

# Data Structures and Algorithms

(CS210A)

Semester I – 2014-15

## Lecture 18:

### Analysis of

- Red Black trees
- Nearly Balanced BST

A **Red Black** Tree is height balanced

A **detailed proof** from **scratch**

# Red Black Tree

**Red Black** tree:

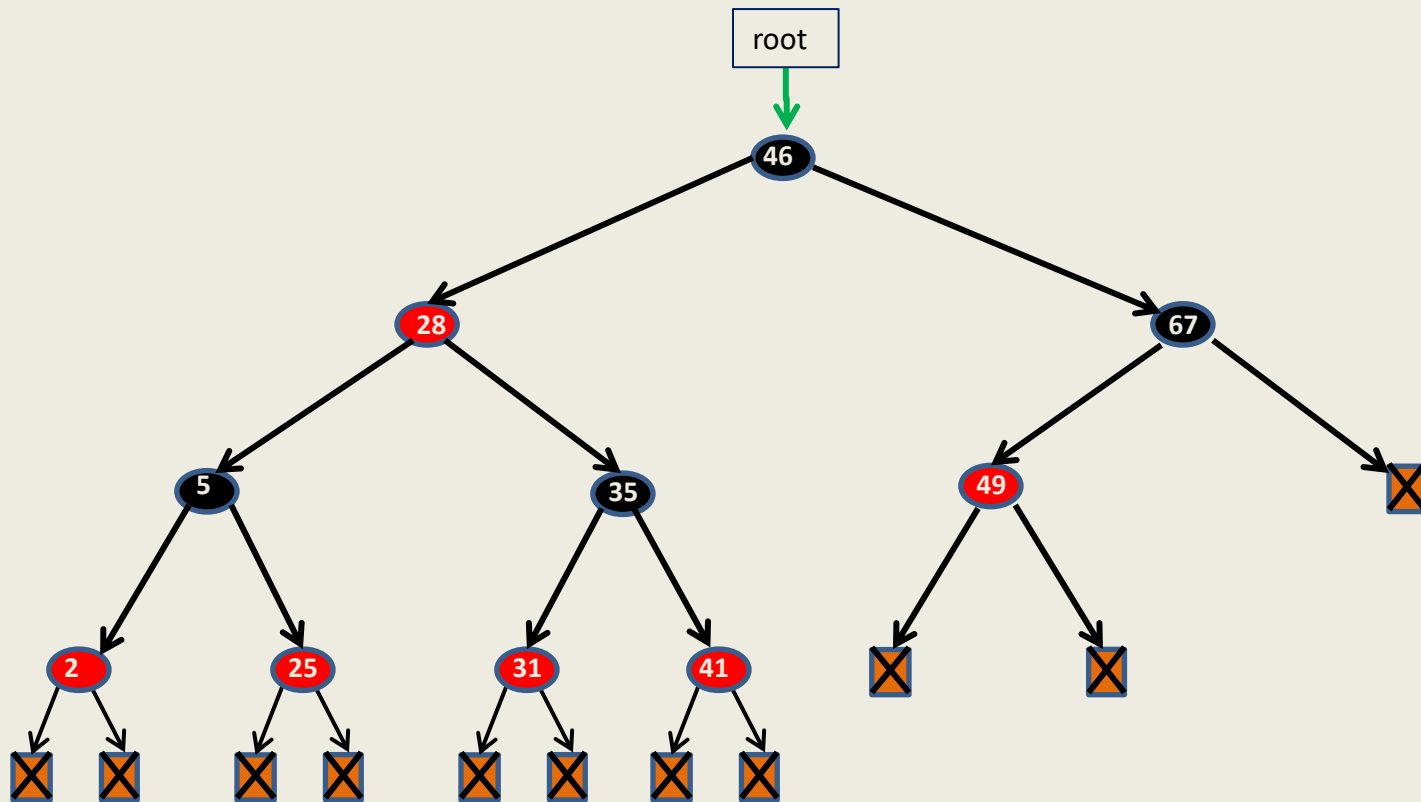
a **full** binary search tree with each leaf as a **null** node and satisfying the following properties.

- Each node is colored **red** or **black**.
- Each leaf is colored **black** and so is the root.
- Every **red** node will have both its children **black**.
- No. of **black nodes** on a path from root to each leaf node is same.



