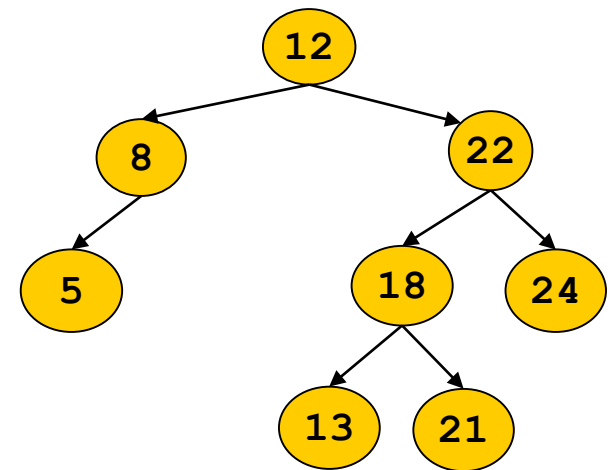
- Definition

A *search tree*  $T=(V,E)$  for a set of  $n$  *unique keys* is a *labeled binary tree* with  $|V|=n$  and

- $label(v) > \max(label(left\_child(v)), label(successors(left\_child(v))))$
- $label(v) < \min(label(right\_child(v)), label(successors(right\_child(v))))$

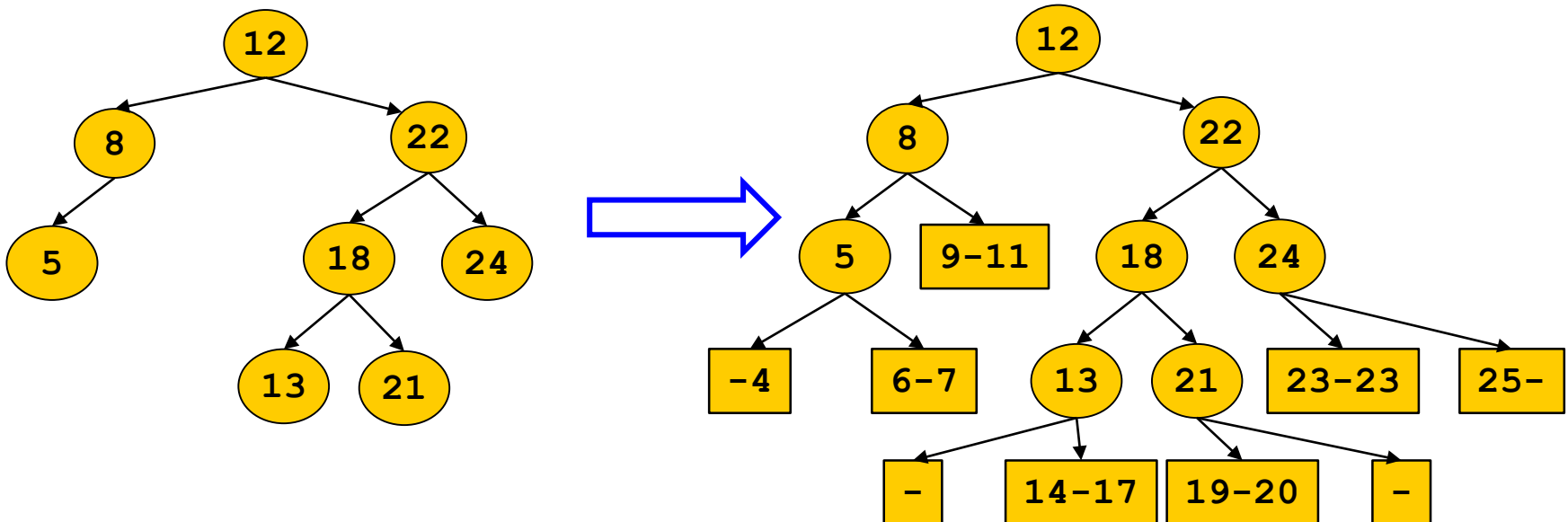
- Remarks

- For simplicity, we use integer labels
- “node”  $\sim$  “label of a node”
- We only consider search trees *without duplicate* keys (easy to change)
- Search trees are used to manage and search a list of keys
- Operations: *search, insert, delete*



# Complete Trees

- Conceptually, we sometimes **pad search trees** to full rank in all nodes
  - “padded” leaves are usually neither drawn nor implemented (NULL)
- A “padded” leaf represents the interval of values that **would be** below this node



