

## Topic 9

# ***Tree: How to Search Data Faster Than Linear Time***

資料結構與程式設計  
Data Structure and Programming

Sep, 2011

### **What we have learned before...**

- ◆ We have learned some linear type data structures --- list, array, queue, stack, etc.
- ◆ However, in real life, many data types are NOT stored in a linear sequence. For example,
  - Directories and files
  - Employee structure in a company

## In additon, complexity tradeoffs ---

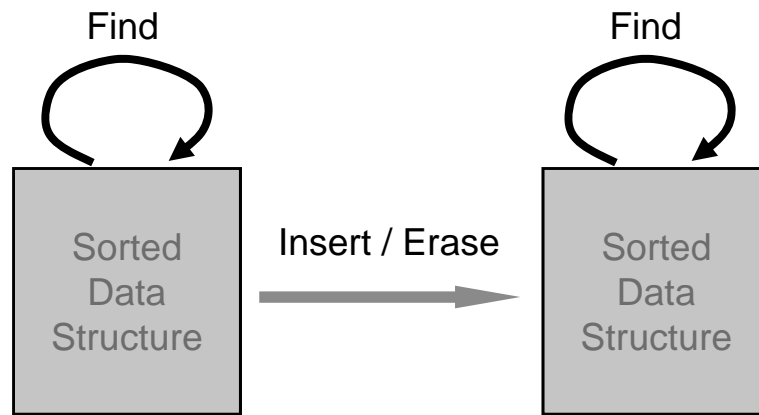
- ◆ Remember in “List and Array” topic we compare the complexity of the following functions

	DList	Array
Insert (any pos)	$O(1)$	$O(n)$ or $O(1)$
Erase (any pos)	$O(1)$	$O(n)$ or $O(1)$
Find	$O(n)$	$O(n \log n)$ to sort the array, $O(\log n)$ to find
Memory Overhead	$8*n + 8$	8

## Remember the difference between → $O(1)$ , $O(\log n)$ , $O(n)$

- ◆ To have better “find” performance
  - Data needs to be sorted
- ◆ List → fast in insert/erase, slow in find
  - Data cannot be sorted
- ◆ Array → not good in insert/erase, OK in find
  - Takes  $O(n \log n)$  to update the order
- ◆ What if we need to
  - Many “find()” operations
  - Some “insert/erase” operation from time to time

## In other words...

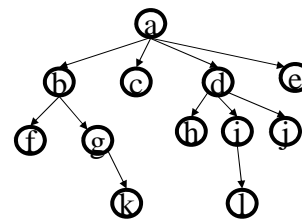


## Better DS for “find”

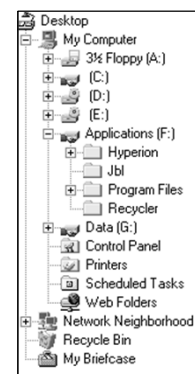
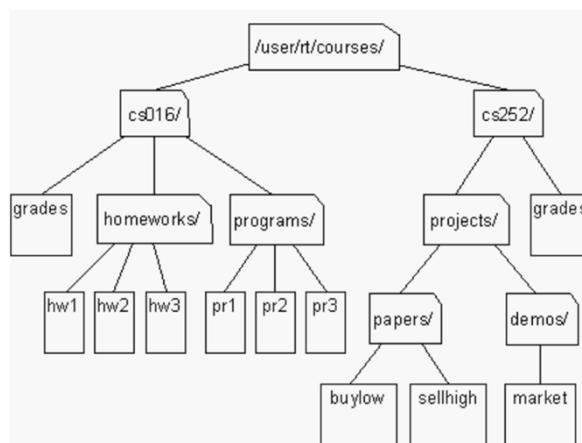
- ◆ We will introduce several data types that have good “find” complexity ( $O(\log n)$ ), and OK “insert/erase” complexity (also  $O(\log n)$ )
  - Heap
  - Set
  - Map
- ➔ They are all different variations of “Tree” data structure

## Tree

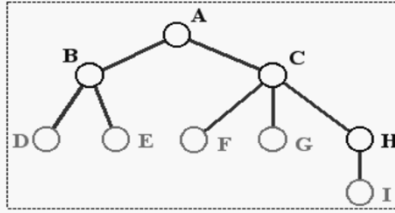
- ◆ In a logic view, it's actually an upside-down tree
- ◆ Usually used in representing hierarchy or relationship
  - e.g. File directories
- ◆ Root
  - branches
  - leaves
  - No re-convergence



## Unix or DOS/Windows file system



- *A* is the **root** node.
- *B* is the **parent** of *D* and *E*.
- *C* is the **sibling** of *B*
- *D* and *E* are the **children** of *B*
- *D, E, F, G, I* are **external nodes**, or **leaves**
- *A, B, C, H* are **internal nodes**
- The **depth (level)** of *E* is 2
- The **height** of the tree is 3
- The **degree** of node *B* is 2

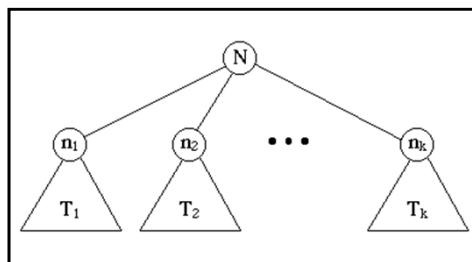


**Property:** (#edges) = (#nodes) - 1

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## Definition of a Tree

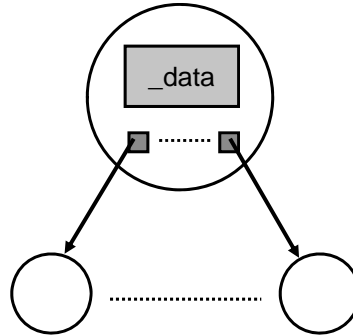
- ◆ This definition is "recursive" and "constructive".
  - 1) A single node is a tree. It is "root."
  - 2) Suppose *N* is a node and *T*<sub>1</sub>, *T*<sub>2</sub>, ..., *T*<sub>*k*</sub> are trees with roots *n*<sub>1</sub>, *n*<sub>2</sub>, ..., *n*<sub>*k*</sub>, respectively. We can construct a new tree *T* by making *N* the parent of the nodes *n*<sub>1</sub>, *n*<sub>2</sub>, ..., *n*<sub>*k*</sub>. Then, *N* is the root of *T* and *T*<sub>1</sub>, *T*<sub>2</sub>, ..., *T*<sub>*k*</sub> are subtrees.



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## Trees Implementation (1)

```
struct TreeNode
{
    MyClass    _data;
    TreeNode*  _child1;
    TreeNode*  _child2;
    .
    .
    .
    TreeNode*  _childm;
};
```



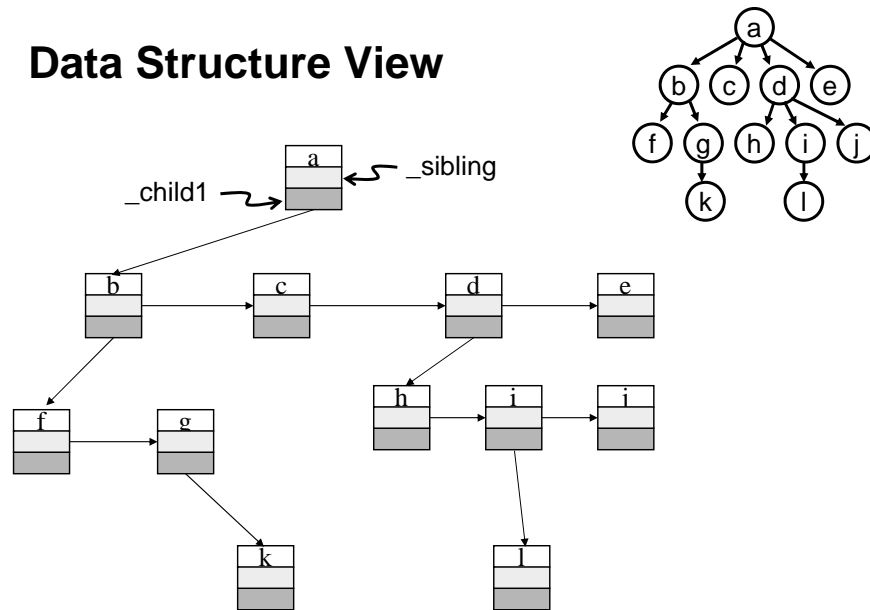
- Straightforward implementation
- Mem usage:  $d * n + 4 * (n * m)$
- Problem
  - ➔ Not flexible in number of children
  - ➔ Good for fixed number of children (e.g. Binary Tree)

## Trees Implementation (2)

```
struct TreeNode
{
    MyClass    _data;
    TreeNode*  _child1;    // head to a list
    TreeNode*  _sibling;   // head to a list
};
```

- Flexible in number of children
- Save memory?
  - Mem usage:  $d * n + 4 * 2n$
- Problem
  - Not straightforward in interpretation
  - Not friendly in child and sibling traversal

## Data Structure View



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## Trees Implementation (3)

```
class TreeNode
{
    MyClass          _data;
    Array<TreeNode *> _children;
};
```

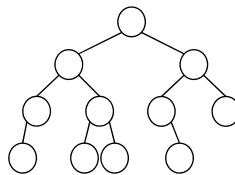
- ◆ Straightforward view
- ◆ Flexible in number of children
- ◆ Mem usage:  $d * n + 4 * (3n - 1)$  (why?)
- ◆ Problem
  - Not easy to access siblings  
(but is that really a problem?)

