# **Binary Search Trees**

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#### Maps

#### **Definition**

A **record** is a key-value pair: (k, v)

A map is an abstract data type for maintaining a set of records

- No two records can have the same key
- However, two records can have same values though
- Association of keys to values define a **mapping**: f(key) = value

#### **Examples of maps**

- UNF maintains a map of (N#, student information) records
- A social media company maintains a map of (email address, user account information) records
- An assembler maintains a symbol table (a map) of (opcode, hex) records
- A text-editor maintains a map of (color, RGB representation) records

## How to implement a map?

#### **Common map operations**

- **Insert** a record (k, v)
- **Retrieve** a record having key k
- **Delete** a record having key k

#### Approach 1: maintain a sorted list of records

- **Insertion.** will take  $O(\log n)$  time for figuring out the correct spot for the incoming record using binary search; then O(n) time for shifting items to the right to accommodate the new record; total time taken is  $O(\log n) + O(n) = O(n)$
- **Retrieval.** will take  $O(\log n)$  time using binary search
- **Deletion.** will take  $O(\log n)$  time to locate it using a binary search; then then O(n) time for left shifting items to kill the empty spot; total time taken  $O(\log n) + O(n) = O(n)$

# **How to implement a map?**

#### **Common map operations**

- **Insert** a record (k, v)
- **Retrieve** a record having key k
- Delete a record having key k

## Approach 2: maintain an unsorted list of records

- **Insertion.** will take O(1) time (add the new record at the end)
- **Retrieval.** will take O(n) time using a linear search; may need to search the whole list in the worst-case
- **Deletion.** will take O(n) time to locate it using a linear search; then O(n) time for left shifting items to kill the empty spot; total time taken O(n) + O(n) = O(n)

#### **Our aim**

#### **Common map operations**

- **Insert** a record (k, v)
- **Retrieve** a record having key k
- **Delete** a record having key k

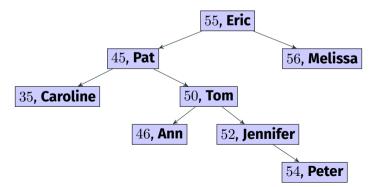
- $\blacksquare$  To accomplish the above three tasks in  $O(\log n)$  time each
- Balanced binary search trees is the solution; stay tuned ...

#### The map ADT

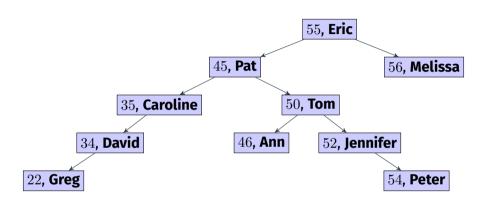
```
public interface MapADT<K,V> {
   boolean put(K k, V v); // adds a new record with key k and value v
   V remove(K k); // removes the record having key k
   V get(K k); // return the value part of the record whose key is k
   V updateValue(K k, V v); // updates the value part of the record whose key is k with a new value v
   int size(); // returns the number of records stored in the map
   void clear(); //Removes all records from the map
}
```

# What is a Binary Search Tree?

- It is a binary tree where every node contains a <key, value> pair (a record); keys
  must be comparable but the values don't need to be
- Moreover, for every node p in the tree, the following 2 properties hold
  - f 1 Keys stored in the left subtree of p are < the key stored at p
  - **2** Keys stored in the right subtree of p are > the key stored at p



# What is a Binary Search Tree?



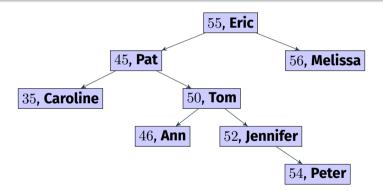
#### Use

BSTs can be used to implement **maps** and are commonly used for fast searching (typically need far less comparisons than lists)

## **Sorted Maps**

#### **BSTs** are sorted maps

An inorder traversal of a BST always gives the sorted sequence based on the keys



#### **Inorder traversal**

35, Caroline; 45, Pat; 46, Ann; 50, Tom; 52, Jennifer; 54, Peter; 55, Eric; 56, Melissa

# Searching

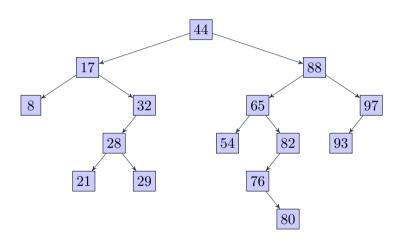
# Let's say you need to look for the record that has the key k; how will you do this?