**black height**

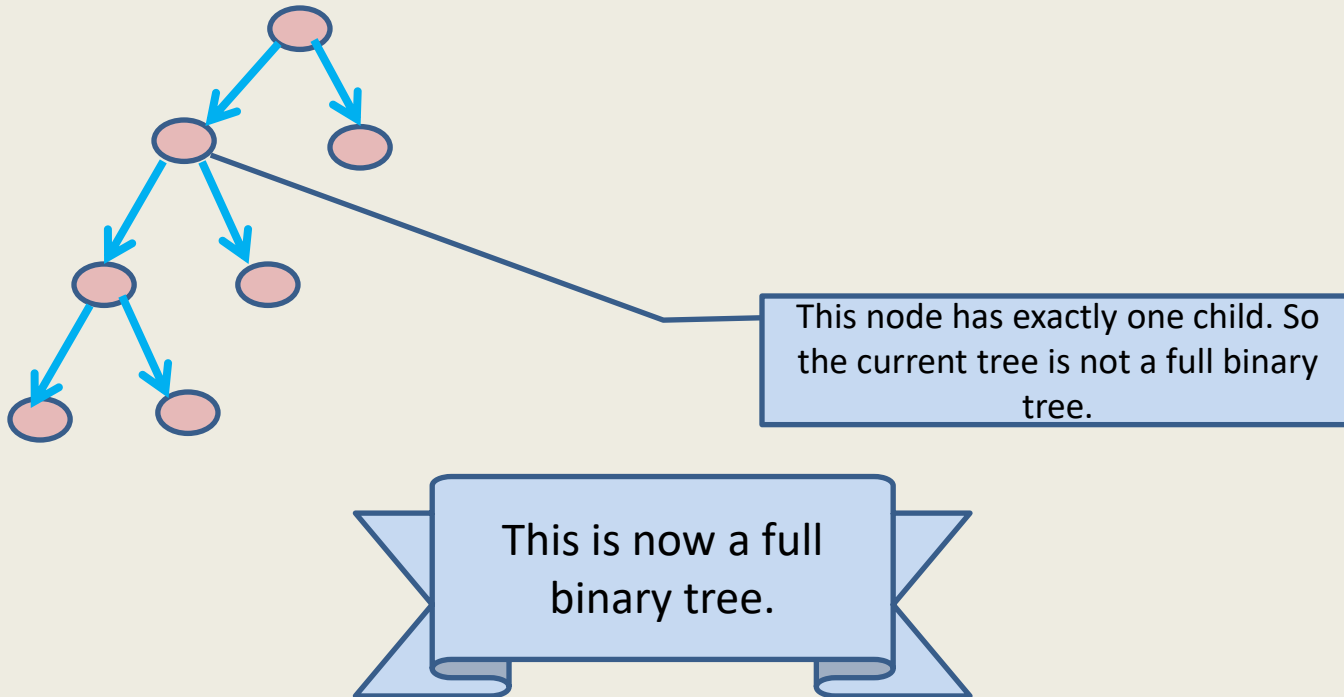
# A red-black tree



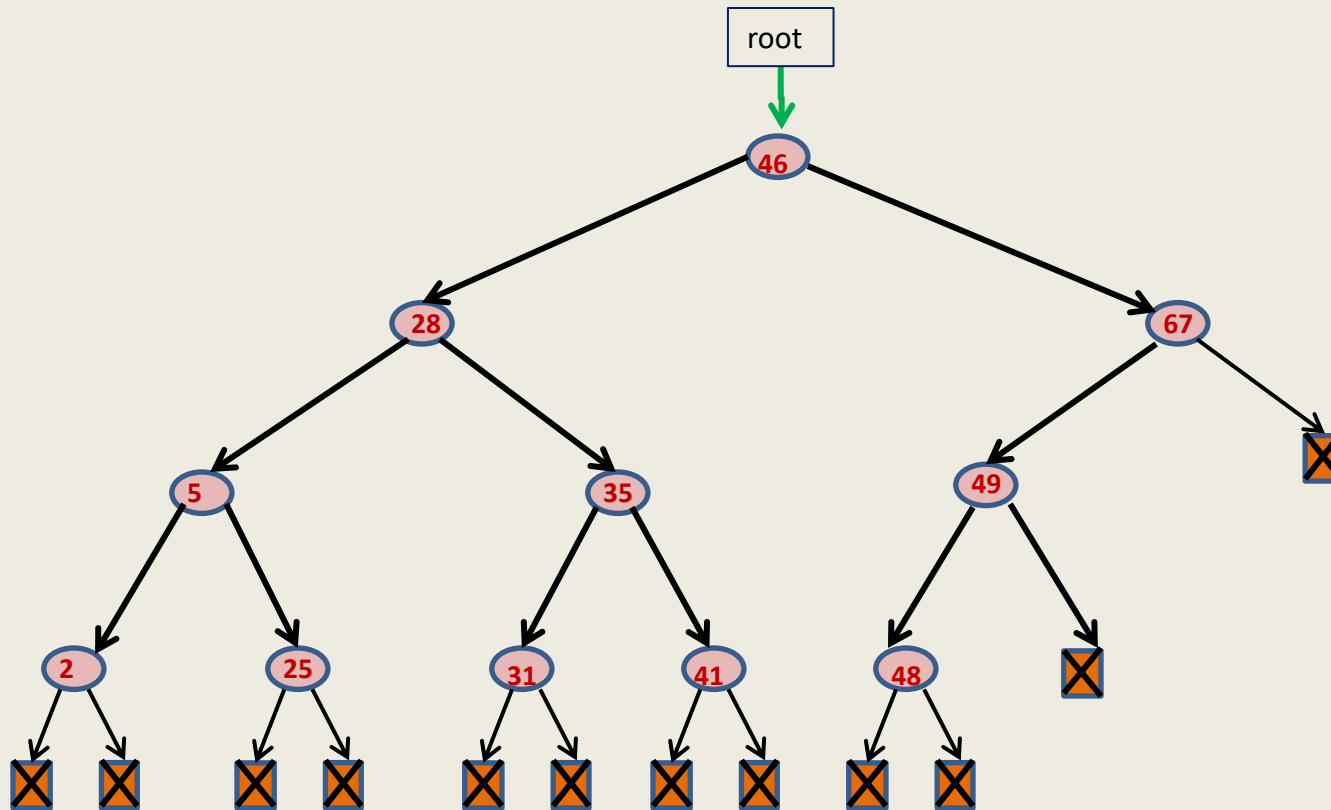
# Terminologies

## Full binary tree:

A binary tree where every internal node has exactly two children.