## Trees Implementation (4)

```
template <class T>
class TreeNode
{
    T _data;
    Array<TreeNode<T> *> _children;
};

template <class T>
class Tree
{
    TreeNode<T>* _root;
};
```

## Traversal of Trees



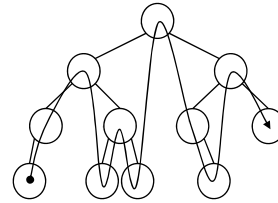
1. Preorder: Process the node, then recursively process the left and right subtrees.
2. Inorder: Process the left subtree, the node, and the right subtree. ← for binary tree
3. Postorder: Process the left subtree, the right subtree, and the node.
4. Levelorder: top-to-bottom, left-to-right order



## Tree Traversal: InOrder

In Order is easily described recursively:

- Visit left subtree (if there is one) In Order
- Visit root
- Visit right subtree (if there is one) In Order



```
algorithm inOrder(TreeNode t)
```

**Input:** a tree node (can be considered to be a tree)

**Output: None.**

```
if t has a left child
```

```
inOrder(left child of t)
```

Visit node  $t$

```
if t has a right child
```

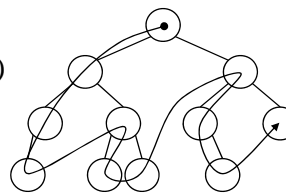
```
inOrder(right child of t)
```

## Tree Traversal: PreOrder

Another common traversal is PreOrder.

It goes as deep as possible (visiting as it goes)

- Visit root
- Visit left subtree in PreOrder
- Visit right subtree in PreOrder



```
algorithm preOrder(TreeNode t)
```

**Input:** a tree node (can be considered to be a tree)

**Output: None.**

```
Visit node t    // Numbering, action, etc
```

```
if t has a left child
```

```
preOrder(left child of t)
```

```

if t has a right child

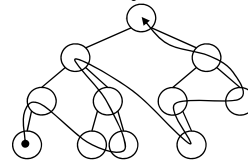
```

```
preOrder(right child of t)
```

## Tree Traversal: PostOrder

PostOrder traversal also goes as deep as possible, but only visits internal nodes during backtracking.

- Visit left subtree in PostOrder
- Visit right subtree in PostOrder
- Visit root



```
algorithm postOrder(TreeNode t)
    Input: a tree node (can be considered to be a tree)
    Output: None.

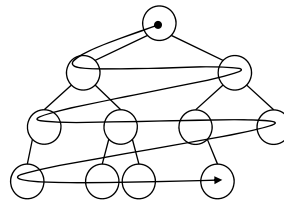
    if t has a left child
        postOrder(left child of t)
    if t has a right child
        postOrder(right child of t)
    Visit node t
```

## LevelOrder Traversal

How to prove the correctness of this algorithm?

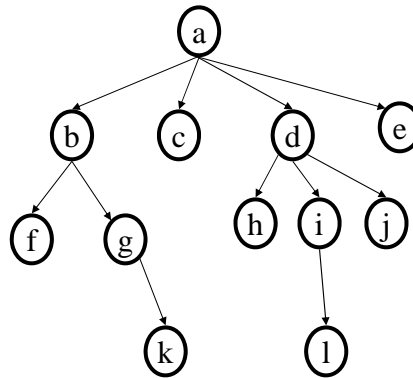
```
algorithm levelOrder(TreeNode t)
    Input: a tree node (can be considered to be a tree)
    Output: None.

    Let Q be a Queue
    Q.enqueue(t)
    while the Q is not empty
        n = Q.dequeue()
        Visit node n
        if n has a left child
            Q.enqueue(left child of n)
        if tree has a right child
            Q.enqueue(right child of n)
```



## Example: Level-Order Queue

Process	Enqueue	Q
	a	a
a	b,c,d,e	bcde
b	f,g	cdefg
c		defg
d	h,i,j	efghij
e		fghij
f		ghij
g	k	hijk
h		ijk
i	l	ijkl
j		kl
k		l
l		



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## Not just learn. Use smartly...

### ◆ Which kind of tree traversal to use?

- Pre-order
- Post-order
- In-order
- Level-order

## Quiz!!

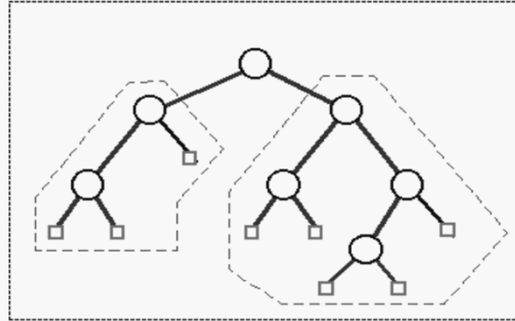
- ◆ Given a tree and we like to perform tree traversal, numbering from 1 to n...
  1. If the number in a node should be smaller than its children (i.e. parent first)...
    - ➔ Pre-order traversal
  2. If the number in a node should be greater than its children (i.e. children first)...
    - ➔ Post-order traversal

## Do we need to have “\_sibling” field?

- ◆ For tree traversal, NO.  
All we need to know is “\_children”.
- ◆ How about other functions?
  - Insert?
  - Erase?Where????
  - ➔ How do these functions operate in a Tree?
- ◆ What if TreeNodes need to be sorted?
  - ➔ We will discuss general sorted tree later.
- ◆ We will look at one special kind of sorted tree  
--- Binary Tree first

## Binary Trees

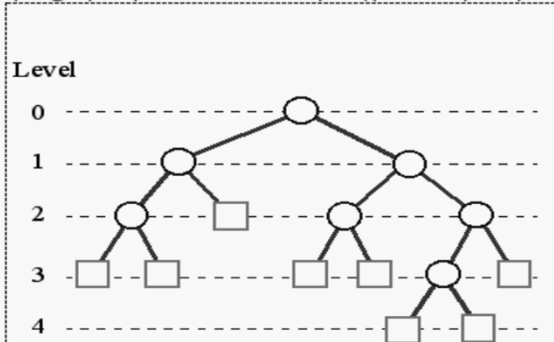
- **Ordered tree:** the children of each node are ordered.
- **Binary tree:** ordered tree with all internal nodes of degree 2.
- Recursive definition of binary tree:
- A **binary tree** is either
  - an external node (leaf), or
  - an internal node (the **root**) and two binary trees (**left subtree** and **right subtree**)



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## Properties of Binary Trees

- $(\# \text{ external nodes }) = (\# \text{ internal nodes }) + 1$
- $(\# \text{ nodes at level } i) \leq 2^i$
- $(\# \text{ external nodes }) \leq 2^{(\text{height})}$
- $(\text{height}) \geq \log_2 (\# \text{ external nodes })$
- $(\text{height}) \geq \log_2 (\# \text{ nodes }) - 1$
- $(\text{height}) \leq (\# \text{ internal nodes }) = ((\# \text{ nodes }) - 1)/2$

