

# *BLM5106- Advanced Algorithm Analysis and Design*

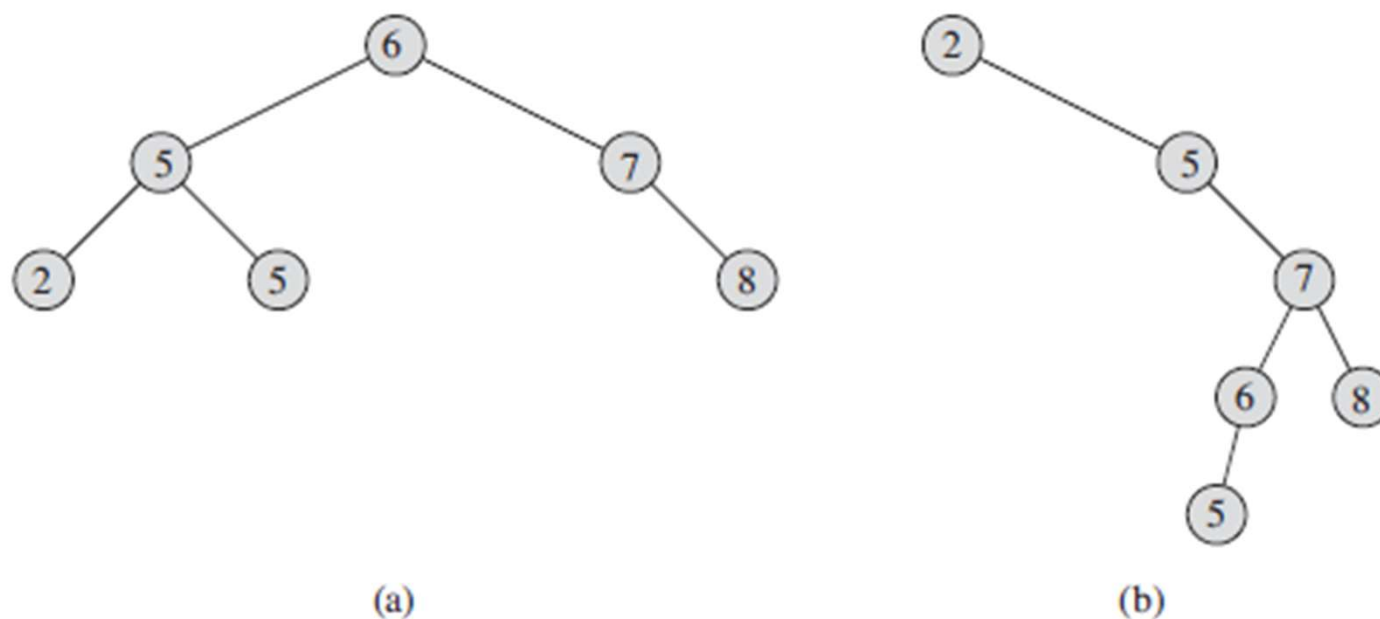
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# Binary Search Trees

The search tree data structure supports many dynamic-set operations, including SEARCH, MINIMUM, MAXIMUM, PREDECESSOR, SUCCESSOR, INSERT, and DELETE. Thus, we can use a search tree both as a dictionary and as a priority queue.

Basic operations on a binary search tree take time proportional to the height of the tree. For a complete binary tree with  $n$  nodes, such operations run in  $\Theta(\lg n)$  worst-case time. If the tree is a linear chain of  $n$  nodes, however, the same operations take  $\Theta(n)$  worst-case time.

# Binary Search Trees



**Figure 12.1** Binary search trees. For any node  $x$ , the keys in the left subtree of  $x$  are at most  $x.key$ , and the keys in the right subtree of  $x$  are at least  $x.key$ . Different binary search trees can represent the same set of values. The worst-case running time for most search-tree operations is proportional to the height of the tree. (a) A binary search tree on 6 nodes with height 2. (b) A less efficient binary search tree with height 4 that contains the same keys.

# Binary Search Trees

TREE-SEARCH( $x, k$ )

```
1  if  $x == \text{NIL}$  or  $k == x.key$ 
2      return  $x$ 
3  if  $k < x.key$ 
4      return TREE-SEARCH( $x.left, k$ )
5  else return TREE-SEARCH( $x.right, k$ )
```

# Binary Search Trees

TREE-MINIMUM( $x$ )

```
1  while  $x.left \neq \text{NIL}$ 
2       $x = x.left$ 
3  return  $x$ 
```

TREE-MAXIMUM( $x$ )

```
1  while  $x.right \neq \text{NIL}$ 
2       $x = x.right$ 
3  return  $x$ 
```

# Binary Search Trees

TREE-SUCCESSOR( $x$ )

```
1  if  $x.right \neq \text{NIL}$ 
2      return TREE-MINIMUM( $x.right$ )
3   $y = x.p$ 
4  while  $y \neq \text{NIL}$  and  $x == y.right$ 
5       $x = y$ 
6       $y = y.p$ 
7  return  $y$ 
```

# Binary Search Trees

TREE-INSERT( $T, z$ )

```
1   $y = \text{NIL}$ 
2   $x = T.\text{root}$ 
3  while  $x \neq \text{NIL}$ 
4       $y = x$ 
5      if  $z.\text{key} < x.\text{key}$ 
6           $x = x.\text{left}$ 
7      else  $x = x.\text{right}$ 
8   $z.p = y$ 
9  if  $y == \text{NIL}$ 
10      $T.\text{root} = z$       // tree  $T$  was empty
11  elseif  $z.\text{key} < y.\text{key}$ 
12      $y.\text{left} = z$ 
13  else  $y.\text{right} = z$ 
```



# Binary Search Trees

## Deletion

The overall strategy for deleting a node  $z$  from a binary search tree  $T$  has three basic cases but, as we shall see, one of the cases is a bit tricky.

- If  $z$  has no children, then we simply remove it by modifying its parent to replace  $z$  with NIL as its child.
- If  $z$  has just one child, then we elevate that child to take  $z$ 's position in the tree by modifying  $z$ 's parent to replace  $z$  by  $z$ 's child.
- If  $z$  has two children, then we find  $z$ 's successor  $y$ —which must be in  $z$ 's right subtree—and have  $y$  take  $z$ 's position in the tree. The rest of  $z$ 's original right subtree becomes  $y$ 's new right subtree, and  $z$ 's left subtree becomes  $y$ 's new left subtree. This case is the tricky one because, as we shall see, it matters whether  $y$  is  $z$ 's right child.