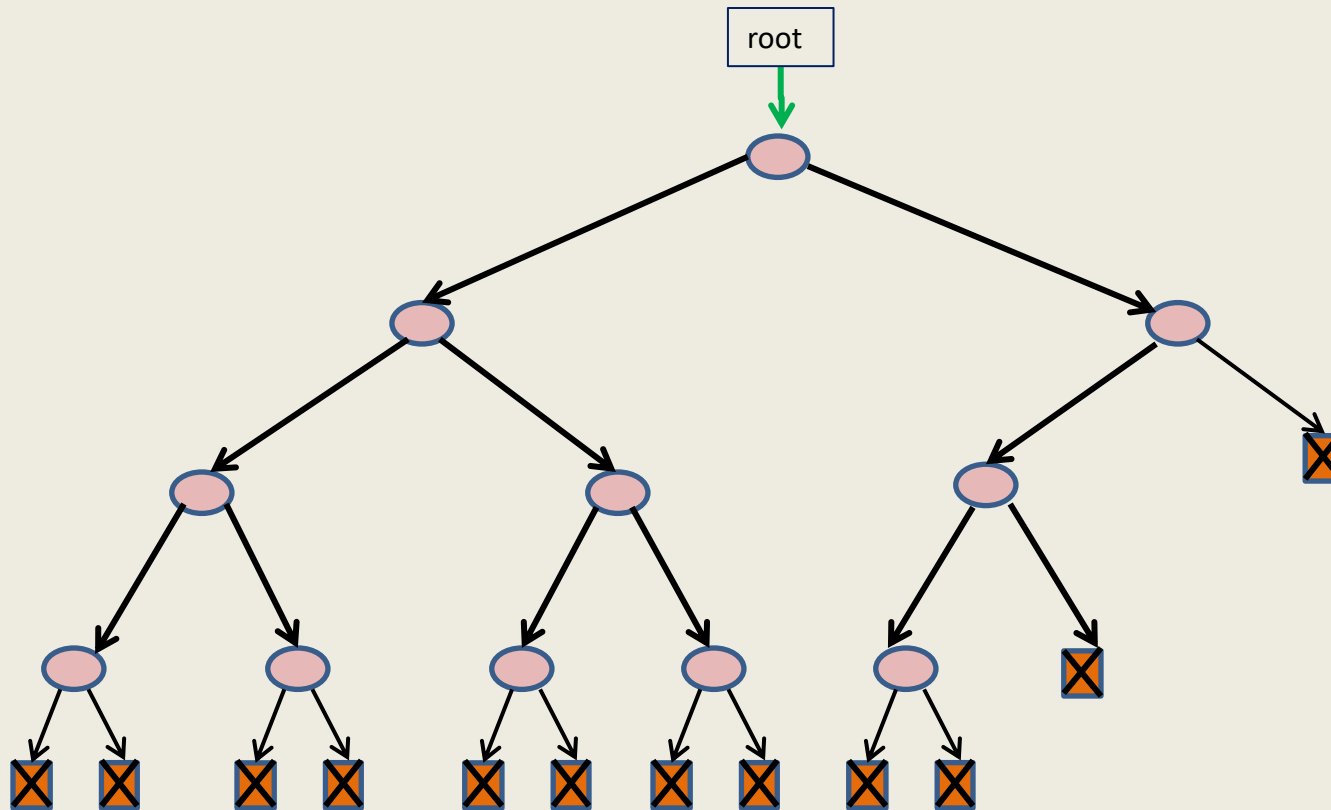


# Red-black tree: as a Full Binary Tree



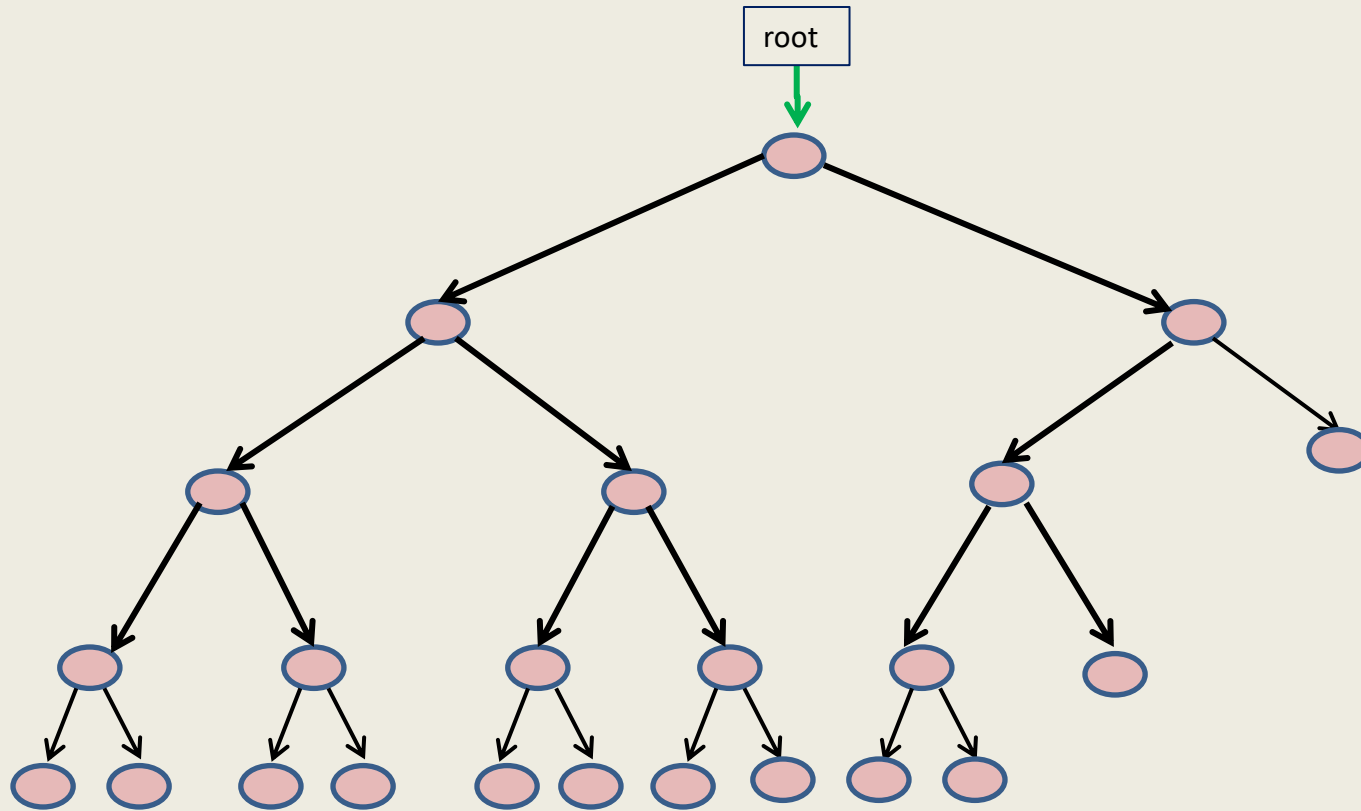
Ignore the values

# Red-black tree: as a Full Binary Tree



Ignore the distinction  
between internal nodes  
and leaf nodes

## Red-black tree: as a Full Binary Tree

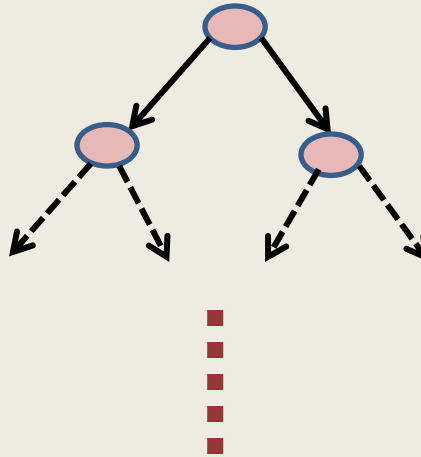




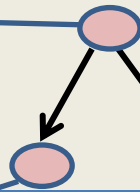
# **Properties of a Red-Black Tree viewed as a full binary tree**

**Relationship between  
Number of leaf nodes and Number of internal nodes**

# A full binary tree



This node must have a left child since the tree is a full binary tree

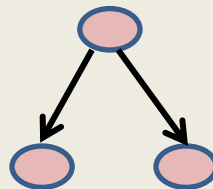
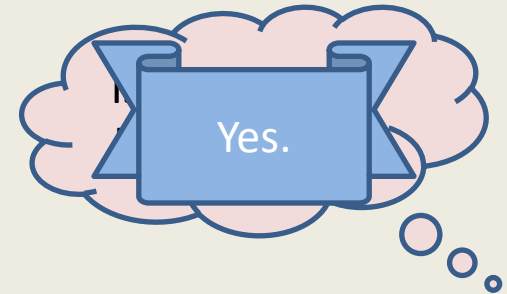
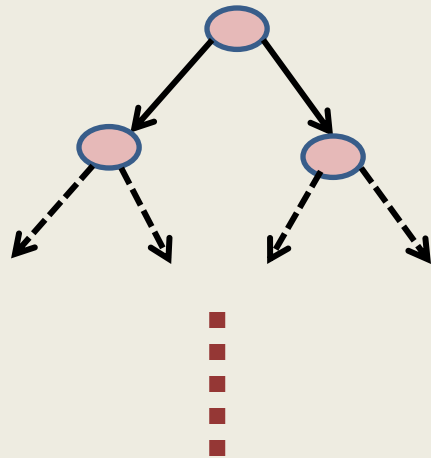


This node must be a leaf node. Give reason.