# What For?

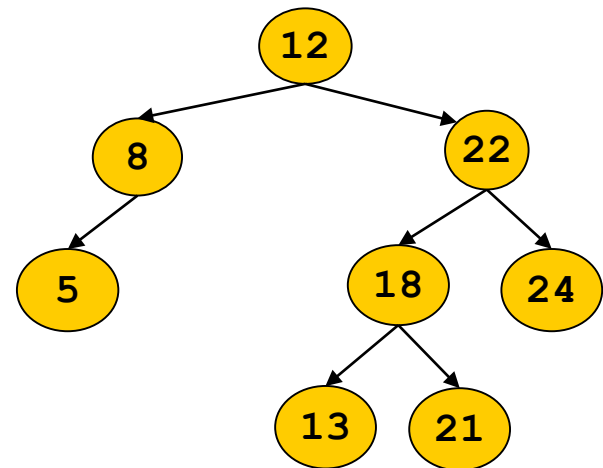
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- For a search tree  $T=(V,E)$ , we eventually will reach  $O(\log(|V|))$  for searching, inserting and deleting a key  $k$ 
  - First: Average Case of natural trees
  - Next: Worst Case for AVL-Trees
- Compared to binsearch on arrays, search trees are a dynamically growing / shrinking data structure
  - But need to store pointers
  - Complete trees can be easily managed in arrays

# Searching

- Searching a key  $k$ 
  - Comparing  $k$  to a node determines whether we have to look further down the **left** or the **right subtree**
    - We stop if  $\text{label}(\text{node}) = k$
  - If there is no child left,  $k \notin T$
- Complexity
  - In the worst case we need to traverse the **longest path** in  $T$  to show  $k \notin T$
  - Thus:  **$O(|V|)$**
  - Wait a bit ...

```
func node search( T search_tree,  
                  k integer) {  
    v := root(T);  
    while v!=null do  
        if label(v)>k then  
            v := v.left_child();  
        else if label(v)<k then  
            v := v.right_child();  
        else  
            return v;  
        end while;  
    return null;  
}
```



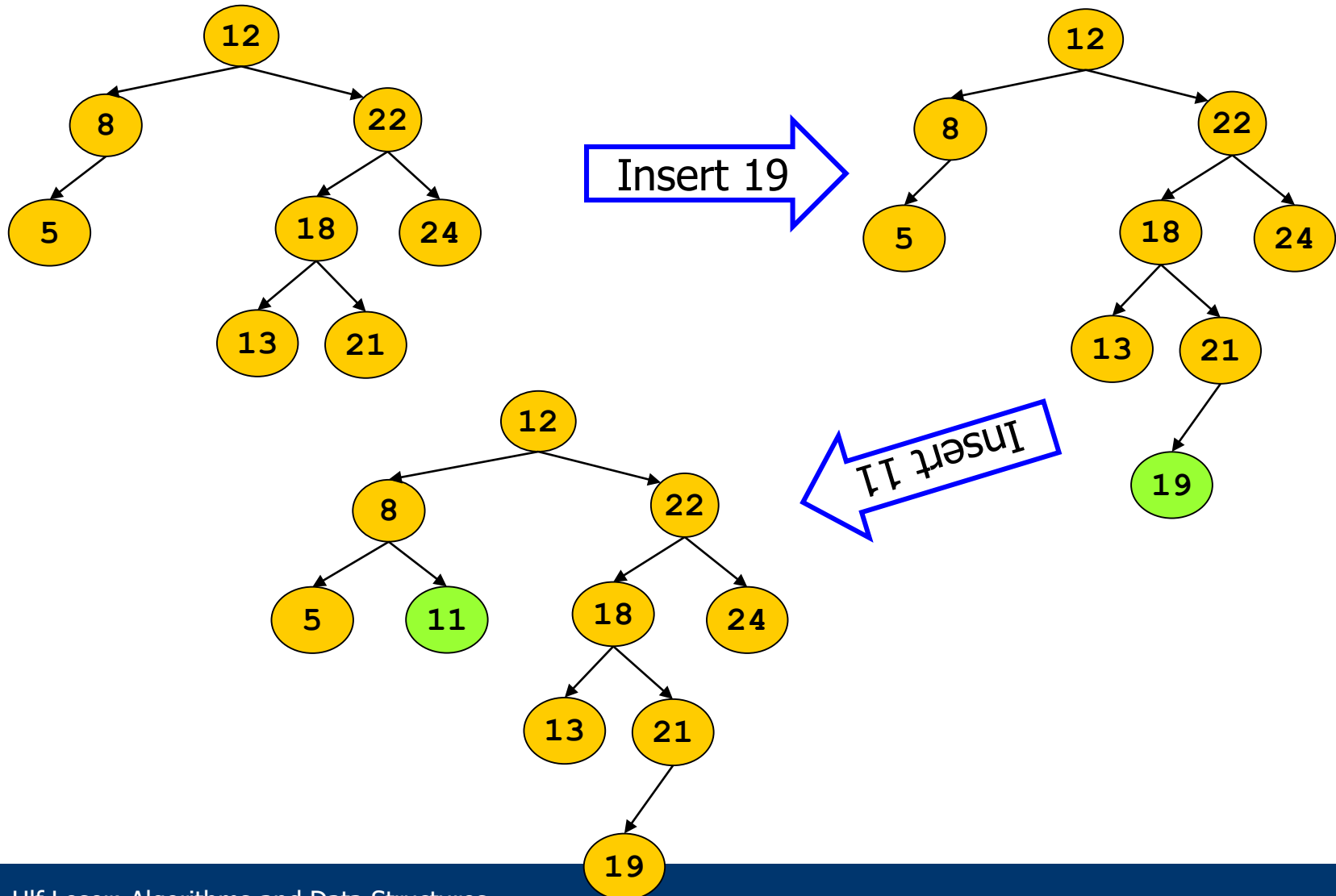
# Insertion

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```