# Binary Search Trees

TRANSPLANT( $T, u, v$ )

```
1  if  $u.p == \text{NIL}$ 
2       $T.root = v$ 
3  elseif  $u == u.p.left$ 
4       $u.p.left = v$ 
5  else  $u.p.right = v$ 
6  if  $v \neq \text{NIL}$ 
7       $v.p = u.p$ 
```

# Binary Search Trees

TREE-DELETE( $T, z$ )

```
1  if  $z.left == \text{NIL}$ 
2      TRANSPLANT( $T, z, z.right$ )
3  elseif  $z.right == \text{NIL}$ 
4      TRANSPLANT( $T, z, z.left$ )
5  else  $y = \text{TREE-MINIMUM}(z.right)$ 
6      if  $y.p \neq z$ 
7          TRANSPLANT( $T, y, y.right$ )
8           $y.right = z.right$ 
9           $y.right.p = y$ 
10     TRANSPLANT( $T, z, y$ )
11      $y.left = z.left$ 
12      $y.left.p = y$ 
```

# Red-Black Trees

- A binary search tree of height  $h$  can support any of the basic dynamic-set operations—such as SEARCH, PREDECESSOR, SUCCESSOR, MINIMUM, MAXIMUM, INSERT, and DELETE—in  $O(h)$  time.
- Thus, the set operations are fast if the height of the search tree is small. If its height is large, however, the set operations may run no faster than with a linked list.
- Red-black trees are one of many search-tree schemes that are “balanced” in order to guarantee that basic dynamic-set operations take  $O(\lg n)$  time in the worst case.

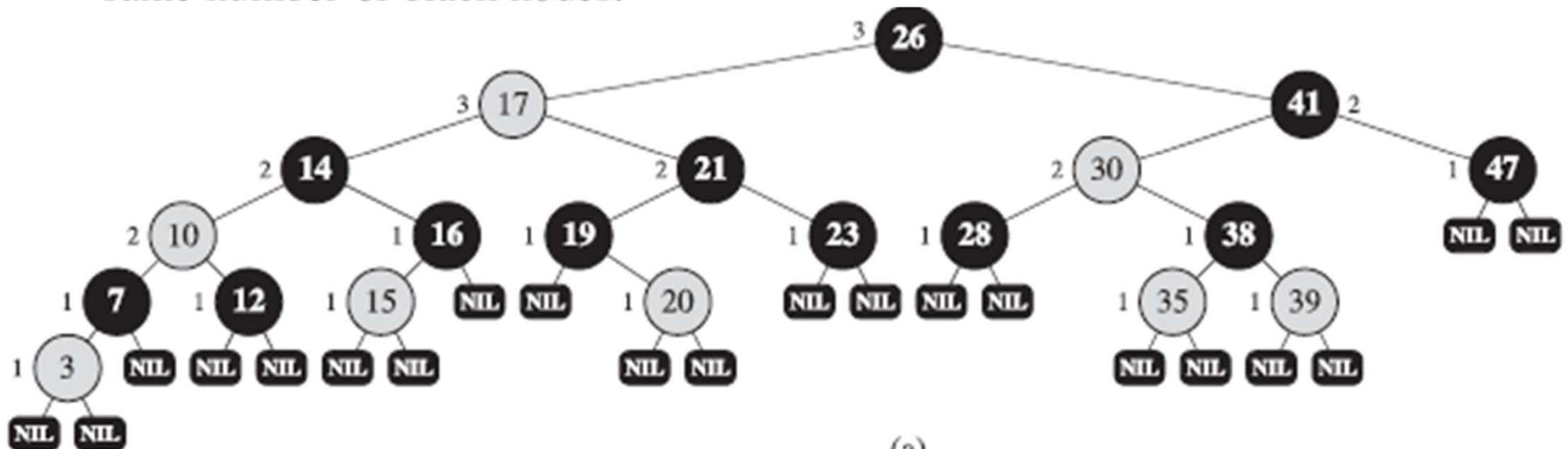
# Red-Black Trees

- A *red-black tree* is a binary search tree with one extra bit of storage per node: its *color*, which can be either RED or BLACK
- By constraining the node colors on any simple path from the root to a leaf, red-black trees ensure the tree is approximately balanced.
- Each node of the tree now contains the attributes color, key, left, right, and p. If a child or the parent of a node does not exist, the corresponding pointer attribute of the node contains the value NIL

# Red-Black Trees

A red-black tree is a binary tree that satisfies the following *red-black properties*:

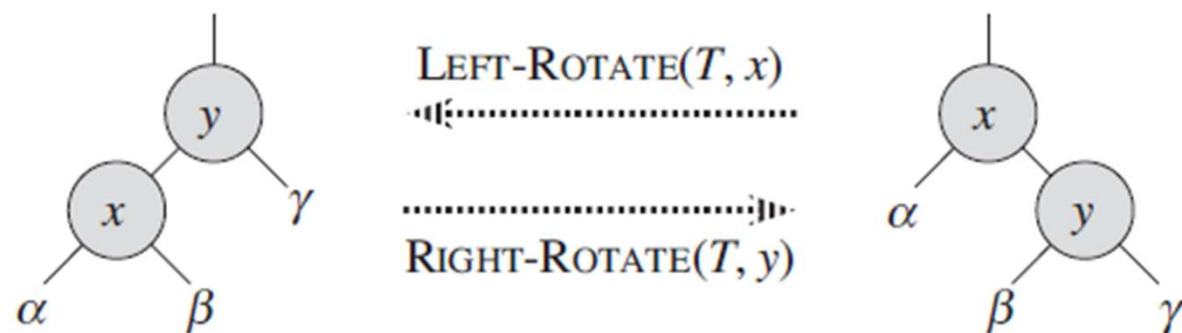
1. Every node is either red or black.
2. The root is black.
3. Every leaf (NIL) is black.
4. If a node is red, then both its children are black.
5. For each node, all simple paths from the node to descendant leaves contain the same number of black nodes.



# Red-Black Trees

- A red-black tree with  $n$  internal nodes has height at most  $2 \lg(n+1)$
- Proof?