Otherwise this node won't be the deepest node.

Any deepest node

# A full binary tree



What happened to the number of leaf nodes ?

Reduced by one

What happened to the number of internal nodes ?

Reduced by one

# A full binary tree

Analyze the process:

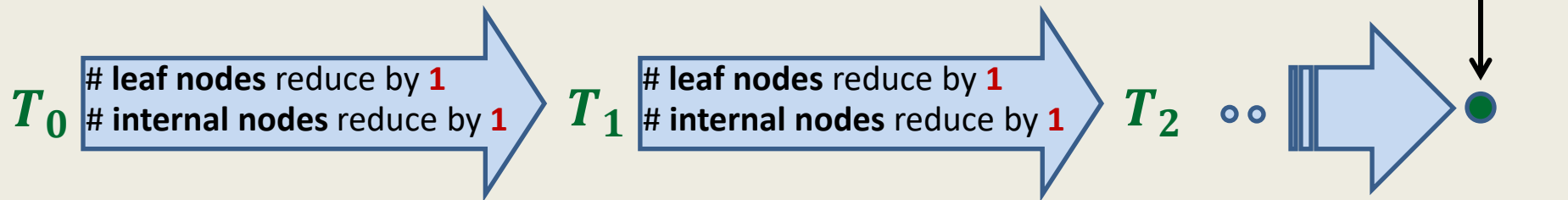
Repeat

{ Delete the deepest node and its sibling }

until only **root** remains

Let  $T_0$  be the full binary tree before the process starts.

Let  $T_1, T_2, \dots$  be the full binary trees after 1<sup>st</sup>, 2<sup>nd</sup>, ... iterations of the process.



**Question:** What might be the relation between leaf nodes and internal nodes in  $T_0$  ?

**Answer:** No. of leaf nodes in  $T_0$  = No. of internal nodes in  $T_0$  + 1.

# A full binary tree

**Question:** If  $i$  is the number of internal nodes in a full binary tree  $T$ , what is the size (number of nodes) of the tree ?

**Answer:**  $2i + 1$

**Question:** What is the size of a **Red Black** tree storing  $n$  keys ?

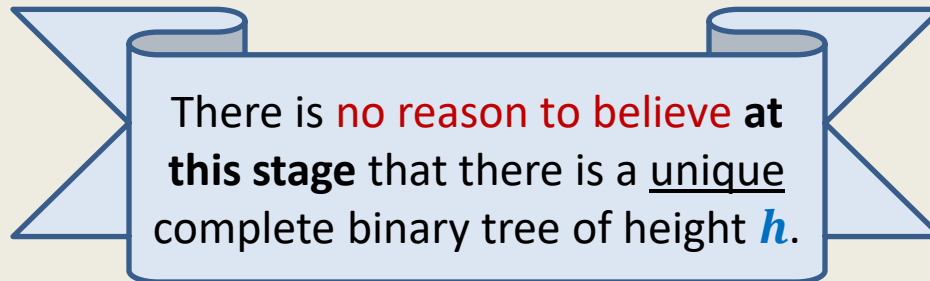
**Answer:**  $2n + 1$

# A complete binary tree of height $h$ and its Properties

# A complete binary tree of height $h$

## Definition:

A full binary tree of height  $h$  is said to be a **complete** binary tree of height  $h$  if every leaf node is at depth  $h$ .



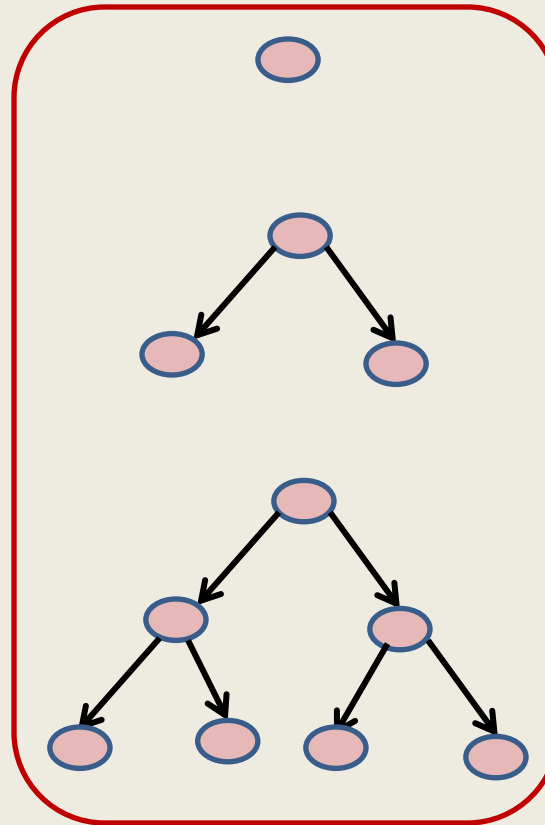
**Question:** How will any complete binary tree of height  $h$  look like ?

# A complete binary tree of height $h$

Complete binary tree of height 1 ?

Complete binary tree of height 2 ?

Complete binary tree of height 3 ?