func bool insert( T search_tree,
                  k integer) {
    v := root(T);
    while v!=null do
        p := v;
        if label(v)>k then
            v := v.left_child();
        else if label(v)<k then
            v := v.right_child();
        else
            return false;
    end while;
    if label(p)>k then
        p.left_child := new node(k);
    else
        p.right_child := new node(k);
    end if;
    return true;
}
```

- First search the new key  $k$ 
  - If  $k \in T$ , we do nothing
  - If  $k \notin T$ , the search must finish at a **null pointer** in a node  $p$ 
    - A “right pointer” if  $\text{label}(p) < k$ , otherwise a “left pointer”
- We replace the null with a pointer to a new node  $k$
- Complexity: Same as search

# Example



# Deletion

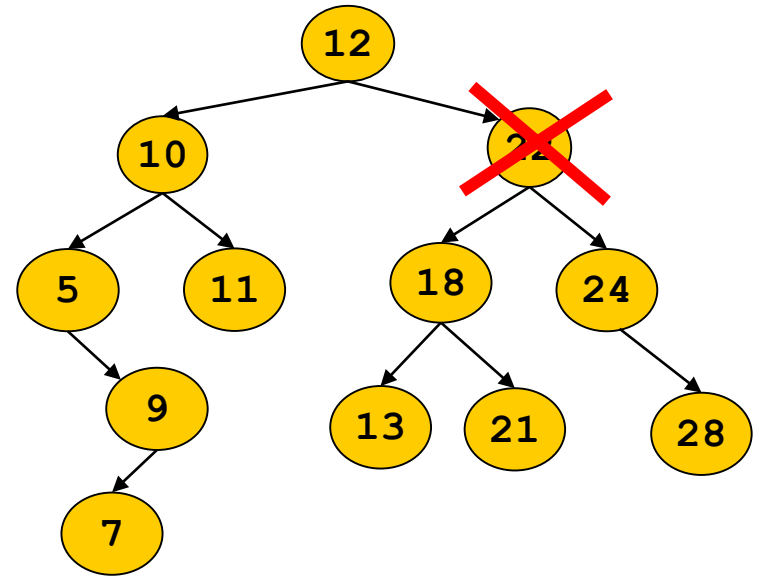
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- Again, we first search  $k$
- If  $k \notin T$ , we are done
- Assume  $k \in T$ . The following situations are possible
  - $k$  is **stored in a leaf**. Then simply remove this leaf
  - $k$  is stored in an inner node  $q$  with **only one child**. Then remove  $q$  and connect  $\text{parent}(q)$  to  $\text{child}(q)$
  - $k$  is stored in an inner node  $q$  with **two children**. Then ...