#### **Algorithm**

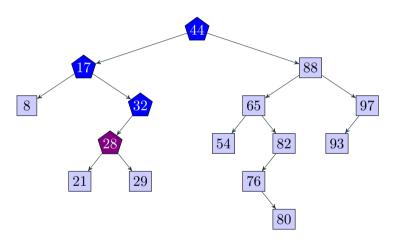
- Start at the root
- If the root's key is k, then search is successful
- ullet If  $k < {
  m root's}$  key, search recursively (or iteratively) in the left subtree of the root
- Otherwise, search recursively (or iteratively) in the right subtree of the root
- ullet If we have reached a null link, no record exists in the tree with key k

To save space in the figures, we will write only the record keys inside the nodes and avoid the corresponding values

# **Example**

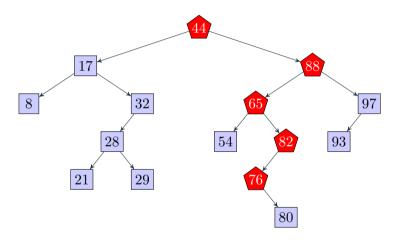


# Example: search for 28



Found!

# **Example: search for 68**



Not found; a null link is reached (the left link of 76)

# Time complexity of searching

O(h), where h is the height of the BST under consideration, since, for searching, we need to traverse a path whose length is h in the worst-case

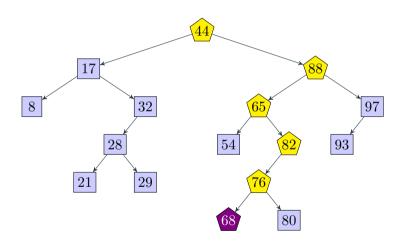
#### Insertion

# How to insert a record into a BST having key *k*?

#### **Algorithm**

- Start at the root
- If the root is null, replace empty tree with a new tree with the new record as the root, and signal **SUCCESS**
- If k equals root's key, signal **FAILURE** since a record with key k already exists
- ullet If  $k < {
  m root's}$  key, insert recursively (or, iteratively) in the left subtree of the root
- Otherwise, insert recursively (or, iteratively) in the right subtree of the root

# Example: insert 68



# Time complexity of insertion

O(h), where h is the height of the BST under consideration, since, for inserting a new record, we need to traverse a path whose length is h in the worst-case

#### **Deletion**

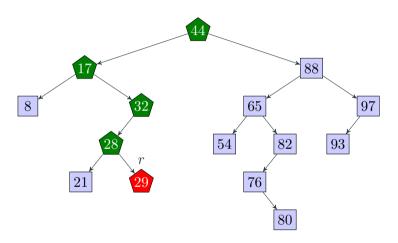
# How to delete the record from a BST having key k?

#### Idea

- ullet Traverse through the tree to find the record having key k
- If the record cannot be found, no action is needed
- Now assume that we have found the record that has key k at node r
  - $\mathbf{1}$  r is a leaf node (no child)

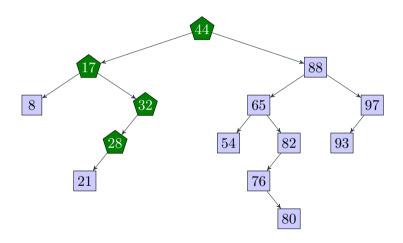
  - 3 r has two children (both left and right)