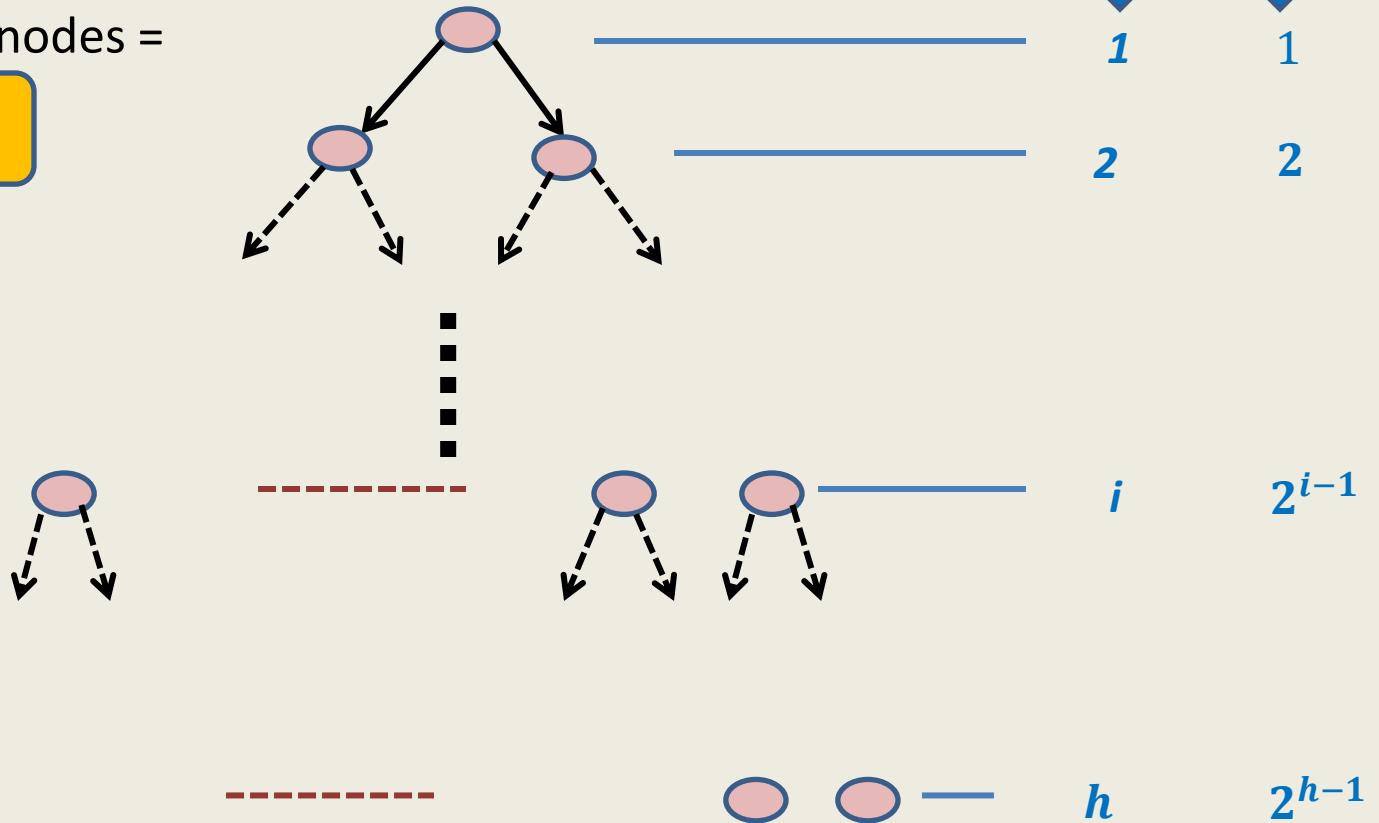
Try to generalize  
the type of tree  
shown here to  
tree of height  $h$ .



# A complete binary tree of height $h$

Total number of nodes =

$$2^h - 1$$



Certainly this tree is a complete binary tree of height  $h$ .  
We shall now show that this is **the only possible** complete binary tree of height  $h$ .

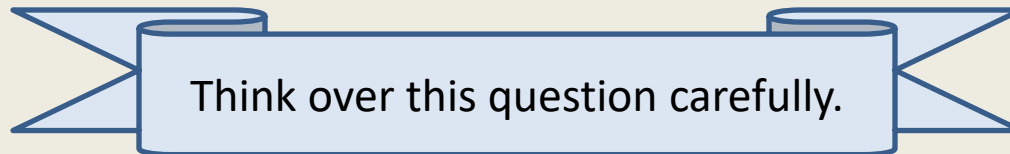
# Uniqueness of a complete binary tree of height $h$

Let  $T^*$  be the complete binary tree of height  $h$  shown in previous slide.

Notice that this is **densest** possible tree of height  $h$ .

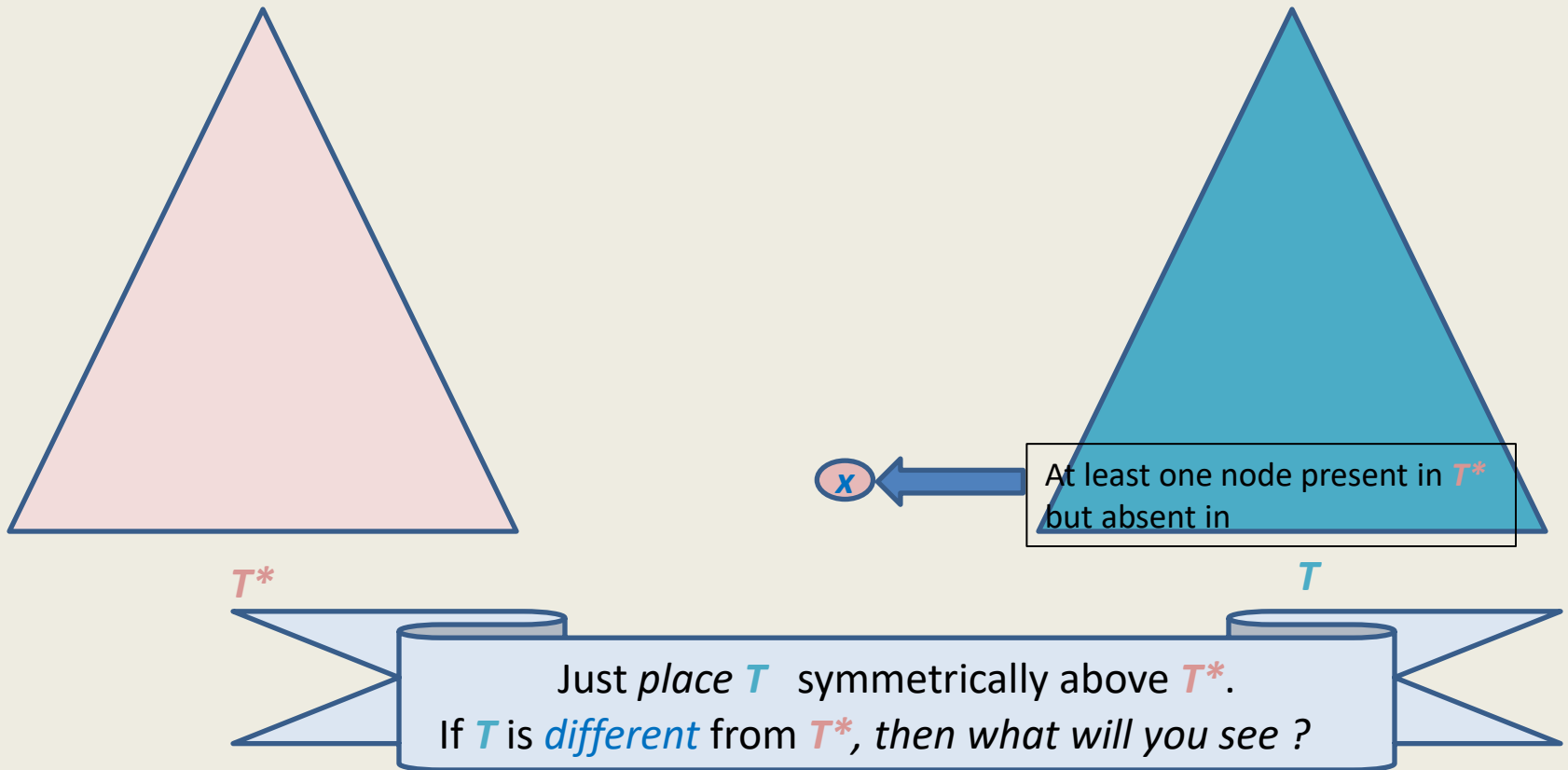
Let  $T$  be any other complete binary tree of height  $h$  different from  $T^*$ .