# Observations

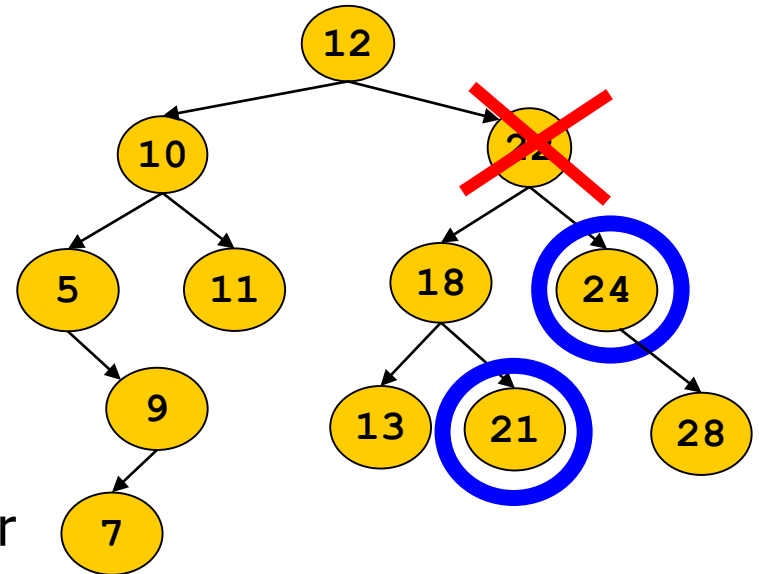
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- We cannot remove  $q$ , but we can **replace the label of  $q$**  with another label - and remove this node
- We need a node  $q'$  which can be removed and whose **label  $k'$  can replace  $k$**  without hurting the **search tree constraints**
  - $\text{label}(k') > \max(\text{label}(\text{left\_child}(k')), \text{label}(\text{successors}(\text{left\_child}(k'))))$
  - $\text{label}(k') < \min(\text{label}(\text{right\_child}(k')), \text{label}(\text{successors}(\text{right\_child}(k'))))$



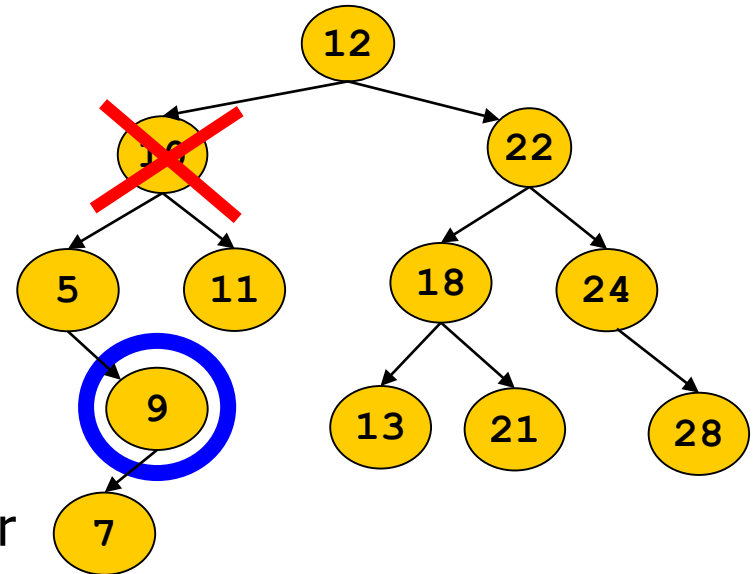
# Observations

- Two candidates
  - Largest value in the left subtree  
(**symmetric predecessor** of k)
  - Smallest value in the right subtree  
(**symmetric successor** of k)
- We can choose any of those
  - Let's use the symmetric predecessor
  - This is either a leaf – no problem