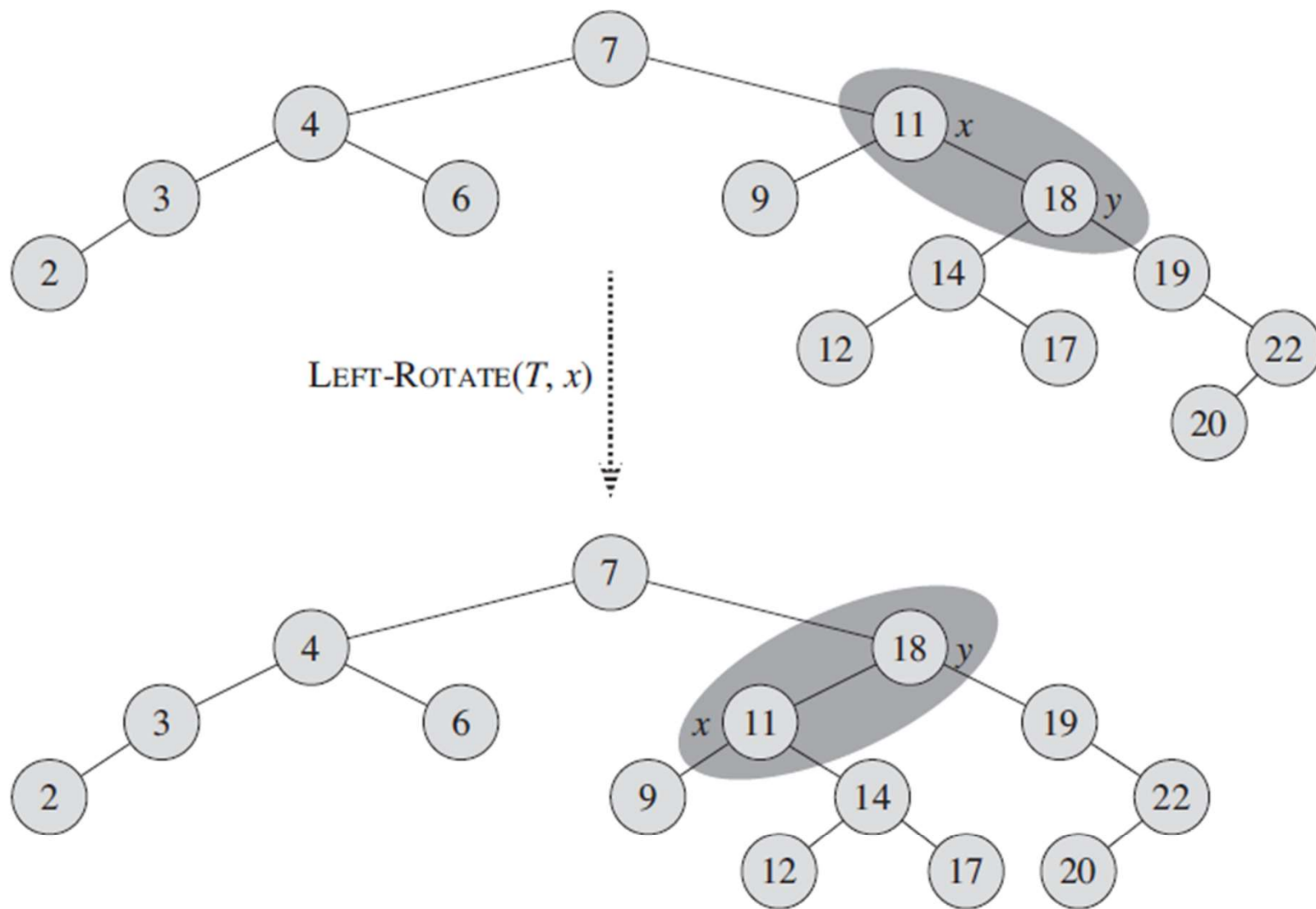
# Rotations



**Figure 13.2** The rotation operations on a binary search tree. The operation  $\text{LEFT-ROTATE}(T, x)$  transforms the configuration of the two nodes on the right into the configuration on the left by changing a constant number of pointers. The inverse operation  $\text{RIGHT-ROTATE}(T, y)$  transforms the configuration on the left into the configuration on the right. The letters  $\alpha$ ,  $\beta$ , and  $\gamma$  represent arbitrary subtrees. A rotation operation preserves the binary-search-tree property: the keys in  $\alpha$  precede  $x.\text{key}$ , which precedes the keys in  $\beta$ , which precede  $y.\text{key}$ , which precedes the keys in  $\gamma$ .



# Rotations



**Figure 13.3** An example of how the procedure  $\text{LEFT-ROTATE}(T, x)$  modifies a binary search tree. Inorder tree walks of the input tree and the modified tree produce the same listing of key values.

# RB-INSERT( $T, z$ )

```

1   $y = T.nil$ 
2   $x = T.root$ 
3  while  $x \neq T.nil$ 
4       $y = x$ 
5      if  $z.key < x.key$ 
6           $x = x.left$ 
7      else  $x = x.right$ 
8   $z.p = y$ 
9  if  $y == T.nil$ 
10      $T.root = z$ 
11  elseif  $z.key < y.key$ 
12      $y.left = z$ 
13  else  $y.right = z$ 
14   $z.left = T.nil$ 
15   $z.right = T.nil$ 
16   $z.color = RED$ 
17  RB-INSERT-FIXUP( $T, z$ )

```

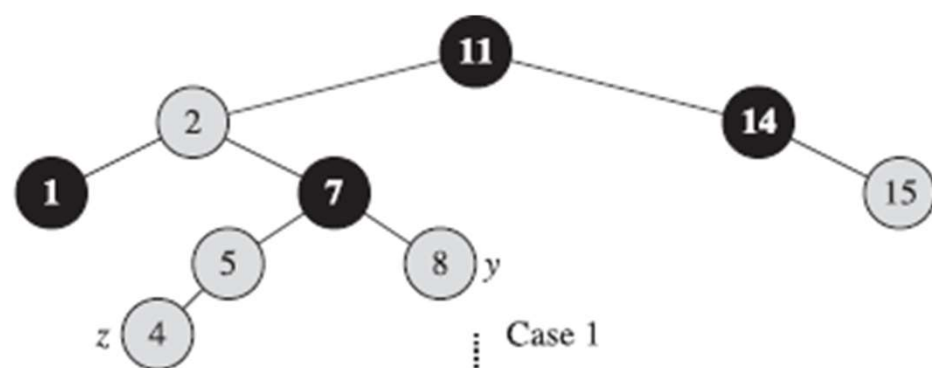
# TREE-INSERT( $T, z$ )

```

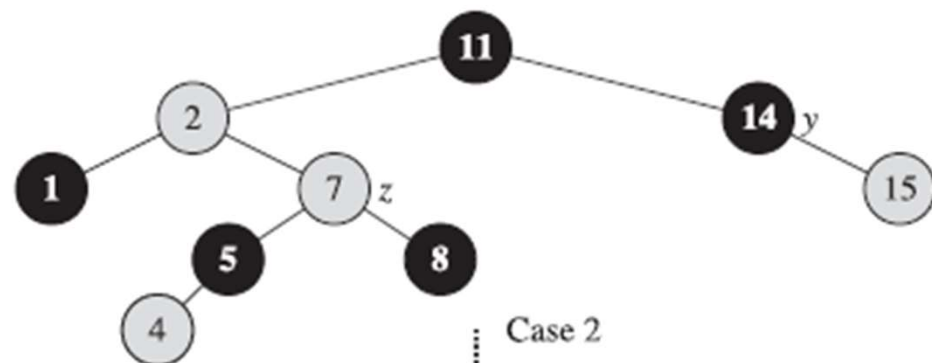
1   $y = NIL$ 
2   $x = T.root$ 
3  while  $x \neq NIL$ 
4       $y = x$ 
5      if  $z.key < x.key$ 
6           $x = x.left$ 
7      else  $x = x.right$ 
8   $z.p = y$ 
9  if  $y == NIL$ 
10      $T.root = z$  // tree  $T$  was empty
11  elseif  $z.key < y.key$ 
12      $y.left = z$ 
13  else  $y.right = z$ 