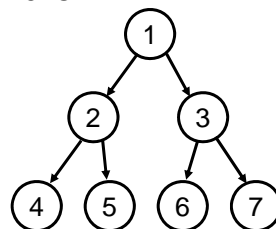


## Special Binary Trees

1. Full / Complete binary tree
2. Binary Search Tree (BST)
3. Balanced Binary Search Tree

## Full Binary Tree

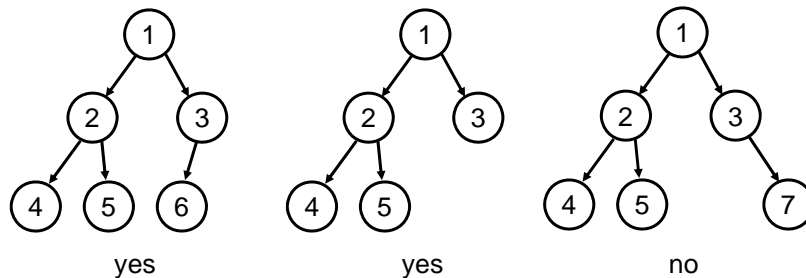
- ◆ A **full** binary tree of height  $h$  is a binary tree of height  $h$  having exactly  $2^{(h+1)} - 1$  nodes
  - All external nodes have same depth =  $h$
  - All internal nodes have non-empty left and right children



height = 2  
#nodes = 7

## Complete Binary Tree

- ◆ A **complete** binary tree is a special case of a binary tree, in which all the levels, except perhaps the last, are full; while on the last level, any missing nodes are to the right of all the nodes that are present.

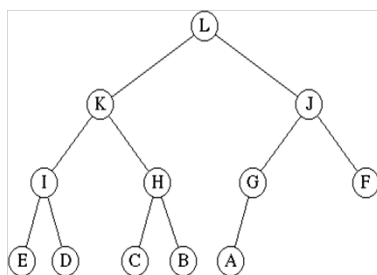


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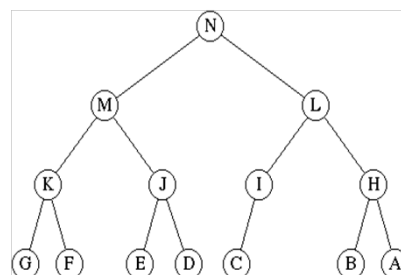
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## Complete/Full Binary Tree Example



- Is this a full binary tree?
  - No – not all leaf nodes are at the same level.
  - No - node G has an empty right and a non-empty left.
- Is it a complete binary tree?
  - Yes



- Is this a full binary tree?
  - No - node I has an empty right and a non-empty left.
- Is it a complete binary tree?
  - No – the final level is not completed in left-to-right fashion

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## Binary Tree Implementation (1)

- ◆ Since the number of children of a binary tree is fixed, we can implement it as

```
template <class T>
class BinaryTreeNode
{
    T                _data;
    BinaryTreeNode<T>* _left;
    BinaryTreeNode<T>* _right;
};

template <class T>
class BinaryTree
{
    BinaryTreeNode<T>* _root;
};

➔ Memory usage:  $d \cdot n + 4 \cdot 2n$ 
```

## Memory Usage Consideration

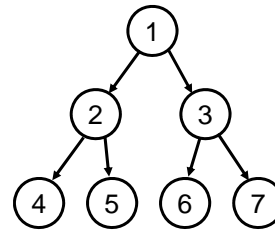
- ◆ An observation ---
    - In 64-bit or higher platform, the memory usage of pointer variables is bigger than that of (unsigned) integers. So we should use “indices” for the child nodes, instead of pointers. That is,
- ```
class BinaryTreeNode {
    T                _data;
    unsigned         _left;
    unsigned         _right;
};
```
- What do you think?
  - What extra data structure do you need?
  - What is the total memory usage?



## Binary Tree Implementation (2)

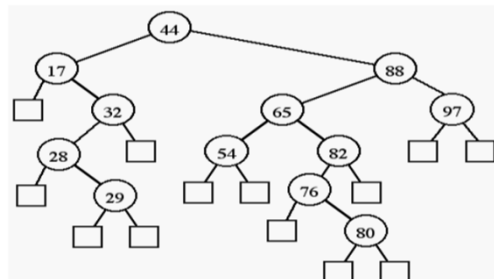
- ◆ If a binary tree is complete
  - ➔ Use array for implementation

- ◆ Let the height of the tree = 'h'
  - #nodes must  $\geq 2^h$  and  $\leq 2^{(h+1)} - 1$
  - root has index = 1
  - A node with index t
    - index of left child =  $2t$
    - index of right child =  $2t + 1$
    - Index of parent =  $t / 2$

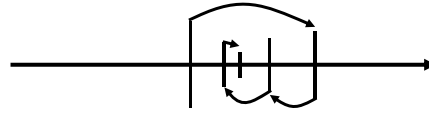


## Binary Search Trees (BST)

- ◆ A binary search tree is a binary tree T such that
  - each internal node stores an item (k, e) of a dictionary.
  - keys stored at nodes in the left subtree of v are less than or equal to k.
  - keys stored at nodes in the right subtree of v are greater than or equal to k.
  - external nodes do not hold elements but serve as place holders.



## Search in BST



- ◆ A binary search tree  $T$  is a **decision tree**, where the question asked at an internal node  $v$  is whether the search key  $k$  is less than, equal to, or greater than the key stored at  $v$ .

```

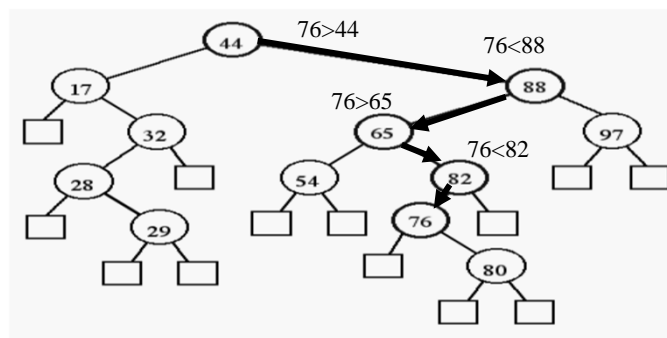
if  $v$  is an external node then
    return  $v$  // mean the key should be inserted here
if  $k = \text{key}(v)$  then
    return  $v$  // find a match
else if  $k < \text{key}(v)$  then
    return  $\text{TreeSearch}(k, T.\text{leftChild}(v))$ 
else  $\{ k > \text{key}(v) \}$ 
    return  $\text{TreeSearch}(k, T.\text{rightChild}(v))$ 
    
```

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## Search Example I

What is the running time ?

Successful findElement(76)



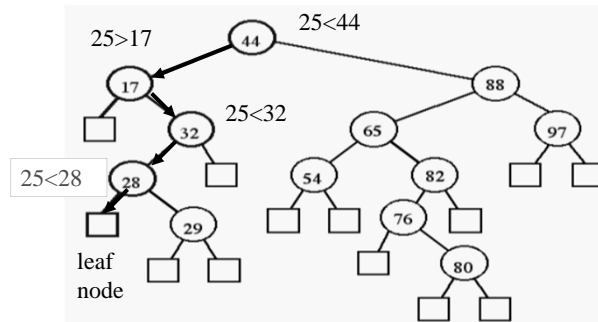
- ◆ A successful search traverses a path starting at the root and ending at an internal node

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## Search Example II

Unsuccessful findElement(25)

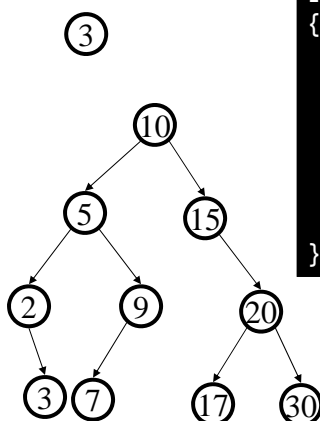
What is the running time ?



- ◆ An unsuccessful search traverses a path starting at the root and ending at an external node

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## Insert a Key

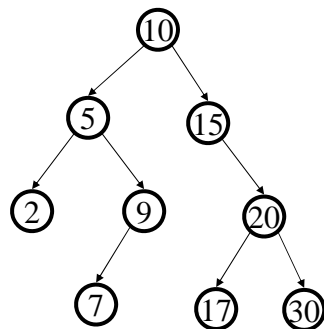


```
TreeNode insert(int x, TreeNode T)
{
    if ( T == NULL )
        return new TreeNode(x,null,null);
    if (x == T.Element)
        return T;
    if (x < T.Element)
        T.Left = insert(x, T.Left);
    else T.Right = insert(x, T.Right);
    return T;
}
```

What is the running time ?