**Question:** How to show that  $T$  can not exist ?



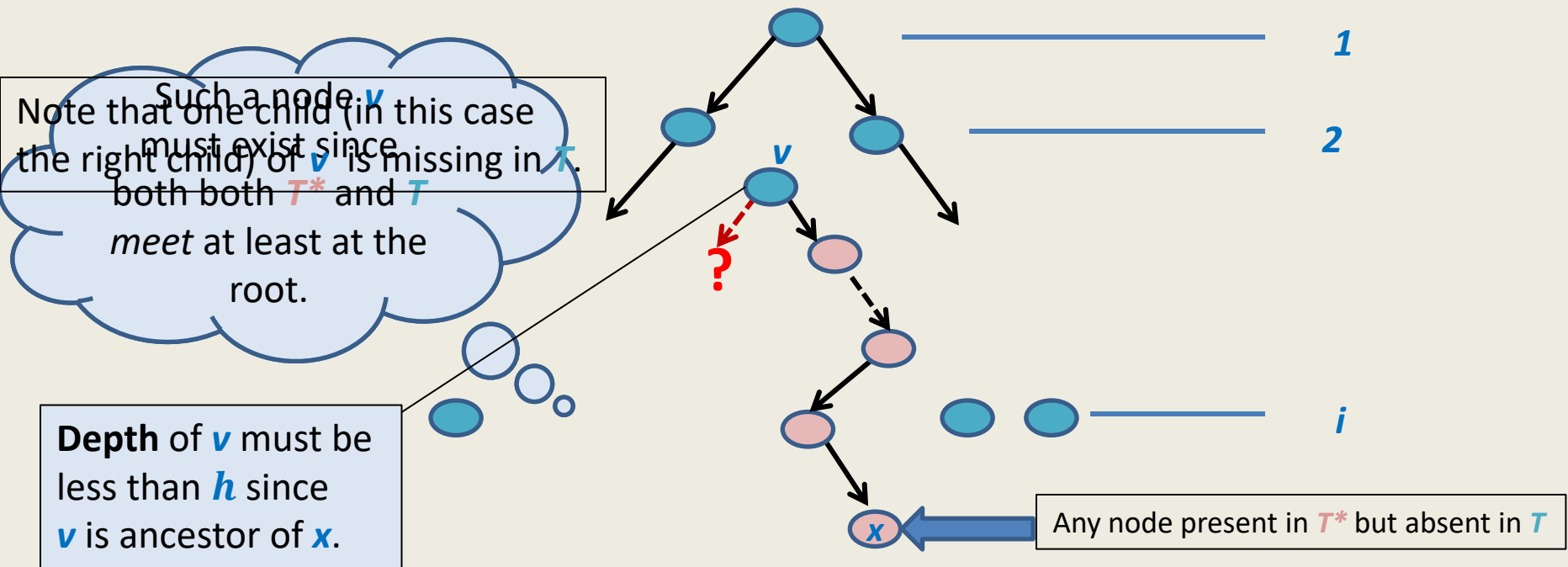
*Watch the following slide carefully.*

# Uniqueness of a complete binary tree of height $h$



# Uniqueness

## of a complete binary tree of height $h$



Since  $T$  is a full binary tree and **right child** of  $v$  is missing,

→  $v$  can not be an internal node in  $T$ .

→  $v$  **must be** leaf node.

Hence  $v$  is a leaf node of  $T$  at depth  $< h$

Hence

$T$  is not a complete binary tree of height  $h$

Hence there is no **complete** binary tree of height  $h$  different from  $T^*$ .

→ There exists a unique **complete** binary tree of height  $h$ .

**Theorem:**

A complete binary tree of height  $h$  has exactly  $2^h - 1$  nodes.

A **Red Black** Tree is height balanced

The final proof

**$T$**  : a red black tree storing  $n$  keys.

**Total number of nodes** =  $2n + 1$

$h$  : the black height

**Every leaf node is at depth  $\geq h$**

**Question:** How does  **$T$**  look like if we remove all nodes at depth  $> h$  ?

**Answer:** a complete binary tree of height  $h$

Hence  $2n + 1 \geq 2^h - 1$

$$\Rightarrow 2^h \leq 2n + 2$$

$$\Rightarrow h \leq 1 + \log_2(n + 1)$$

So Height of  **$T$**   $\leq 2h - 1 \leq 2 \log_2(n + 1) + 1$

# Analysis

## **NEARLY BALANCED BST**



# Nearly balanced Binary Search Tree

## Terminology:

**size** of a binary tree is the number of nodes present in it.

**Definition:** A binary search tree **T** is said to be nearly balanced at node **v**, if

$$\text{size}(\text{left}(\mathbf{v})) \leq \frac{3}{4} \text{size}(\mathbf{v})$$

and

$$\text{size}(\text{right}(\mathbf{v})) \leq \frac{3}{4} \text{size}(\mathbf{v})$$