```

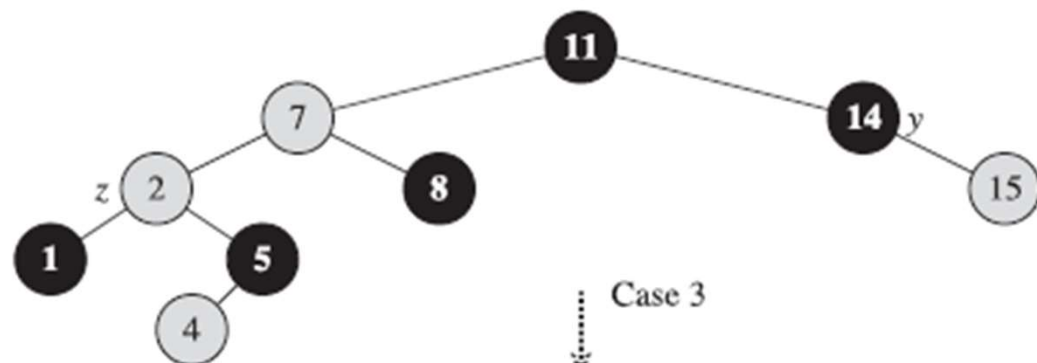
(a)



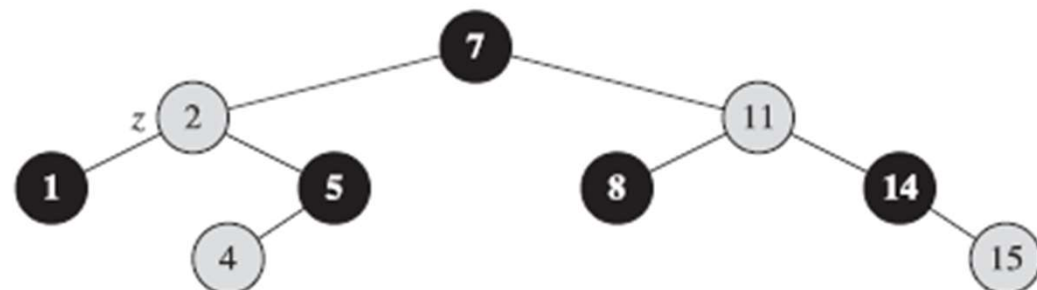
(b)



(c)



(d)

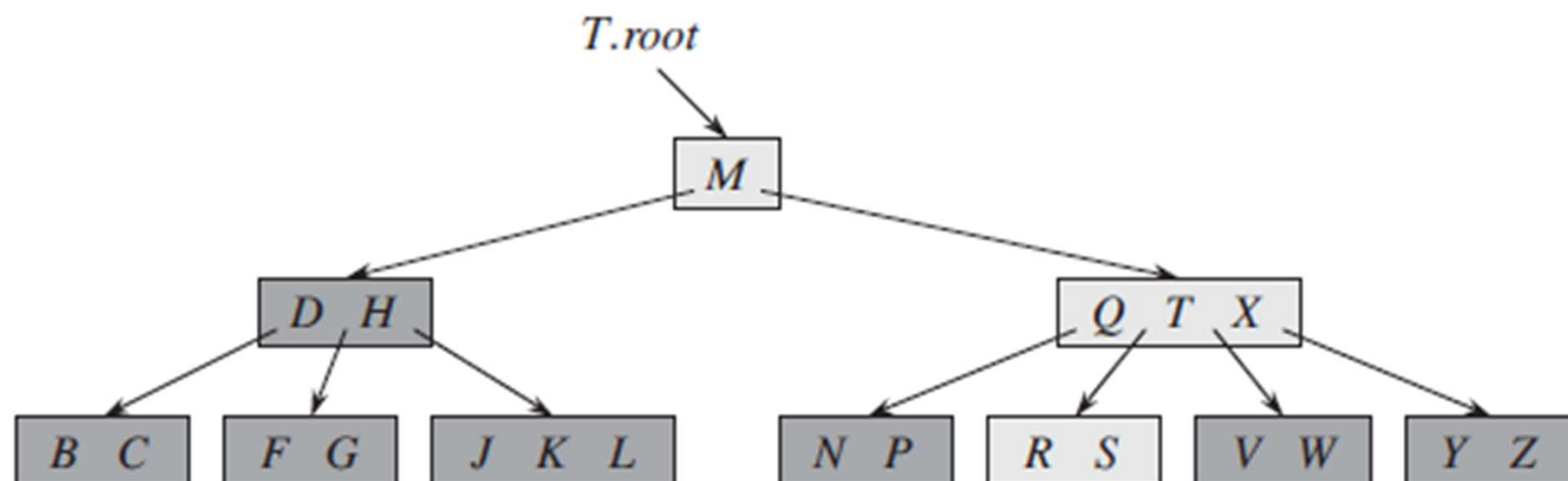


# B-Trees

- B-trees are balanced search trees designed to work well on disks or other direct access secondary storage devices.
- B-trees are similar to red-black trees but they are better at minimizing disk I/O operations.
- Many database systems use B-trees, or variants of B-trees, to store information.
- B-trees differ from red-black trees in that B-tree nodes may have many children, from a few to thousands. That is, the “branching factor” of a B-tree can be quite large, although it usually depends on characteristics of the disk unit used.
- B-trees are similar to red-black trees in that every  $n$ -node B-tree has height  $O(\lg n)$
- We can also use B-trees to implement many dynamic-set operations in time  $O(\lg n)$