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## Delete a Key



How do you delete:

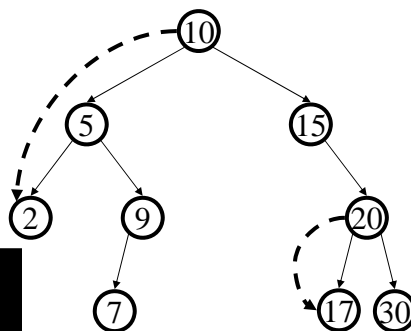
17 ?

9 ?

20 ?????

Let's look at two basic operations:  
min() and successor() first!!

## FindMin



```
TreeNode min(Node T) {  
    if (T.Left == NULL)  
        return T;  
    else  
        return min(T.Left); }  
}
```

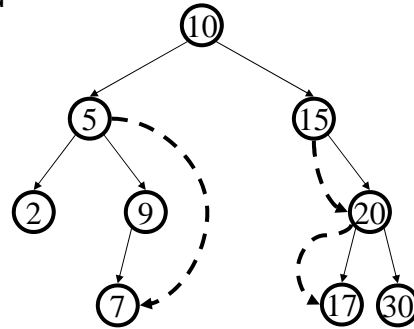
*How many children can the min of a node have?*

## Successor

Find the next larger node in this node's subtree.

- Find “min” of the right child
- [Compare] second largest

```
TreeNode succ(TreeNode T) {  
    if (T.right == NULL)  
        return NULL;  
    else  
        return min(T.right);  
}
```

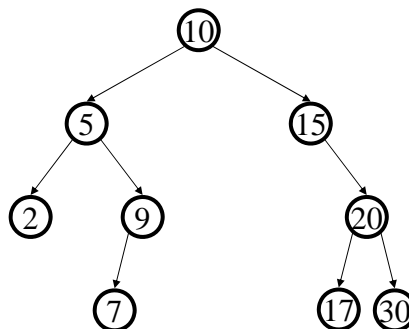


*How many children can the successor of a node have?*

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## Deletion (1) - Leaf Case

Delete(17)

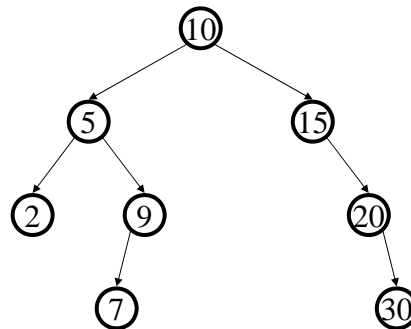


Trivial !!

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## Deletion (2) - One Child Case

Delete(15)

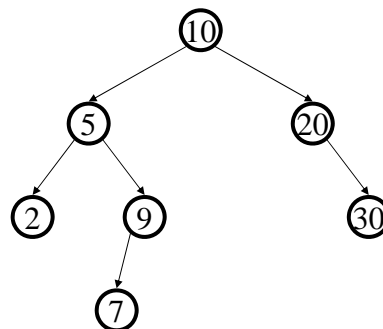


Trivial !!

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## Deletion (3) - Two Children Case

Delete(5)



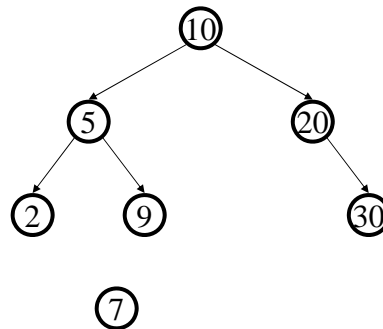
Replace node with value guaranteed to be between the left and right subtrees

→ the successor

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## Deletion (3) - Two Children Case

Delete(5)

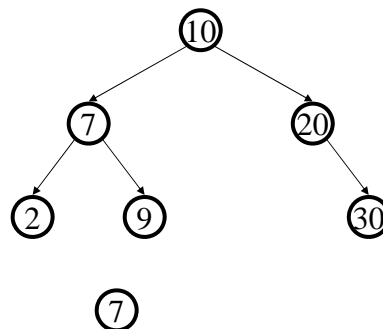


Always easy to delete the successor – always has either 0 or 1 children!

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## Deletion (3) - Two Child Case

Delete(5)



Finally copy data value from deleted successor into original node

What is the cost of a delete operation ?

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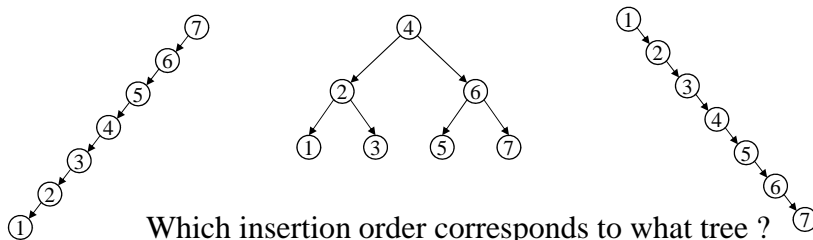
## Cost of the Operations

- ◆ find, insert, delete : **time =  $O(\text{height}(T))$**
- ◆ Need to compute  $\text{height}(T)$
- ◆ For a tree  $T$  with  $n$  nodes:
  - $\text{height}(T) \leq n$
  - $\text{height}(T) \geq \log_2(n)$  (why ?)

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## Height of the Binary Search Tree

- ◆ Height depends critically on the order in which we insert the data:
  - E.g. 1,2,3,4,5,6,7 or 7,6,5,4,3,2,1, or 4,2,6,1,3,5,7



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## Random Input vs. Random Trees

For three items, the shallowest tree is twice as likely as any other – effect grows as  $n$  increases. For  $n=4$ , probability of getting a shallow tree  $> 50\%$

Inputs

1,2,3

3,2,1

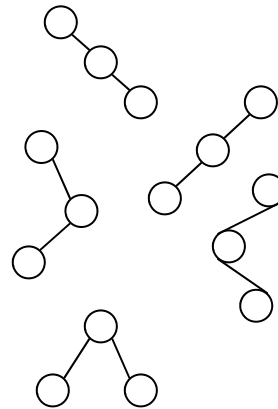
1,3,2

3,1,2

2,1,3

2,3,1

Trees



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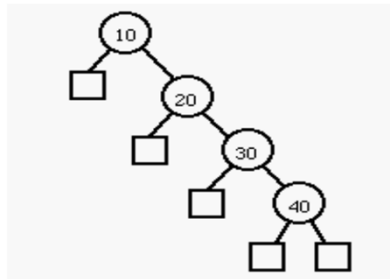
## Average cost

- ◆ The *average, amortized* cost of  $n$  insert/find operations is  $O(\log(n))$
- ◆ But the *average, amortized* cost of  $n$  insert/find/delete operations can be as bad as  $\sqrt{n}$ 
  - $\log 10000$  vs.  $\sqrt{10000}$
  - Deletions make life harder
  - Read the book for details
- ◆ Need guaranteed cost  $O(\log n)$

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## Time Complexity

- ◆ The height of binary search tree is  $n$  in the worst case, where a binary search tree looks like a sorted sequence



- ◆ To achieve good running time, we need to keep the tree **balanced**, i.e., with  $O(\log n)$  height.

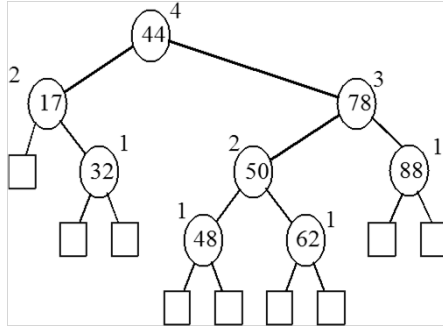
## Self Adjusting Binary Search Trees

- ◆ Insertions/removals may “deepen” and “unbalance” a binary search tree.
- ◆ Self-adjusting binary search trees automatically restore balance after each insertion/removal by performing a series of **rotations**.
- ◆ Self-adjusting binary search trees **insure** good worst-case performance.