**Definition:** A binary search tree **T** is said to be **nearly balanced** if it is nearly balanced at each node.

**Theorem:** Height of a **nearly balanced** BST on  $n$  nodes is  $O(\log_{4/3} n)$

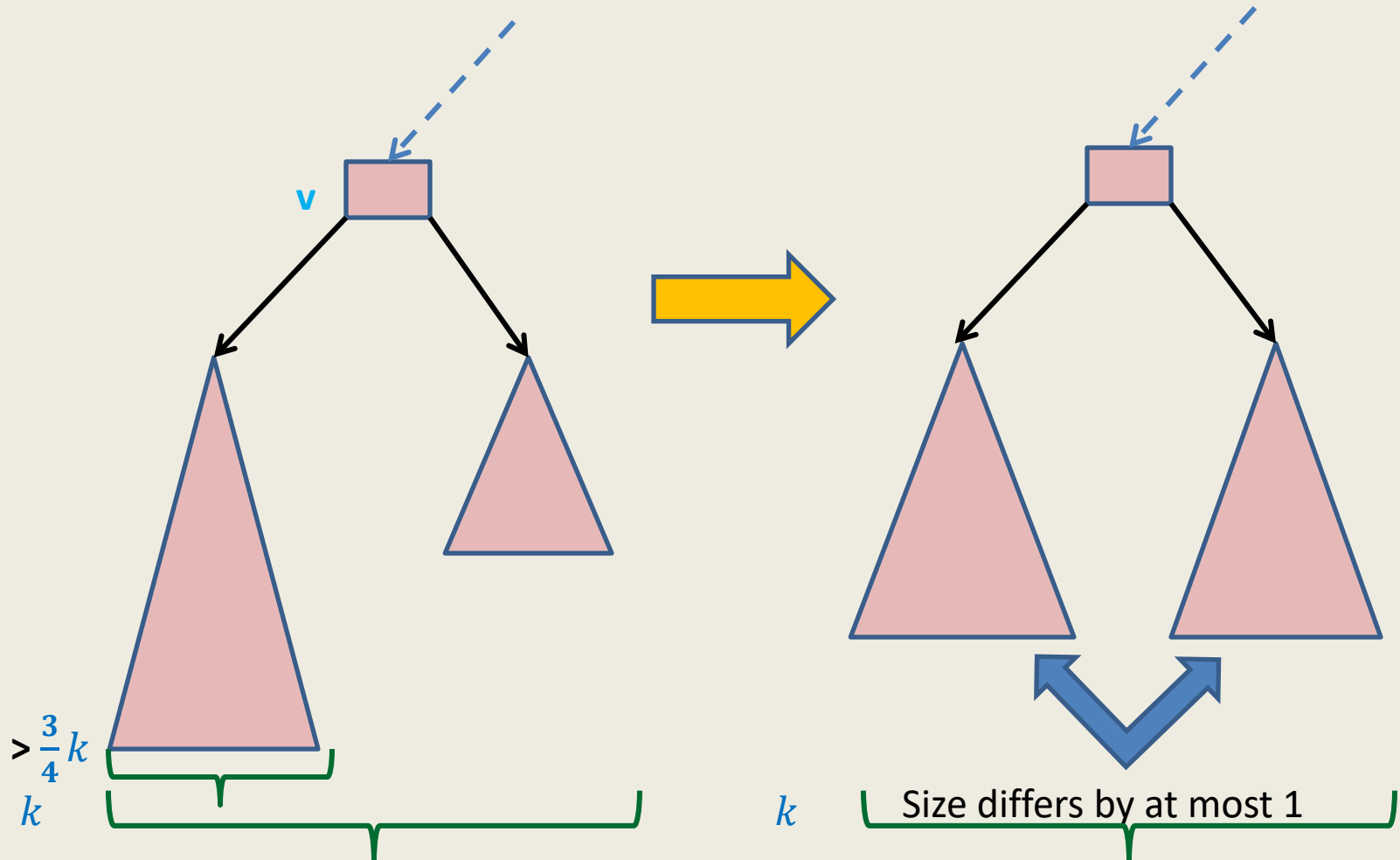
# Nearly balanced Binary Search Tree

## Maintaining under Insertion

Each node  $v$  in  $T$  maintains additional field  $\text{size}(v)$  which is the number of nodes in the  $\text{subtree}(v)$ .

- Keep  $\text{Search}(T, x)$  operation unchanged.
- Modify  $\text{Insert}(T, x)$  operation as follows:
  - Carry out normal insert and update the  $\text{size}$  fields of nodes traversed.
  - If BST  $T$  ceases to be **nearly imbalanced** at any node  $v$ , transform  $\text{subtree}(v)$  into **perfectly balanced** BST.

# “Perfectly Balancing” subtree at a node $v$



# Nearly balanced Binary Search Tree

## Observation :

It takes  $O(k)$  time to transform an imbalanced tree of size  $k$  into a perfectly balanced BST.

**Observation:** Worst case search time in nearly balanced BST is  $O(\log n)$

## Theorem:

For any arbitrary sequence of  $n$  operations, total time will be  $O(n \log n)$ .

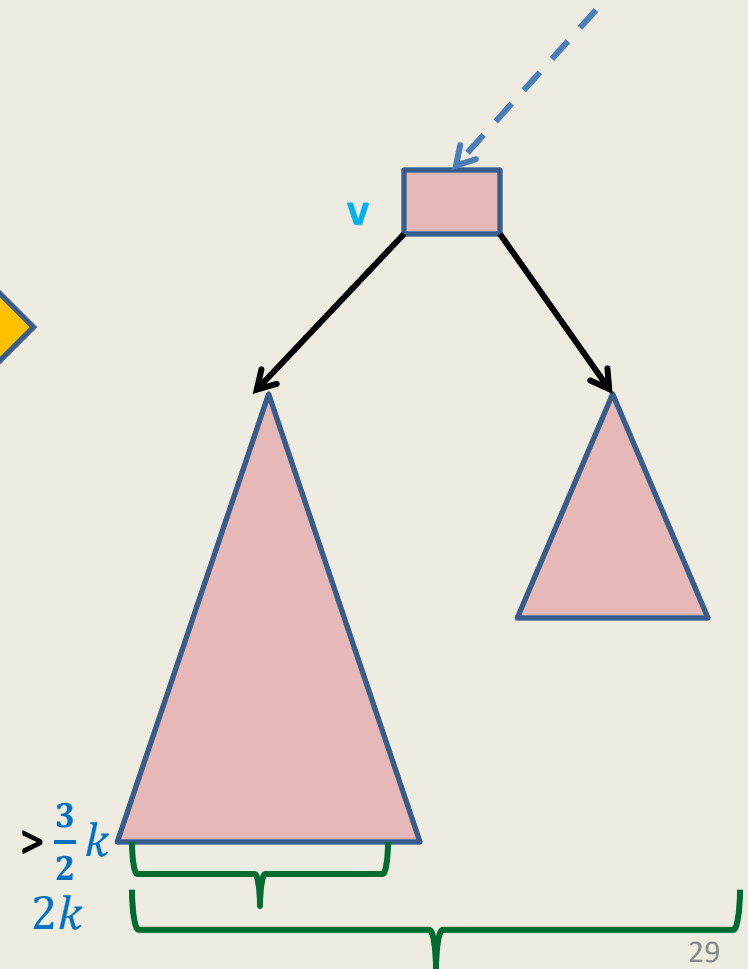
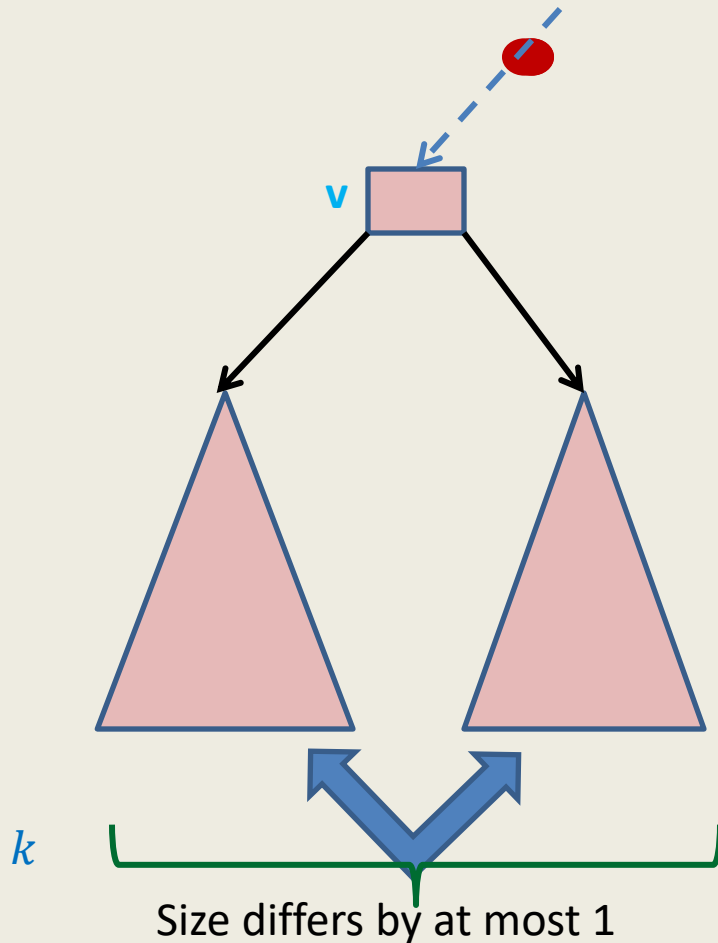
We shall now prove this theorem formally.