# Case 1: r is a leaf node, delete 29

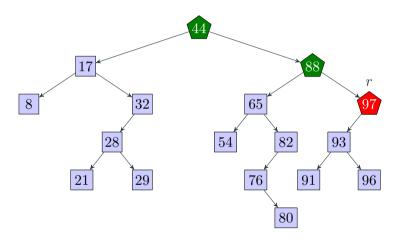


Easy! just delete it right away

# Case 1: r is a leaf node, delete 29

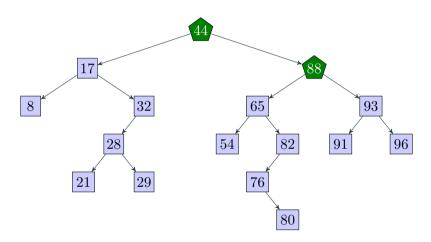


# Case 2: r has one child, delete 97

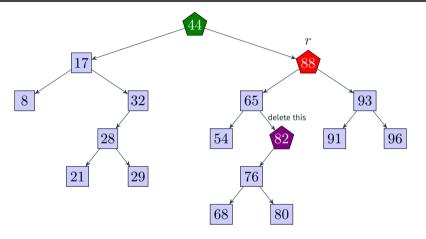


Make the parent of  $\boldsymbol{r}$  point to the only child of  $\boldsymbol{r}$  instead of  $\boldsymbol{r}$ 

# Case 2: r has one child, delete 97



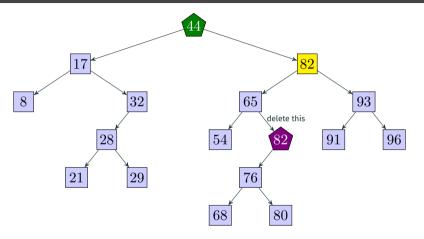
# Case 3: r has two children, delete 88



Replace the record at r with the record at the inorder predecessor p of n; then delete p using Case 1 or 2, depending on the number of children of p. Note that the inorder predecessor is either a leaf node or has a left child only (no right child).

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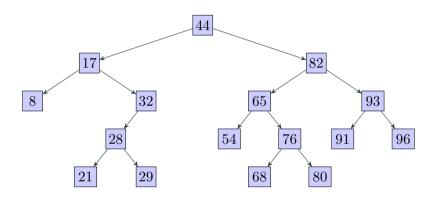
# Case 3: r has two children, delete 88



Replace the record at r with the record at the inorder predecessor p of n; then delete p using Case 1 or 2, depending on the number of children of p. Note that the inorder predecessor is either a leaf node or has a left child only (no right child).

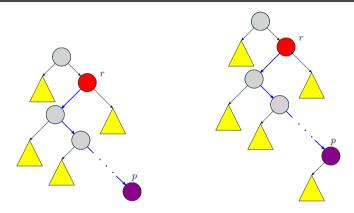
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#### Case 3: r has two children, delete 88



Replace the record at r with the record at the inorder predecessor p of n; then delete p using Case 1 or 2, depending on the number of children of p. Note that the inorder predecessor is either a leaf node or has a left child only (no right child).

# Finding the inorder predecessor in Case 3



The inorder predecessor p of node r is either a leaf node (left figure) or has a left child but no right child (right figure). So, to find p, go left of r and then keep on moving right as long as possible. It takes O(h) time to find p, where h is the height of the tree. The yellow triangles represent subtrees (some of them could be empty).

# Time complexity of deletion

- It takes O(h) time for locating the record to be deleted, where h is the height of the BST
- Case 1 takes O(1) time to delete a record
- Case 2 takes O(1) time to delete a record
- Case 3 takes O(h) time to delete a record since we need to locate the inorder predecessor p of node r in O(h) time and then delete p in O(1) time using Case 1 or 2
- So, a single deletion takes O(h) time in the worst-case

#### Code

#### A built-in record class in Java

java.util.AbstractMap.SimpleEntry<K,V>

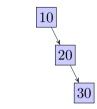
**Reference.** https://docs.oracle.com/en/java/javase/17/docs//api/java.base/java/util/AbstractMap.SimpleEntry.html

# See the class TreeMapBST

## BSTs are not unique



Insertion sequence: 20, 10, 30



Insertion sequence: 10, 20, 30



# Their structures really depend on the insertion sequence of records

# **An application**

#### **Problem**

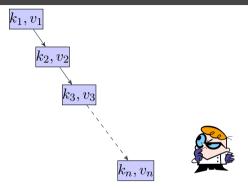
Given a text, find out the unique words in it along with their counts. We also need to output the distinct words with their count.