# B-Tree

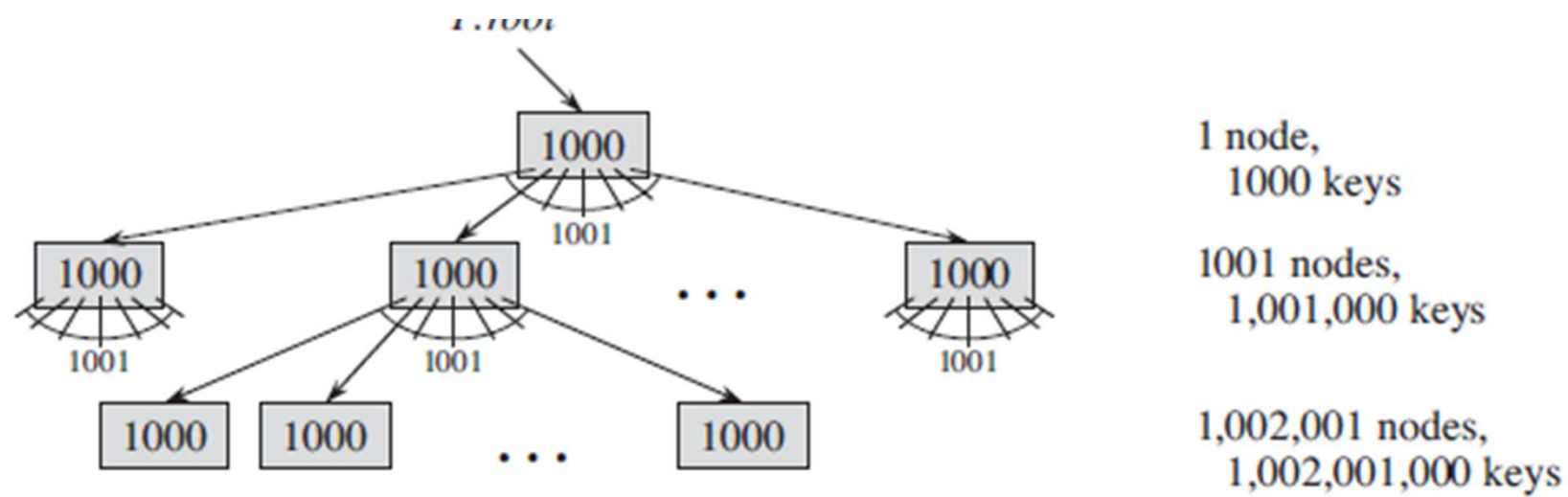
If an internal B-tree node  $x$  contains  $x.n$  keys, then  $x$  has  $x.n + 1$  children. The keys in node  $x$  serve as dividing points separating the range of keys handled by  $x$  into  $x.n + 1$  subranges, each handled by one child of  $x$ .



**Figure 18.1** A B-tree whose keys are the consonants of English. An internal node  $x$  containing  $x.n$  keys has  $x.n + 1$  children. All leaves are at the same depth in the tree. The lightly shaded nodes are examined in a search for the letter  $R$ .

# B-Tree

- In a typical B-tree application, the amount of data handled is so large that all the data do not fit into main memory at once.
- The B-tree algorithms copy selected pages from disk into main memory as needed and write back onto disk the pages that have changed.
- B-tree algorithms keep only a constant number of pages in main memory at any time; thus, the size of main memory does not limit the size of B-trees that can be handled.
- A B-tree node is usually as large as a whole disk page, and this size limits the number of children a B-tree node can have.
- For a large B-tree stored on a disk, we often see branching factors between 50 and 2000, depending on the size of a key relative to the size of a page.
- A large branching factor dramatically reduces both the height of the tree and the number of disk accesses required to find any key



**Figure 18.3** A B-tree of height 2 containing over one billion keys. Shown inside each node  $x$  is  $x.n$ , the number of keys in  $x$ . Each internal node and leaf contains 1000 keys. This B-tree has 1001 nodes at depth 1 and over one million leaves at depth 2.



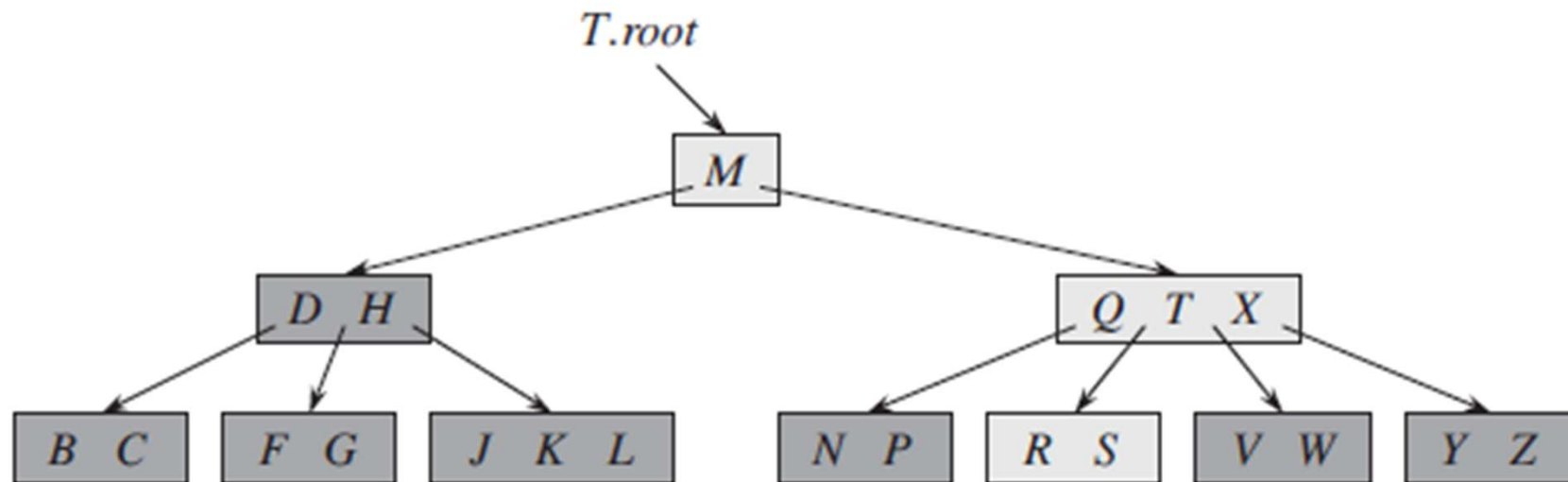
A *B-tree*  $T$  is a rooted tree (whose root is  $T.root$ ) having the following properties:

1. Every node  $x$  has the following attributes:
  - a.  $x.n$ , the number of keys currently stored in node  $x$ ,
  - b. the  $x.n$  keys themselves,  $x.key_1, x.key_2, \dots, x.key_{x.n}$ , stored in nondecreasing order, so that  $x.key_1 \leq x.key_2 \leq \dots \leq x.key_{x.n}$ ,
  - c.  $x.leaf$ , a boolean value that is TRUE if  $x$  is a leaf and FALSE if  $x$  is an internal node.
2. Each internal node  $x$  also contains  $x.n + 1$  pointers  $x.c_1, x.c_2, \dots, x.c_{x.n+1}$  to its children. Leaf nodes have no children, and so their  $c_i$  attributes are undefined.
3. The keys  $x.key_i$  separate the ranges of keys stored in each subtree: if  $k_i$  is any key stored in the subtree with root  $x.c_i$ , then

$$k_1 \leq x.key_1 \leq k_2 \leq x.key_2 \leq \dots \leq x.key_{x.n} \leq k_{x.n+1} .$$

4. All leaves have the same depth, which is the tree's height  $h$ .
5. Nodes have lower and upper bounds on the number of keys they can contain. We express these bounds in terms of a fixed integer  $t \geq 2$  called the *minimum degree* of the B-tree:
  - a. Every node other than the root must have at least  $t - 1$  keys. Every internal node other than the root thus has at least  $t$  children. If the tree is nonempty, the root must have at least one key.
  - b. Every node may contain at most  $2t - 1$  keys. Therefore, an internal node may have at most  $2t$  children. We say that a node is *full* if it contains exactly  $2t - 1$  keys.<sup>2</sup>

- What values of  $t$  makes this b-tree legal?



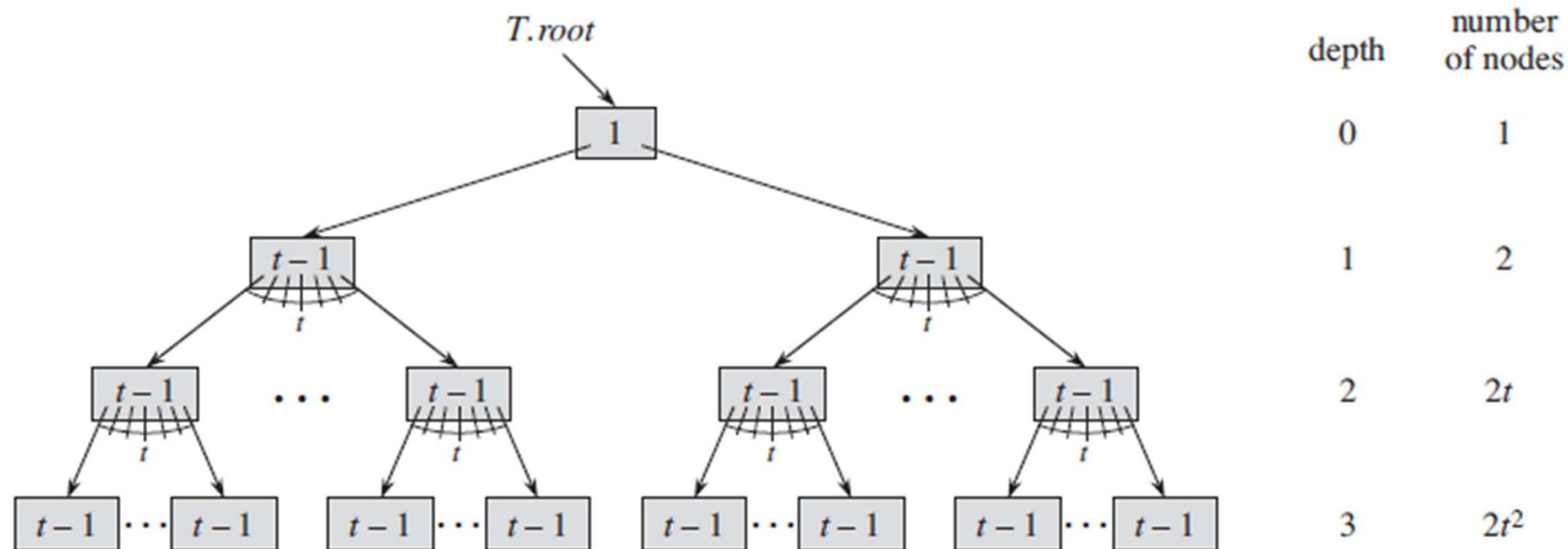
**Figure 18.1** A B-tree whose keys are the consonants of English. An internal node  $x$  containing  $x.n$  keys has  $x.n + 1$  children. All leaves are at the same depth in the tree. The lightly shaded nodes are examined in a search for the letter  $R$ .

# Proof?

## *Theorem 18.1*

If  $n \geq 1$ , then for any  $n$ -key B-tree  $T$  of height  $h$  and minimum degree  $t \geq 2$ ,

$$h \leq \log_t \frac{n+1}{2} .$$



$$\begin{aligned}
 n &\geq 1 + (t-1) \sum_{i=1}^h 2t^{i-1} \\
 &= 1 + 2(t-1) \left( \frac{t^h - 1}{t-1} \right) \\
 &= 2t^h - 1.
 \end{aligned}$$

# Basic Operations on B-Trees

- In this section, we present the details of the operations B-TREE-SEARCH, B-TREE-CREATE, and B-TREE-INSERT. In these procedures, we adopt two conventions:
  - The root of the B-tree is always in main memory, so that we never need to perform a DISK-READ on the root; we do have to perform a DISK-WRITE of the root, however, whenever the root node is changed.
  - Any nodes that are passed as parameters must already have had a DISK-READ operation performed on them.
- The procedures we present are all “one-pass” algorithms that proceed downward from the root of the tree, without having to back up.

# B-Tree Search

B-TREE-SEARCH( $x, k$ )

```
1   $i = 1$ 
2  while  $i \leq x.n$  and  $k > x.key_i$ 
3       $i = i + 1$ 
4  if  $i \leq x.n$  and  $k == x.key_i$ 
5      return  $(x, i)$ 
6  elseif  $x.leaf$ 
7      return NIL
8  else DISK-READ( $x.c_i$ )
9      return B-TREE-SEARCH( $x.c_i, k$ )
```

- See a sample
- What about complexity?