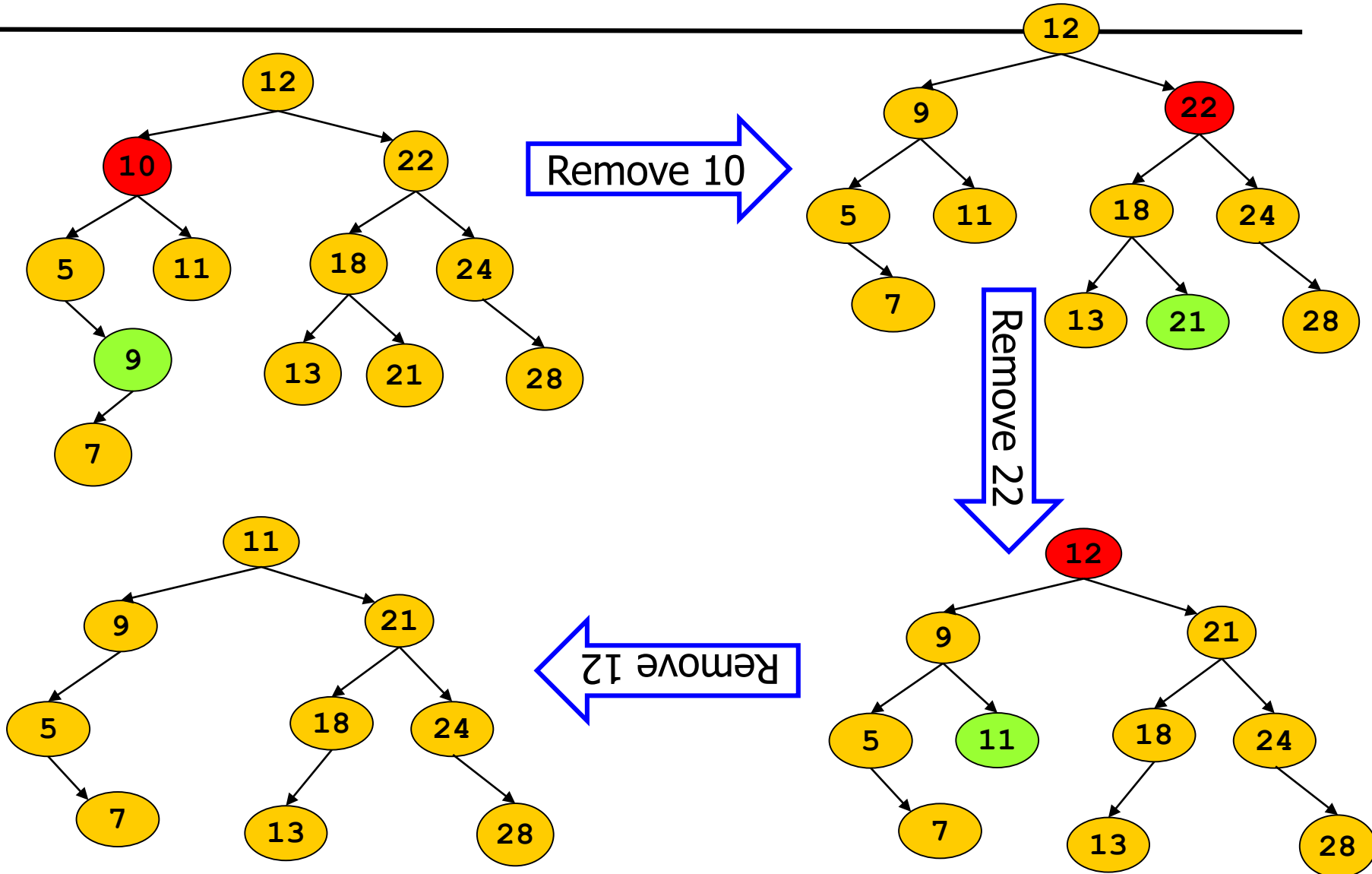


# Observations

- Two candidates
  - Largest value in the left subtree (symmetric predecessor of k)
  - Smallest value in the right subtree (symmetric successor of k)
- We can choose any of those
  - Let's use the symmetric predecessor
  - This is either a leaf
  - Or an **inner node**; but since its label is larger than that of all other labels in the left subtree of q, it can only have a left child
  - Thus it is a node with one child - and can be removed easily

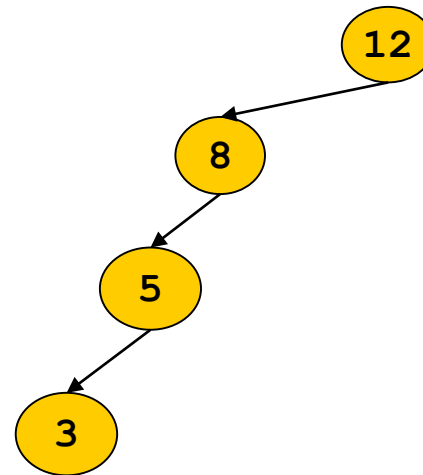
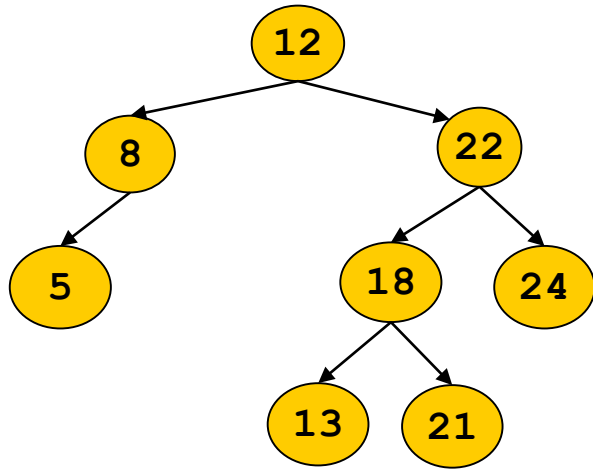


# Example



# Quiz

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# Content of this Lecture

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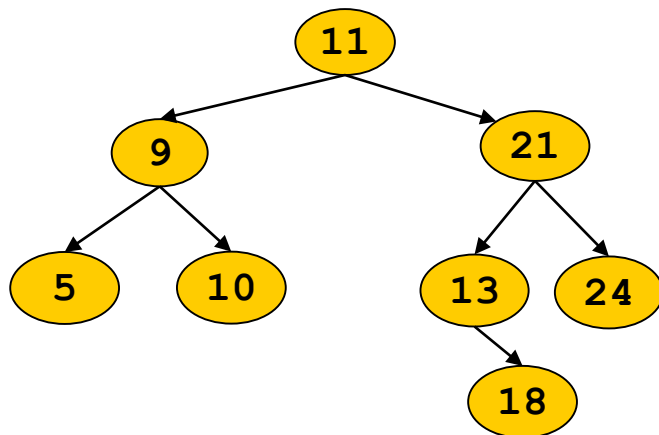
- Trees
- Search Trees
  - Definition
  - Searching
  - Inserting
  - Deleting
- Natural Trees

# Natural Trees

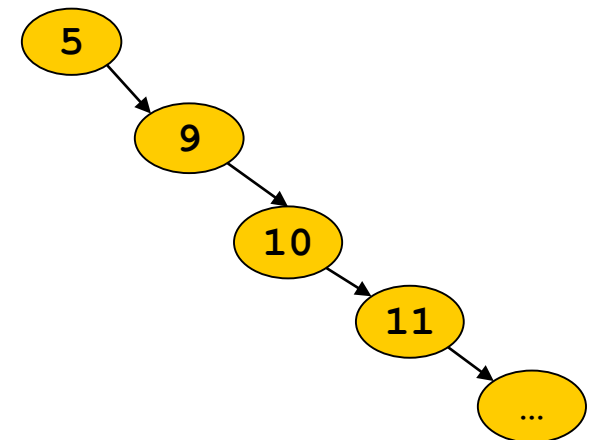