#### **A solution**

- Use a BST T where the key-type is String and the value-type is Integer. This means every node will store a word from the text and its count in the same text.
- $\bullet$  For every word w in the text, first check if a node exists in T , where the stored key is w
  - ullet If such a node does not exist, insert a new node in T with key w and value 1
  - If such a node exists in T, increment the stored value (essentially a counter) by 1

#### See the class WordCounter

# **Worst case scenario: skewed binary trees**



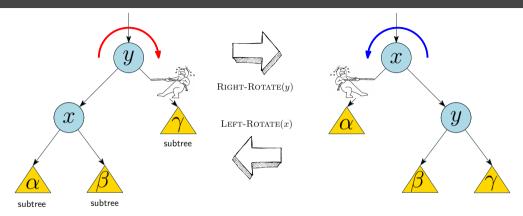
In this case,  $k_1 < k_2 < k_3 < \ldots < k_n$  and the tree is right-skewed Note that when  $k_n < k_{n-1} < \ldots < k_1$ , the tree will be left-skewed (try!) So, in the worst-case, h = n - 1 = O(n) Therefore, searching, insertion, deletion take O(h) = O(n) time each! This is as bad as using singly linked-lists!

Do something so that the height h remains under control (logarithmic) We aim for  $h = O(\log n)$ 

A solution. use **Red-Black** trees

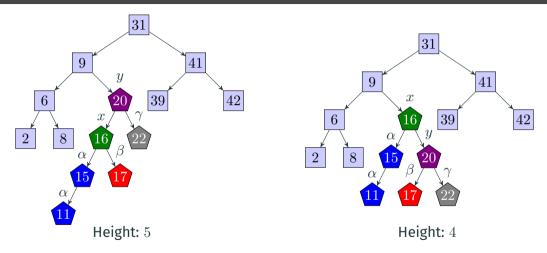
Red-Black trees are never skewed or close to being skewed unlike the plain binary search trees we just talked about

#### **Rotation on BSTs**



Rotations help in **reducing** height of BSTs; this means faster operations on BSTs A single rotation can be done in O(1) time Some of the three subtrees  $(\alpha, \beta, \gamma)$  can be empty (devoid of nodes).

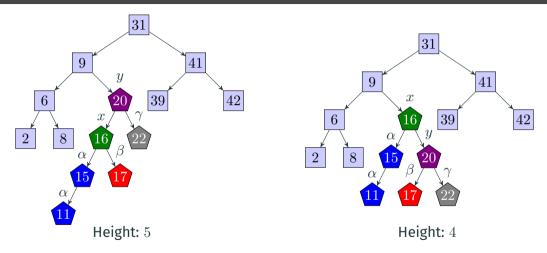
# An example of rotation



If a **right** rotation is performed at node y in the left tree, we get the tree on the right.

If a **left** rotation is performed at node x in the right tree, we get back the tree on the left.

# An example of rotation



Note that rotations never alter the inorder traversal sequences. It also means that after a rotation, the resulting binary tree is still a binary search tree.

#### **Code for rotation**

#### Right rotation at node y

```
private void rightRotateAt(Node<K,V> y) {
   Node<K,V> x = y.left;
  v.left = x.right:
  if( x.right != null )
      x.right.parent = v:
   x.parent = y.parent;
  if( v.parent == null )
      root = x:
   else if( v == v.parent.right )
     v.parent.right = x:
   else
     v.parent.left = x:
  x.right = y;
  v.parent = x:
```

#### Left rotation at node x

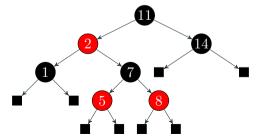
```
private void leftRotateAt(Node<K.V> x) {
 Node < K.V > v = x.right:
 x.right = y.left;
 if( v.left != null )
    v.left.parent = x;
 v.parent = x.parent:
 if( x.parent == null )
     root = v:
  else if( x == x.parent.left )
     x.parent.left = v:
 else
    x.parent.right = v:
 v.left = x:
 x.parent = y;
```