## Balanced Binary Search Trees

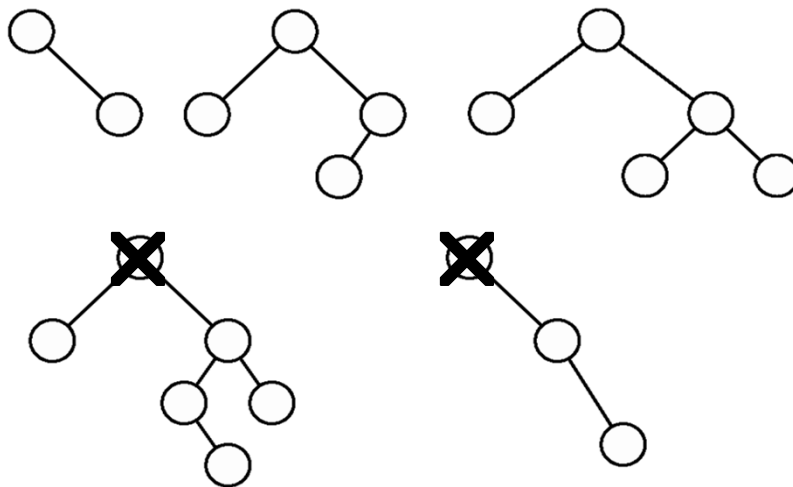
## AVL Tree

- ◆ G. M. Adel'son-Vel'skii and E. M. Landis, "An Algorithm for the Organization of Information," *Soviet Math. Doklady* **3** (1962), pp. 1259—1262.
- ◆ **AVL trees are balanced.**
- ◆ An AVL Tree is a **binary search tree** such that for every internal node  $v$  of  $T$ , the *heights of the children of  $v$*  can differ by at most 1.



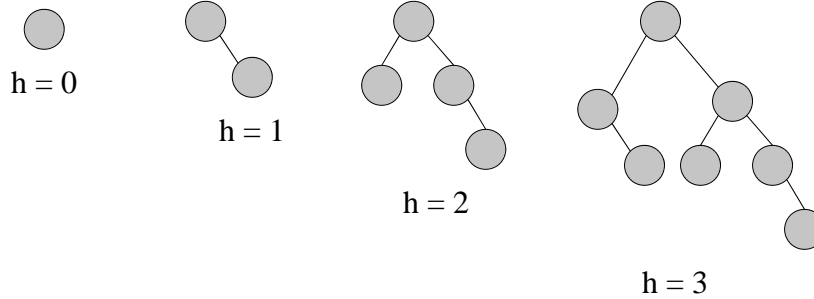
An example of an AVL tree where the heights are shown next to the nodes:

## Example



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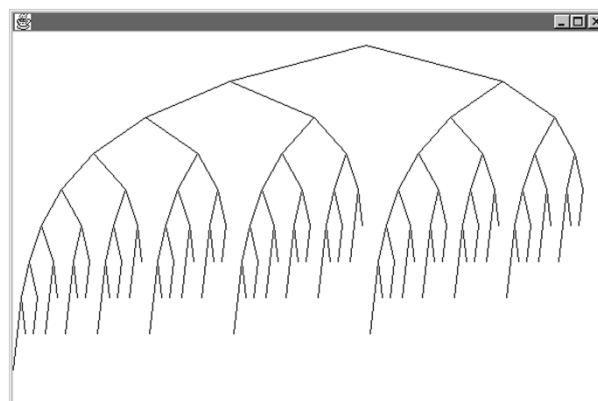
## AVL Trees



Given an AVL tree of height  $h$ , what is the minimal number of nodes that the tree may contain?

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## AVL - height 9



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## Height Of An AVL Tree

The height of an AVL tree that has  $n$  nodes is at most  $1.44 \log_2 (n+2)$ .

The height of every  $n$  node binary tree is at least  $\log_2 (n+1)$ .

$$\log_2 (n+1) \leq \text{height} \leq 1.44 \log_2 (n+2)$$

$$\underline{O(\log(n))}$$

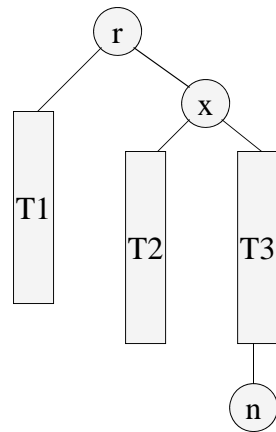
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## Insertion

- ◆ A binary search tree  $T$  is called **balanced** if for every node  $v$ , the height of  $v$ 's children differ by at most one.
- ◆ Inserting a node into an AVL tree involves performing an `expandExternal(w)` on  $T$ , which changes the heights of some of the nodes in  $T$ .
  - If an insertion causes  $T$  to become unbalanced, we travel up the tree from the newly created node until we find the first node  $x$  such that its grandparent  $z$  is unbalanced node.
  - Since  $z$  became unbalanced by an insertion in the subtree rooted at its child  $y$ ,  $\text{height}(y) = \text{height}(\text{sibling}(y)) + 2$
  - Now to rebalance...

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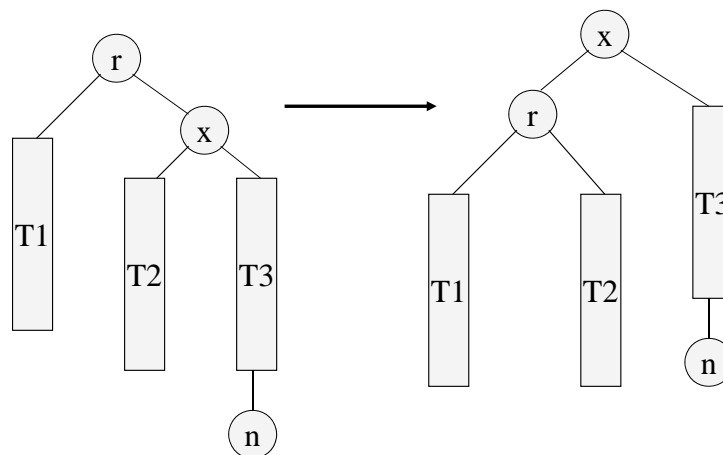
## The “outie” case



- ◆  $r$  - the nearest ancestor which is out of balance
- ◆  $n$  - the newly inserted node
- ◆ height of  $T1$ ,  $T2$ , and  $T3$  are all the same, say  $h$

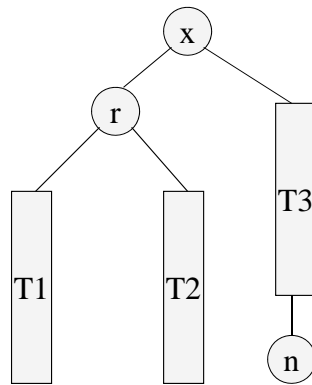
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## single rotation



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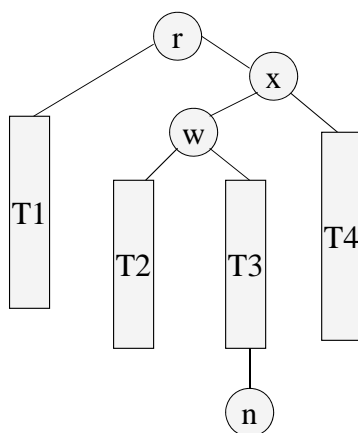
## After the rotation



- ◆ x is now the root
- ◆ the height of the tree is the same as it was before inserting the node, so no other ancestor is unbalanced
- ◆ the root x is balanced

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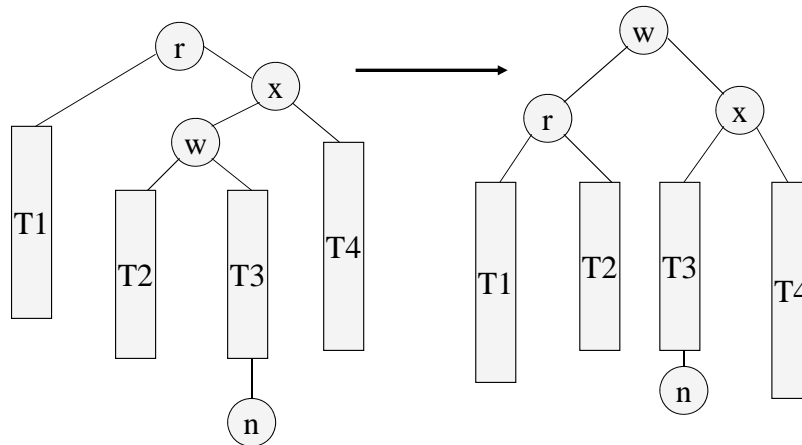
## The "innie" case