# Create an empty B-tree

To build a B-tree  $T$ , we first use B-TREE-CREATE to create an empty root node and then call B-TREE-INSERT to add new keys. Both of these procedures use an auxiliary procedure ALLOCATE-NODE, which allocates one disk page to be used as a new node in  $O(1)$  time. We can assume that a node created by ALLOCATE-NODE requires no DISK-READ, since there is as yet no useful information stored on the disk for that node.

B-TREE-CREATE( $T$ )

```
1   $x = \text{ALLOCATE-NODE}()$ 
2   $x.\text{leaf} = \text{TRUE}$ 
3   $x.n = 0$ 
4   $\text{DISK-WRITE}(x)$ 
5   $T.\text{root} = x$ 
```

B-TREE-CREATE requires  $O(1)$  disk operations and  $O(1)$  CPU time.



# Inserting a key into a B-Tree

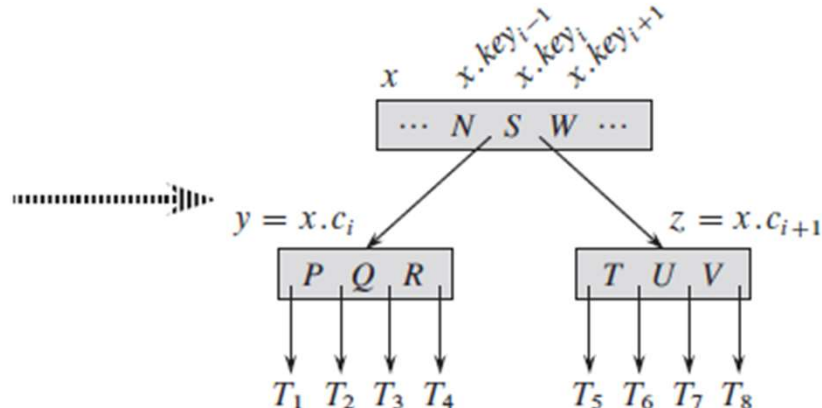
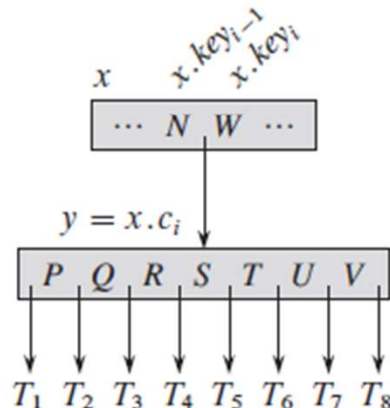
- As with binary search trees, we search for the leaf position at which to insert the new key.
- With a B-tree, however, we cannot simply create a new leaf node and insert it, as the resulting tree would fail to be a valid B-tree.
- Instead, we insert the new key into an existing leaf node. Since we cannot insert a key into a leaf node that is full, we introduce an operation that *splits* a full node  $y$  (having  $2t-1$  keys) around its *median key*  $y:key_t$  into two nodes having only  $t-1$  keys each.
- The median key moves up into  $y$ 's parent to identify the dividing point between the two new trees.
- But if  $y$ 's parent is also full, we must split it before we can insert the new key, and thus we could end up splitting full nodes all the way up the tree.

# B-TREE-SPLIT-CHILD( $x, i$ )

```

1   $z = \text{ALLOCATE-NODE}()$ 
2   $y = x.c_i$ 
3   $z.\text{leaf} = y.\text{leaf}$ 
4   $z.n = t - 1$ 
5  for  $j = 1$  to  $t - 1$ 
6       $z.\text{key}_j = y.\text{key}_{j+t}$ 
7  if not  $y.\text{leaf}$ 
8      for  $j = 1$  to  $t$ 
9           $z.c_j = y.c_{j+t}$ 
10  $y.n = t - 1$ 
11 for  $j = x.n + 1$  downto  $i + 1$ 
12      $x.c_{j+1} = x.c_j$ 
13  $x.c_{i+1} = z$ 
14 for  $j = x.n$  downto  $i$ 
15      $x.\text{key}_{j+1} = x.\text{key}_j$ 
16  $x.\text{key}_i = y.\text{key}_t$ 
17  $x.n = x.n + 1$ 
18 DISK-WRITE( $y$ )
19 DISK-WRITE( $z$ )
20 DISK-WRITE( $x$ )

```



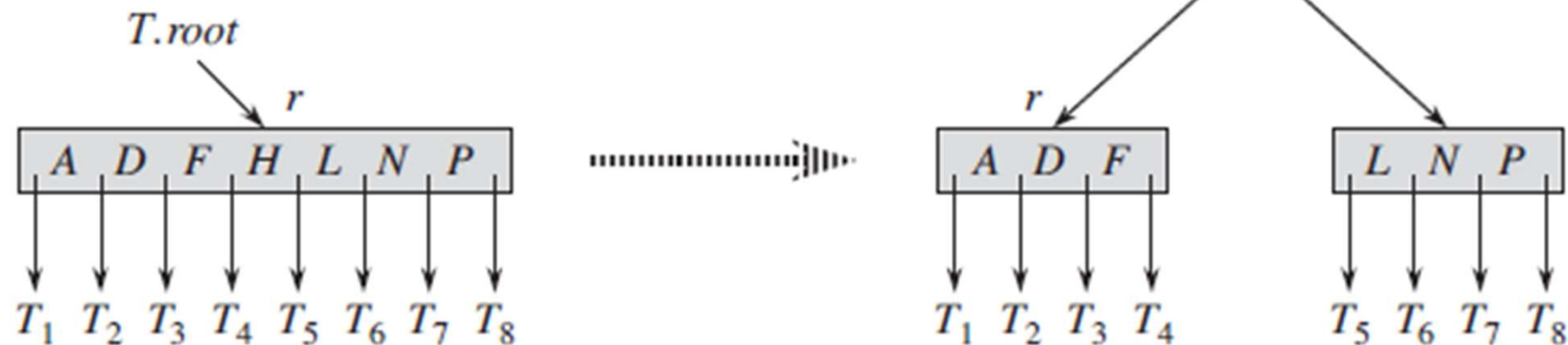
B-TREE-INSERT( $T, k$ )

```
1   $r = T.root$ 
2  if  $r.n == 2t - 1$ 
3       $s = \text{ALLOCATE-NODE}()$ 
4       $T.root = s$ 
5       $s.leaf = \text{FALSE}$ 
6       $s.n = 0$ 
7       $s.c_1 = r$ 
8      B-TREE-SPLIT-CHILD( $s, 1$ )
9      B-TREE-INSERT-NONFULL( $s, k$ )
10 else B-TREE-INSERT-NONFULL( $r, k$ )
```

The auxiliary recursive procedure B-TREE-INSERT-NONFULL inserts key  $k$  into node  $x$ , which is assumed to be nonfull when the procedure is called. The operation of B-TREE-INSERT and the recursive operation of B-TREE-INSERT-NONFULL guarantee that this assumption is true.