A single rotation (left/right) takes O(1) time

#### **Definition**

A **red-black** tree is a self-balancing BST (height never gets too bad) that has the following properties:

- 1 Every node is either red or black
- The root is black
- 3 If a node is **red**, then both its children are **black**. The **null** links (shown using black squares in the figure) of the leaves are **black**.
- The number of black nodes in any path from the root to a leaf is the same



#### How to maintain the colors?

Since the colors are bichromatic, a boolean variable color is enough to specify the color of a node. Set color to false, when the node is painted RED, or else, set it to true to specify the color is BLACK.

There is no magic behind these two colors; feel free to use any two colors

## A node class implementing for RB-tree

```
public class TreeMapRBTree<K extends Comparable<K>. V> implements MapADT<K.V>. Iterable< SimpleEntry<K.V> > {
  private static final boolean RED = false, BLACK = true;
  private static class Node<K. V> {
      final private K kev:
      private V val;
      private Node<K, V> left, right, parent;
      private boolean color:
      public Node(K k, V v) {
        kev = k: val = v:
        left = right = parent = null:
        color = RED:
      public String toString() {
        String colorString = (color == RED)? "RED" : "BLACK":
        if (val != null) return "<" + kev.toString() + ". " + val.toString() + ". " + colorString + ">" :
        else return kev.toString():
      // other variables and methods
```

#### Height of a red-black tree

- A RB-tree containing n records has height  $\leq 2\log_2(n+1) = O(\log n)$
- **Intuition.** By constraining the node colors on any simple path from the root to a leaf (property 4), it can be ensured that no such path is more than twice as long as any other so that the tree's height is always logarithmic.

#### **Implication**

When n = 1,000,000,  $h \le 2\log_2(n+1) < 40$ .

To search for a record, at most 41 comparisons are required in this case.

In contrast, when a plain binary search tree is used,  $h \le 999,999$ .

For searching for a record, at most 1,000,000 comparisons are required in this case.

This demonstrates the advantage of using RB-trees over plain BSTs and also shows that RB-trees can never be badly skewed (can never look like a very long linked-list).

Searching, insertion, and deletion in RB-trees take  $O(h) = O(\log n)$  time each.

# The three primary operations

- **Search**. same as the search operation for plain BSTs; takes  $O(h) = O(\log n)$  time (note that RB-trees are also BSTs, so the same search algorithm works here too!)
- **2** Insertion. we will discuss this; takes  $O(\log n)$  time
- **3 Deletion**. we won't discuss this; takes  $O(\log n)$  time

# The **TreeMap** class in Java implements RB-Tree

https://docs.oracle.com/en/java/javase/17/docs/api/java.base/java/util/TreeMap.html

#### Insertion

#### Inserting a new node z into a RB-tree

- $\bullet$  Color z red and insert it as you would in a plain BST
- 2 If necessary, start fixing the tree (using rotations and recoloring) as long as you see z's parent is red (z may change as we climb up the tree); see the cases next while (z.parent != null && z.parent.color == RED ){
  // deal with the cases inside this loop
- 3 At the end, color the root using **black**

## **Terminologies**

#### Uncle of a node

The uncle of a node is the sibling of its parent. In some cases, it could be a null link if there is no such sibling node.

```
private boolean isLeftChild( Node<K,V> node ) {
    return node.parent != null && node.parent.left == node;
}

private Node<K,V> uncle( Node<K,V> node ){
    if( node == null ) return null;
    return ( isLeftChild(node.parent) ) ? node.parent.parent.right : node.parent.left;
}
```

#### **Grandparent of a node**

The grandparent of a node is the parent of its parent. In some cases, it could be a null link if there is no such grandparent node.

```
private Node<K,V> grandParent( Node<K,V> node ){
    return node.parent.parent;
}
```