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- A search tree  $T$  created by inserting and deleting  $n$  keys in **random order** is called a **natural tree**
- As any binary tree, it has  $\text{height}(T) \in [n-1, \log(n)]$
- Height depends on **the order in which keys were inserted**
- Example

11,9,10,5,21,13,24,18



5,9,10,11,13,18,21,24



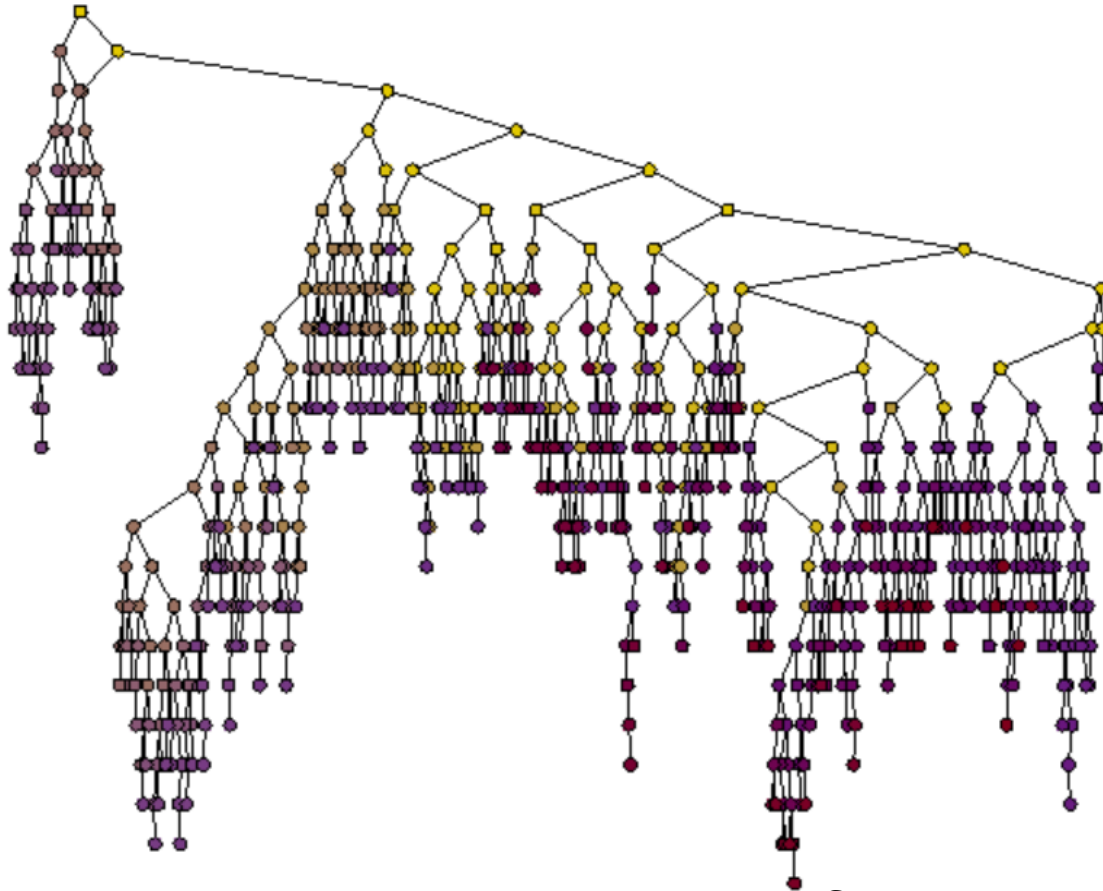
# Average Case

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- A natural tree with  $n$  nodes has maximal height  $n-1$
- Thus, searching will need  $O(n)$  comparisons in worst-case
  - Same for inserting and deleting
- But: Natural trees are not bad on average
  - The average case is  $O(\log(n))$
  - More precisely, a natural tree is on average only  $\sim 1.4$  times deeper than the optimal search tree (with height  $h \sim \log(n)$ )
  - We skip the proof (argue over all possible orders of inserting  $n$  keys), because balanced search trees (AVL trees) are  $O(\log(n))$  also in worst-case and are not much harder to implement

# Example

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Source: [cg.scs.carleton.ca/](http://cg.scs.carleton.ca/)

# Exemplary Questions

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- Construct a natural search tree from the following input, showing all intermediate steps (I: insert; D: delete): I5, I7, I3, I10, D7, I7, I13, I12, D5
- The worst case complexity for inserting/deleting a key into a search tree with  $n=|V|$  nodes is  $O(n)$ . Give an order of the following operations such that this worst case happens for every operation: I5, I7, I3, I10, D7, I7, I13, I12, D5
- For deleting a given key  $k$  in a natural search tree, one may need to find the symmetric predecessor (SP) of a key. Define what a SP is, give an algorithm for finding it (starting from  $k$ ), and analyze its complexity