B-TREE-INSERT( $T, k$ )

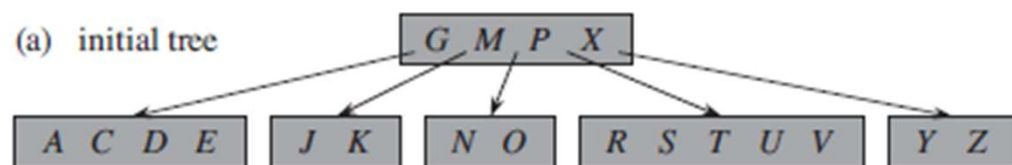
```
1   $r = T.root$ 
2  if  $r.n == 2t - 1$ 
3       $s = \text{ALLOCATE-NODE}()$ 
4       $T.root = s$ 
5       $s.leaf = \text{FALSE}$ 
6       $s.n = 0$ 
7       $s.c_1 = r$ 
8      B-TREE-SPLIT-CHILD( $s, 1$ )
9      B-TREE-INSERT-NONFULL( $s, k$ )
10 else B-TREE-INSERT-NONFULL( $r, k$ )
```



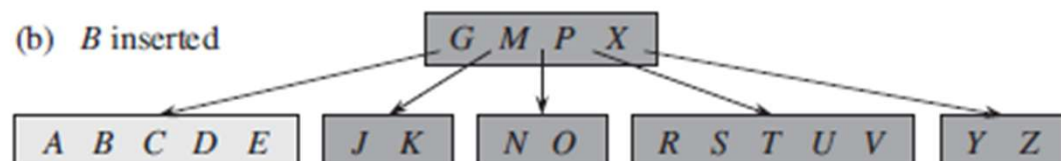
B-TREE-INSERT-NONFULL( $x, k$ )

```
1   $i = x.n$ 
2  if  $x.leaf$ 
3      while  $i \geq 1$  and  $k < x.key_i$ 
4           $x.key_{i+1} = x.key_i$ 
5           $i = i - 1$ 
6       $x.key_{i+1} = k$ 
7       $x.n = x.n + 1$ 
8      DISK-WRITE( $x$ )
9  else while  $i \geq 1$  and  $k < x.key_i$ 
10       $i = i - 1$ 
11       $i = i + 1$ 
12      DISK-READ( $x.c_i$ )
13      if  $x.c_i.n == 2t - 1$ 
14          B-TREE-SPLIT-CHILD( $x, i$ )
15          if  $k > x.key_i$ 
16               $i = i + 1$ 
17      B-TREE-INSERT-NONFULL( $x.c_i, k$ )
```

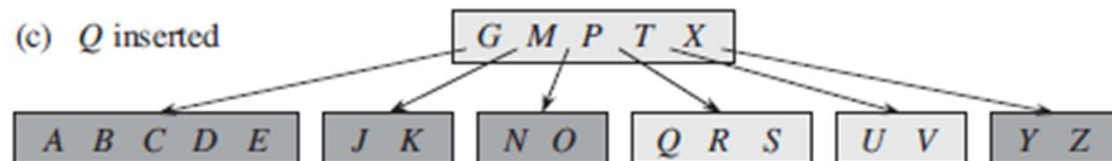
(a) initial tree



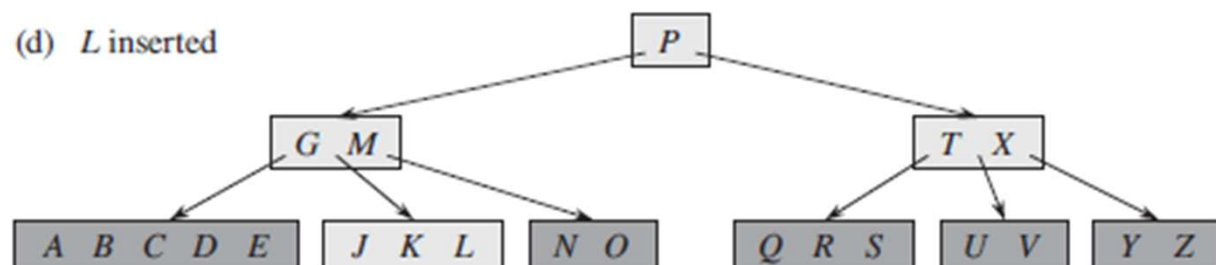
(b)  $B$  inserted



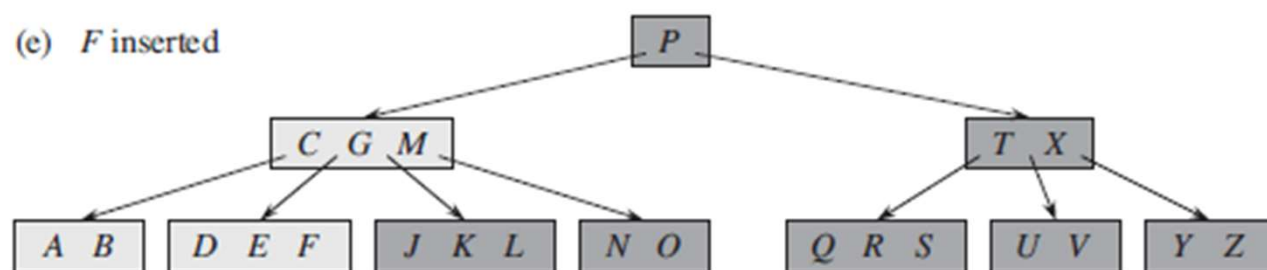
(c)  $Q$  inserted



(d)  $L$  inserted



(e)  $F$  inserted



# Deleting a key from a B-Tree

- Deletion from a B-tree is analogous to insertion but a little more complicated, because we can delete a key from any node—not just a leaf—and when we delete a key from an internal node, we will have to rearrange the node's children.
- As in insertion, we must guard against deletion producing a tree whose structure violates the B-tree properties.
- Just as we had to ensure that a node didn't get too big due to insertion, we must ensure that a node doesn't get too small during deletion (except that the root is allowed to have fewer than the minimum number  $t - 1$  of keys).



# Deleting a key from a B-Tree