#### The six cases of insertion

- (Case 1a) z's uncle y is RED and z is a right child
- (Case 1b) z's uncle y is RED and z is a left child
- (Case 2a) z is a right child, z's parent is a left child, and z's uncle y is **BLACK**
- (Case 2b) z is a left child, z's parent is a right child, and z's uncle y is BLACK
- (Case 3a) z is a left child, z's parent is left child, and z's uncle y is **BLACK**
- (Case 3b) z is a right child, z's parent is a right child, and z's uncle y is **BLACK**

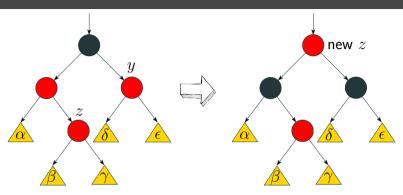
#### Things to note

- 1 The first node inserted into an empty red-black tree is always painted BLACK
- 2 Since the null links are considered to be **BLACK**, make sure that you consider their existence and color while finding the uncles of nodes

# Case 1

(Case 1a) z's uncle y is RED and z is a right child (Case 1b) z's uncle y is RED and z is a left child

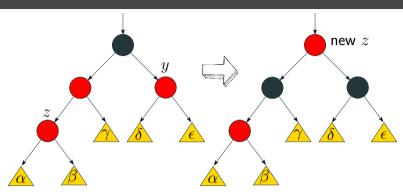
#### Case 1a



z's uncle y is **red** and z is a right child; recoloring is needed but no rotation; takes O(1) time; **continue fixing up** using the new z node

- z.parent.color = **BLACK**; y.color = **BLACK**; grandParent(z).color = **RED**;
- z = grandParent(z); // continue fixing up using the new 'z'

#### Case 1b



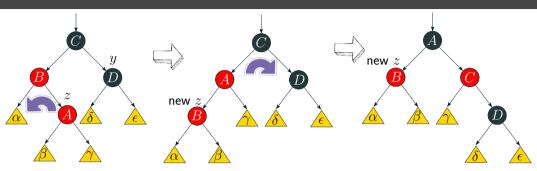
z's uncle y is **red** and z is a left child; recoloring is needed but no rotation; takes O(1) time; **continue fixing up** using the new z node

- z.parent.color = **BLACK**; y.color = **BLACK**; grandParent(z).color = **RED**;
- z = grandParent(z); // continue fixing up using the new 'z'



(Case 2a) z is a right child, z's parent is a left child, and z's uncle y is **BLACK** (Case 2b) z is a left child, z's parent is a right child, and z's uncle y is **BLACK** 

#### Case 2a



z is a right child, z's parent is a left child, and z's uncle y is **black**; recolorings + 2 rotations are needed; the fixing-up process **terminates** since z's parent is **black** after the two rotations; takes O(1) time

```
    z = z.parent; leftRotateAt(z);
    z.parent.color = BLACK; grandParent(z).color = RED; rightRotateAt(grandParent(z));
```

# Case 2b Property of the control of

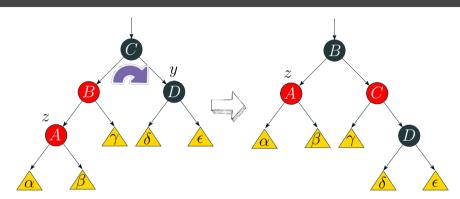
z is a left child, z's parent is a right child, and z's uncle y is **black**; recolorings + 2 rotations are needed; the fixing-up process **terminates** since z's parent is **black** after the two rotations; takes O(1) time

```
    z = z.parent; rightRotateAt(z);
    z.parent.color = BLACK; grandParent(z).color = RED; leftRotateAt(grandParent(z));
```

# Case 3

(Case 3a) z is a left child, z's parent is left child, and z's uncle y is **BLACK** (Case 3b) z is a right child, z's parent is a right child, and z's uncle y is **BLACK** 