- ◆ r is the nearest out-of-balance ancestor
- ◆ T1 and T4 have height  $h$
- ◆ T2 and T3 have height  $h-1$
- ◆ n is the newly inserted node - either in T2 or T3

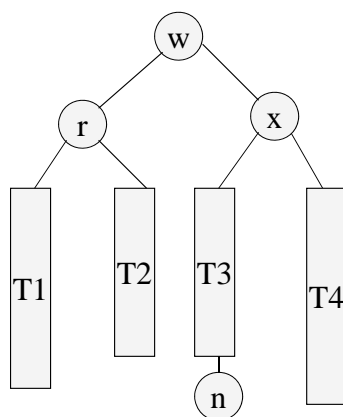
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## Double Rotation



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## After the Rotation

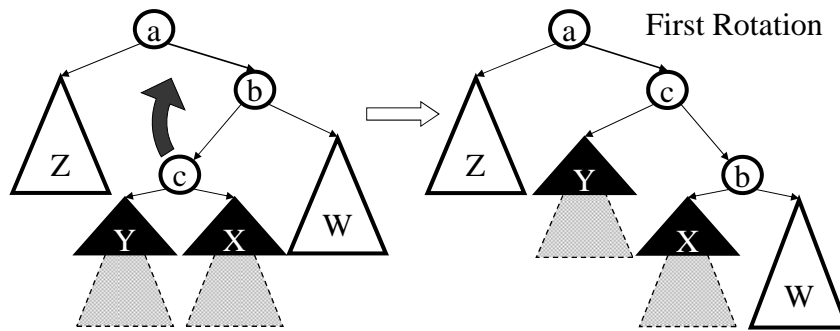


- ◆ w is now the root with left child r and right child x
- ◆ The height of the tree is the same as before the insertion, so no other ancestor is now out-of-balance
- ◆ This tree is balanced

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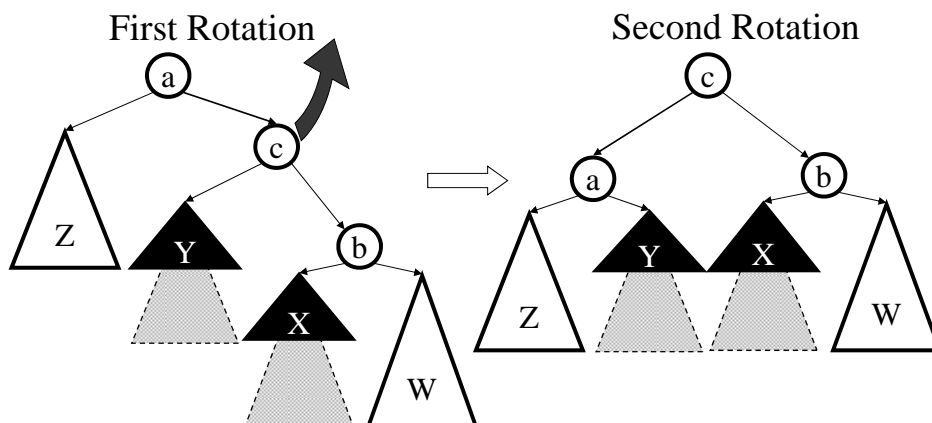


## More precisely



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## Double Rotation Completed



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## The other rotations

- ◆ These two demonstrations show the Single Left rotation and the Double Left rotation (used when the nearest out-of-balance ancestor is too heavy on the right)
- ◆ Similar rotations are performed when the nearest out-of-balance ancestor is heavy on the left -- these are called Single Right and Double Right Rotations

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## Deletion from an AVL Tree

- ◆ Deletion of a node from an AVL tree requires the same basic ideas, including single and double rotations, that are used for insertion
- ◆ We are NOT going into details here.... (Don't need to memorize the steps; understand the principles!!)
  - Please refer to any DS books or the appendix slides at the end

## Building an AVL Tree

Input: sequence of  $n$  keys (unordered)

19 3 4 18 7

Insert each into initially empty AVL tree

$$\sum_{i=1}^n \log i \leq \sum_{i=1}^n \log n = O(n \log n)$$

But, suppose input is already sorted ...

3 4 7 18 19

Can we do better than  $O(n \log n)$ ?

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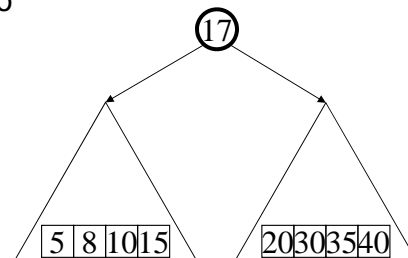
## AVL BuildTree

|   |   |    |    |    |    |    |    |    |
|---|---|----|----|----|----|----|----|----|
| 5 | 8 | 10 | 15 | 17 | 20 | 30 | 35 | 40 |
|---|---|----|----|----|----|----|----|----|

Divide & Conquer

- Divide the problem into parts
- Solve each part recursively
- Merge the parts into a general solution

How long does  
divide & conquer take?



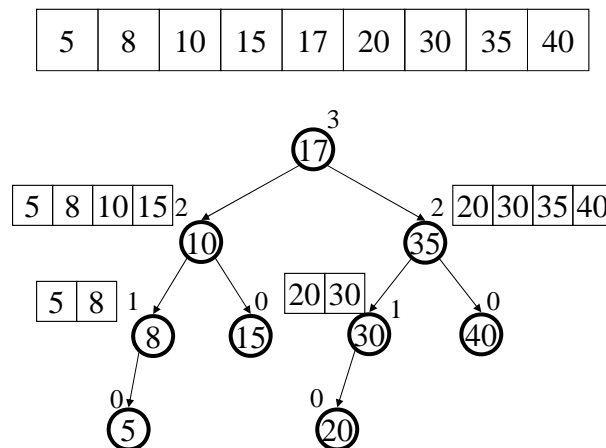
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## BuildTree Example



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## Thinking About AVL

### Observations

- + Worst case height of an AVL tree is about  $1.44 \log n$
- + Insert, Find, Delete in worst case  $O(\log n)$
- + Only one (single or double) rotation needed on insertion
- + Compatible with lazy deletion
- $O(\log n)$  rotations needed on deletion
- Height fields must be maintained (or 2-bit balance)

### Coding complexity?

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## AVL Performance

| Method                               | Worst Case  |
|--------------------------------------|-------------|
| void insert(Comparable element)      | $O(\log N)$ |
| boolean contains(Comparable element) | $O(\log N)$ |
| void delete(Comparable element)      | $O(\log N)$ |
| int size()                           | $O(1)$      |
| boolean isEmpty()                    | $O(1)$      |

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## Pros and Cons of AVL Trees

Pro:

- All operations guaranteed  $O(\log N)$
- The height balancing adds no more than a constant factor to the speed of insertion

Con:

- Space consumed by height field in each node
- Slower than ordinary BST on random data

*Can we guarantee  $O(\log N)$  performance with less overhead?*

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## Alternatives to AVL Trees

- ◆ Weight balanced trees
    - keep about the same number of nodes in each subtree
    - not nearly as nice
  - ◆ Others
    - Splay trees
    - 2-3-4 trees
    - red-black trees
- Too many to cover here...