Watch the **next** slide slowly to get a useful insight.

How many new elements to make  $T(v)$  imbalanced ?

$\geq k$

Suppose  $T(v)$  is perfectly balanced at some moment.



# The intuition for proving the Theorem

“A perfectly balanced subtree  $T(v)$  will have to have large number of insertions before it becomes unbalanced enough to be rebuilt again.”

We shall transform this intuition into a formal proof now.

# Notations

$\text{size}_k(v)$  : no. of nodes in  $T(v)$  at the end of  $k$ th insertion.

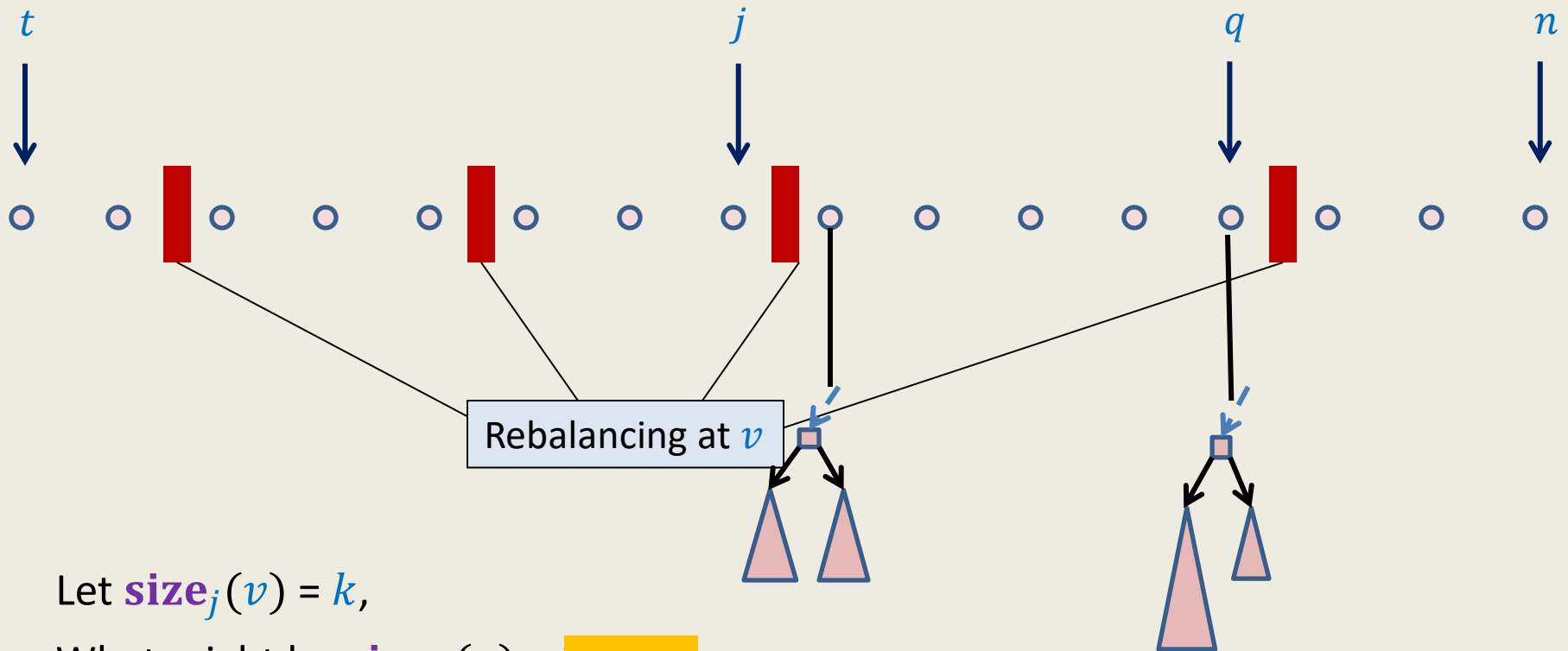
For  $k$ th insertion,

$$I_k(v) = \begin{cases} 1 & \text{if } k\text{th insertion increases } \text{size}(v) \\ 0 & \text{Otherwise} \end{cases}$$

**Question:** For a nearly balanced BST, what is

$$\sum_v I_k(v) = O(\log k)$$
$$\sum_{k=1 \text{ to } n} \sum_v I_k(v) = O(n \log n)$$

# Journey of an element/node $v$ during $n$ insertions



Let  $\text{size}_j(v) = k$ ,

What might be  $\text{size}_q(v) = \geq 2k$

Time complexity of rebalancing  $\mathbf{T}(v)$  after  $q$ th insertion  $= \mathbf{O}(k)$

What might be  $\sum_{r=j+1}^q \mathbf{I}_r(v) \geq k$

→ Time complexity of rebalancing  $\mathbf{T}(v)$  after  $q$ th insertion  $= \mathbf{O}(\sum_{r=j+1}^q \mathbf{I}_r(v))$



# Time complexity of $n$ insertions

For a vertex  $v$ ,

Time complexity of **rebalancing**  $T(v)$  during  $n$  insertions =  $\sum_{r=t}^n I_r(v)$

For all vertices,

the time complexity of **rebalancing** during  $n$  insertions =  $\sum_v \sum_{r=t_v}^n I_r(v)$

After swapping these two “summations”

$$= \sum_{k=1 \text{ to } n} \sum_v I_k(v) = O(n \log n)$$

**Theorem:**

For any arbitrary sequence of  $n$  insert operations, total time to maintain nearly balanced BST will be  $O(n \log n)$ .