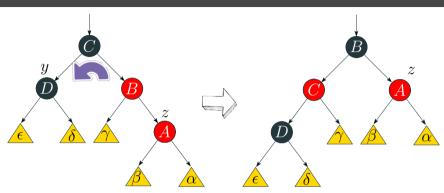
#### Case 3a



z is a left child, z's parent is left child, and z's uncle y is **black**; recolorings + 1 right rotation are needed; the fixing-up process **terminates** since z's parent is **black** after the rotation; takes O(1) time

• parent(z).color = **BLACK**; grandParent(z).color = **RED**; rightRotateAt(grandParent(z));

#### Case 3b



z is a right child, z's parent is a right child, and z's uncle y is **black**; recolorings + 1 left rotation are needed; the fixing-up process **terminates** since z's parent is **black** after the rotation; takes O(1) time

```
• parent(z).color = BLACK; grandParent(z).color = RED; leftRotateAt(grandParent(z));
```

#### Code

The second rotation of Case 2a is exactly an execution of Case 3a Similarly, the second rotation of Case 2b is exactly an execution of Case 3b

# See the class TreeMapRBTree

#### **In-browser visualization**

https://www.cs.csubak.edu/~msarr/visualizations/RedBlack.html

# Insert. 41, 38, 31, 12, 19, 8, 7, 6

The value parts of the above records are ignored in this example

Item	Case Used	Action
41	none	The only node in the tree, color it <b>black</b>
38	none	Parent is <b>black</b> ; no action is needed
31	3a	31's uncle (a null reference) is <b>black</b> ; right rotate at 41
12	1b	Recoloring is needed
19	2a	Two rotations are needed
8	1b	Recoloring is needed
7	3a	Right rotate at 12
6	1b, 3a	Two cases are needed as we climb up to fix the tree

#### **Observations**

- Time taken for inserting a new node is  $O(h) = O(\log n)$ ; after this fix-up may be needed
- During fix-up, we climb up the tree using Case 1, which only recolors but never rotates
- If we ever use Case 2 (uses 2 rotations) or 3 (uses 1 rotation), the fix-up process terminates immediately (no more handling of cases)!
- This means at every insertion of a new item, at most 2 rotations are needed
- At most h executions of Case 1 are needed plus 1 execution of Case 2/3, each taking O(1) time
- So, the total time taken for fix-up is  $(h+1) \times O(1) = O(\log n) \times O(1) = O(\log n)$ , since, for RB-trees,  $h = O(\log n)$
- Total time taken for one insertion equals time taken for inserting a new node plus total time taken for fix-up  $= O(\log n) + O(\log n) = 2 \times O(\log n) = O(\log n)$
- Time complexity of searching is  $O(h) = O(\log n)$ , since, for RB-trees,  $h = O(\log n)$
- Deletion also takes  $O(h) = O(\log n)$  time but we are not going to discuss it

# Plain BSTs vs RB-Trees: tree creation time (using n insertions)

n	RB-tree	Plain BST
10	1	1
100	1	1
1000	3	13
10000	7	214
100000	20	17249

When the input records were already **sorted** in ascending order, RB-trees could easily beat plain BSTs since the heights of the plain BSTs were exactly n-1 everywhere (much worse than the logarithmic heights of RB-trees).

n	RB-tree	Plain BST
10	1	1
100	1	1
1000	3	2
10000	7	8
100000	72	49

When the input records were in **random** order, plain BSTs performed quite as fast as RB-trees since the heights of plain BSTs were not n-1 or even close (in fact, they were almost logarithmic like RB-trees).

Times are reported in milliseconds

### Are plain BSTs completely useless?

- Plain BSTs perform **terribly** when the inputs are sorted (or, almost sorted) in ascending or descending order; the reason is in such cases, we get to see performance-degrading long paths in the tree
- But, BSTs are found to perform great on randomly ordered inputs
- ullet In those cases, h is found to be much less than n-1 and is almost logarithmic
- Consequently, we get to see fast searching, insertion, and deletion times
- Example. when n=5000, heights of plain BSTs are around 30 if the input is randomly ordered. Note that this is much less than 4999 (worst case height)

# Reading

https://opendsa-server.cs.vt.edu/ODSA/Books/Everything/html/BST.html