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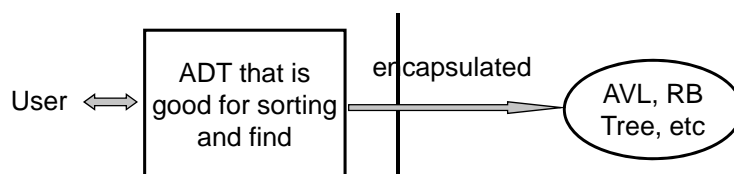
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## Summary about the Tree ADT

- ◆ A good DS for representing hierarchy or relationship
- ◆ Important variations: binary tree, binary search tree, balanced binary search tree
- ◆ Balanced Binary Search Tree
  - All operations are equal or less than  $O(\log(n))$
  - Good example for "Abstract" DT



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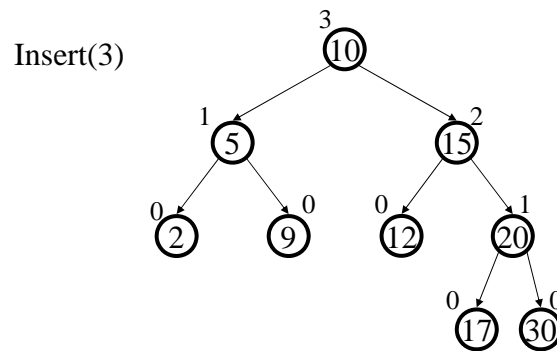
# Appendix Slides

## Steps in deleting X

- ◆ Reduce the problem to the case where X has only one child
- ◆ Delete the node X. The height of the subtree formerly rooted at X has been reduced by one
- ◆ We must trace the effect on the balance from X all the way back to the root until we reach a node which does not need adjustment

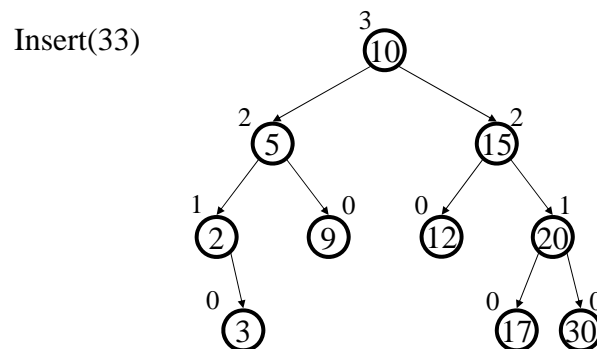
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## Easy Insert



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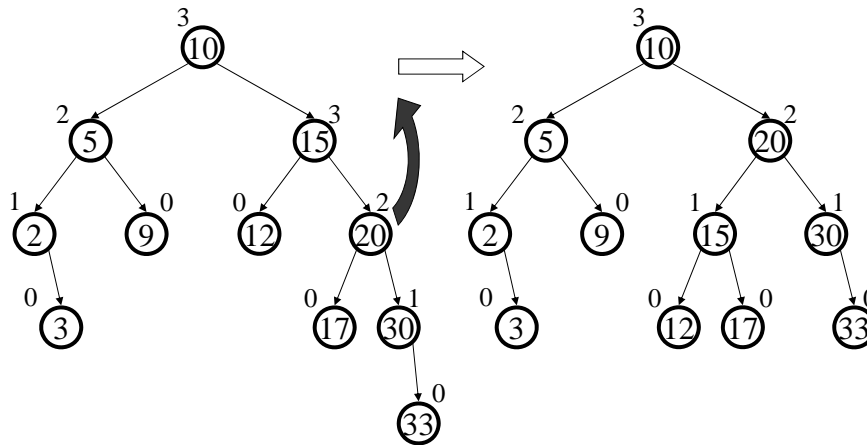
## Hard Insert (Bad Case #1)



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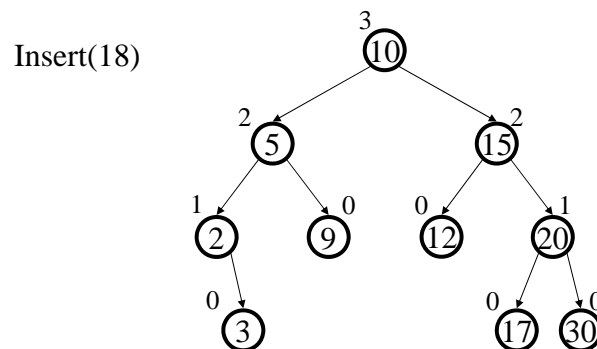


## Single Rotation



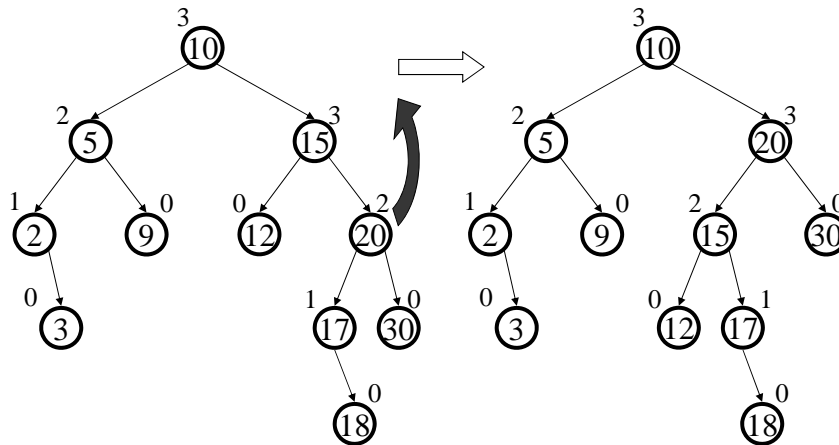
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## Hard Insert (Bad Case #2)



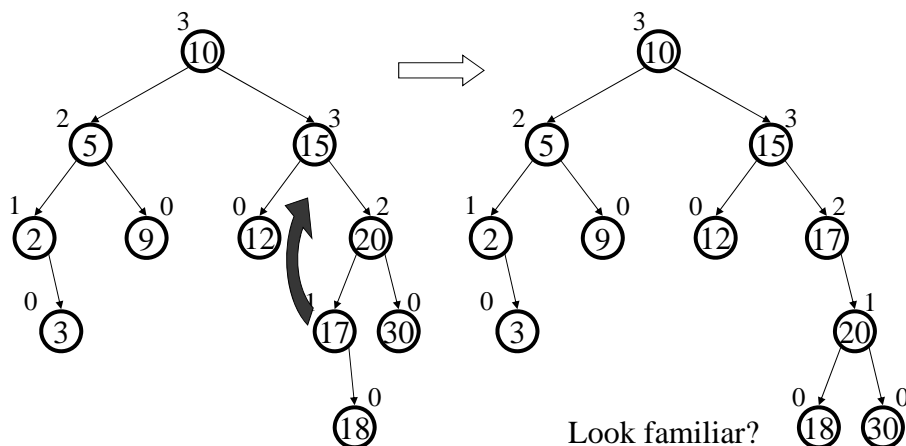
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## Single Rotation (oops!)



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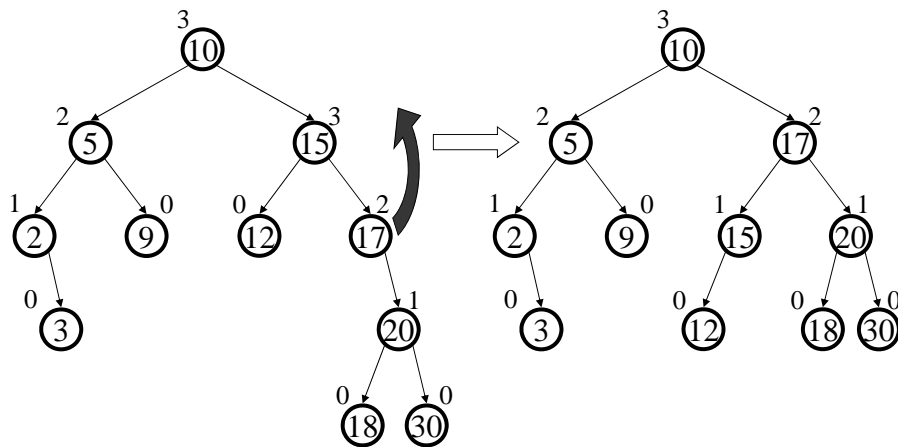
## Double Rotation (Step #1)



Look familiar?

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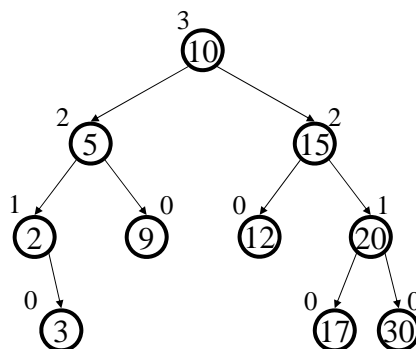
## Double Rotation (Step #2)



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## Deletion: Really Easy Case

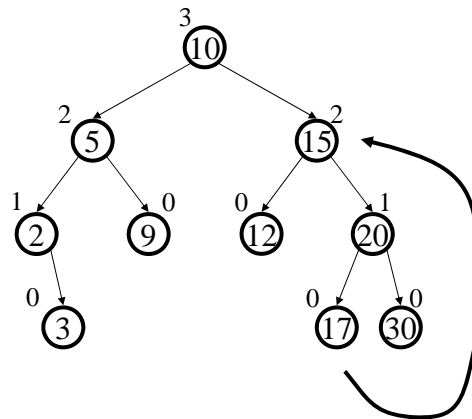
Delete(17)



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## Deletion: Pretty Easy Case

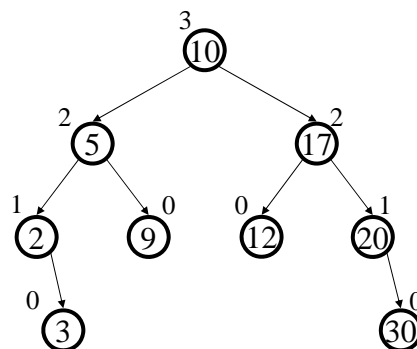
Delete(15)



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## Deletion: Pretty Easy Case (cont.)

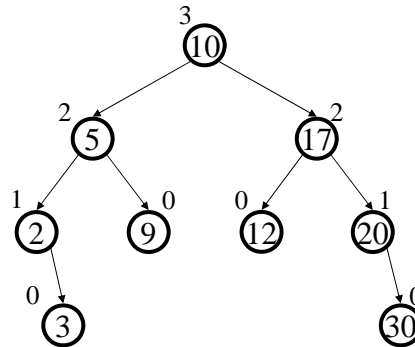
Delete(15)



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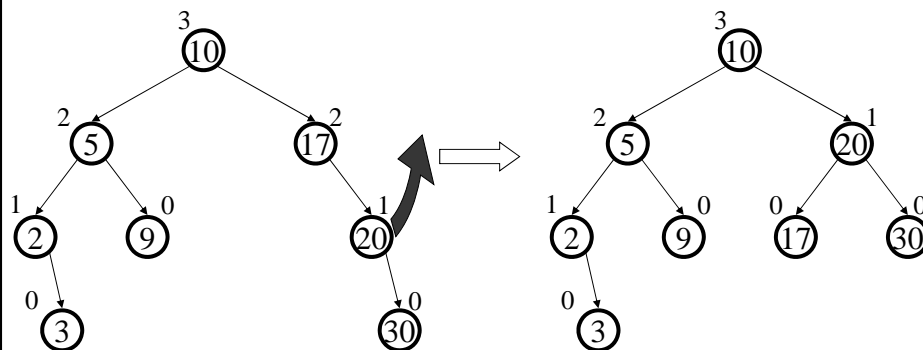
## Deletion (Hard Case #1)

Delete(12)



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## Single Rotation on Deletion

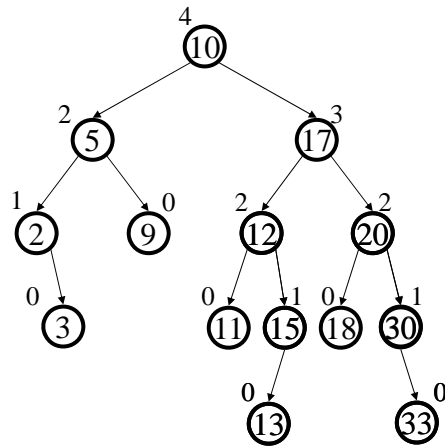


Deletion can differ from insertion – *How?*

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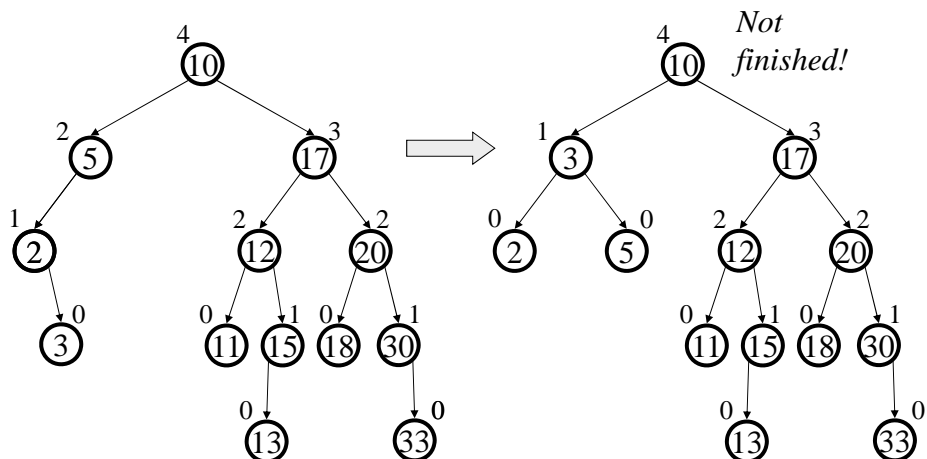
## Deletion (Hard Case)

Delete(9)



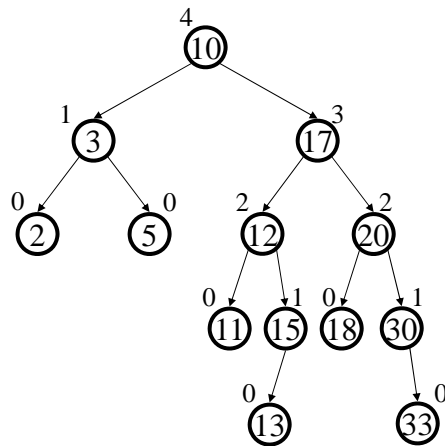
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## Double Rotation on Deletion



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## Deletion with Propagation



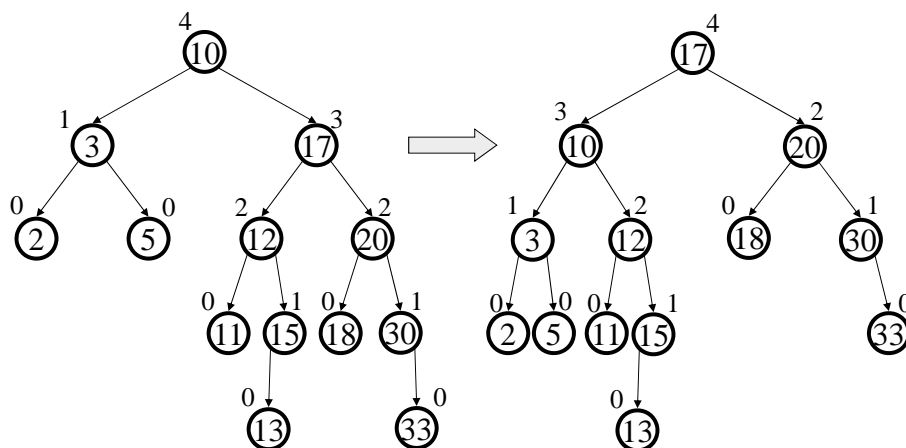
What different about this case?

We get to choose whether to single or double rotate!



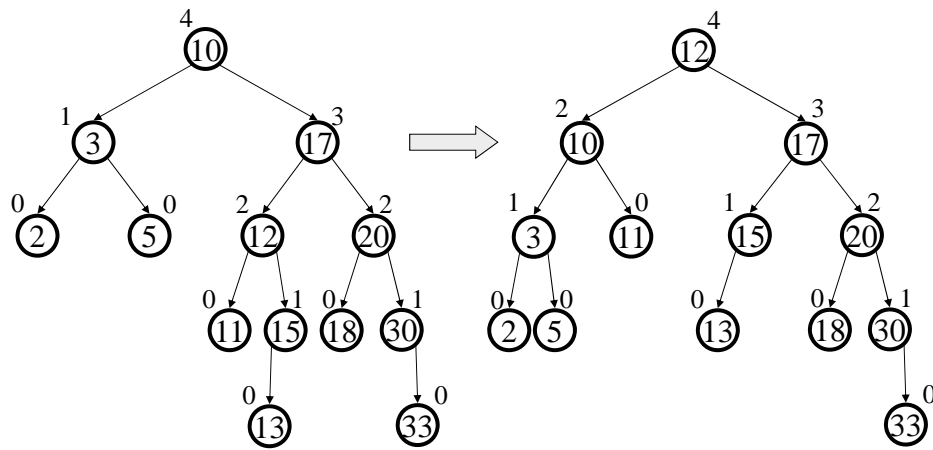
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## Propagated Single Rotation



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## Propagated Double Rotation



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