

Chapter 4

Trees

1

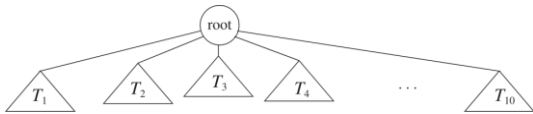
Introduction

- for large input, even linear access time may be prohibitive
- we need data structures that exhibit average running times closer to $O(\log N)$
- binary search tree

2

Terminology

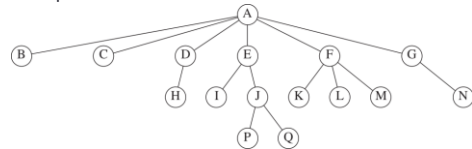
- recursive definition of **tree**
- collection of nodes (may be empty)
- distinguished node, r , is the **root**
- zero or more nonempty subtrees T_1, T_2, \dots, T_k , each of whose roots are connected by a directed **edge** from r
- root of each subtree is a **child** of r
- r is the **parent** of each subtree
- tree of N nodes has $N - 1$ edges



3

Terminology

- example tree

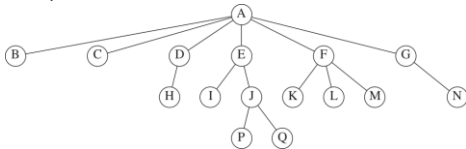


- nodes with no children are called **leaves** (e.g., B, C, H, I, P, Q, K, L, M, N)
- nodes with the same parent are **siblings** (e.g., K, L, M)
- parent**, **grandparent**, **grandchild**, **ancestor**, **descendant**, **proper ancestor**, **proper descendant**

4

Terminology

- example tree

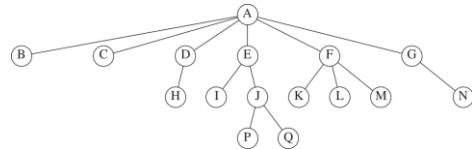


- path** from n_i to n_k is a sequence of nodes n_1, n_2, \dots, n_k where n_1 is the parent of n_{i+1} for $1 \leq i < k$
- length** of path is number of edges on path ($k - 1$)
 - path of length 0 from every node to itself
 - exactly one path from the root to each node

5

Terminology

- example tree

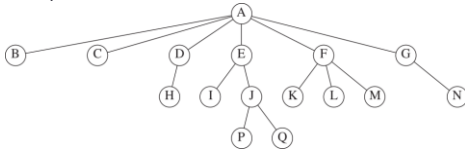


- depth** from n_i is the length of the unique path from the root to n_i
 - root is at depth 0
- height** of n_i is the length of the longest path from n_i to a leaf
 - all leaves at height 0
 - height of the tree is equal to the height of the root

6

Terminology

-example tree



- E is at depth 1 and height 2
- F is at depth 1 and height 1
- depth of tree is 3

7

Implementation of Trees

- each node could have data and a link to each child
- number of children is unknown and may be large, which could lead to wasted space
- instead, keep children in a linked list

```

1 struct TreeNode
2 {
3     Object element;
4     TreeNode *firstChild;
5     TreeNode *nextSibling;
6 };
  
```

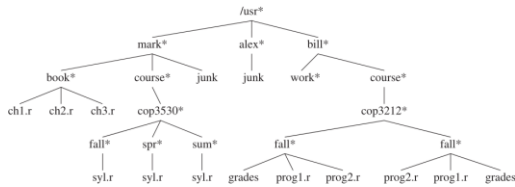


-null links not shown

8

Tree Traversals with Application

- many applications for trees
- subdirectory structure in Unix
- pathname built into tree



9

Tree Traversals with Application

- goal: list all files in a directory
- depth denoted by tabs
- begins at root

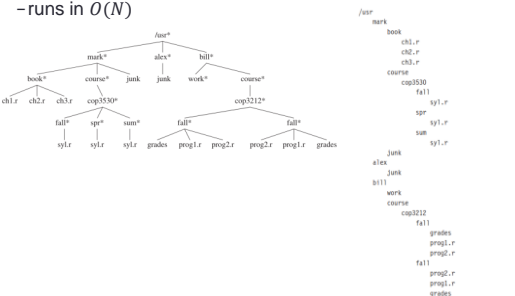
```

void FileSystem::listAll( int depth = 0 ) const
{
    1     printName( depth ); // Print the name of the object
    2     if( isDirectory( ) )
    3         for each file c in this directory (for each child)
    4         c.listAll( depth + 1 );
}
  
```

10

Tree Traversals with Application

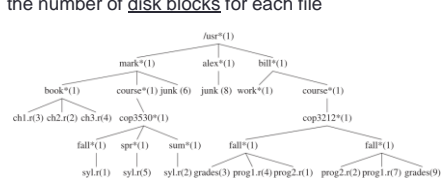
- code prints directories/files in preorder traversal
- runs in $O(N)$



11

Tree Traversals with Application

- for postorder traversal, numbers in parentheses represent the number of disk blocks for each file



12

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12

Tree Traversals with Application

- size method to find number of blocks for each file
- directories use 1 block of space

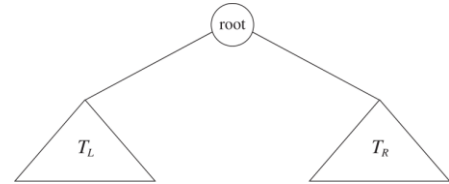
```
int FileSystem::size( ) const
{
    int totalSize = sizeOfThisFile( );
    if( isDirectory( ) )
        for each file c in this directory (for each child)
            totalSize += c.size( );
    return totalSize;
}
```

```
ch1.r 3
ch2.r 2
ch3.r 4
book 10
    syl.r 1
    fall 2
    spr 5
    syl.r 2
    sum 3
    exp9930 12
    course 13
    junk 6
    mark 30
    junk 8
    alex 9
    work 1
    grades 3
    prog1.r 4
    prog2.r 1
    fall 9
    prog1.r 2
    prog1.r 7
    grades 9
    exp9930 19
    course 30
    8011 32
    /usr 72
```

13

Binary Trees

- in binary trees, nodes can have no more than two children
- binary tree below consists of a root and two subtrees, T_L and T_R , both of which could possibly be empty



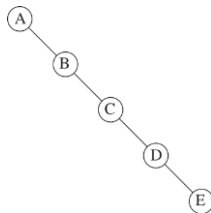
14

13

14

Binary Trees

- depth of a binary tree is considerably smaller than N
- average depth is $O(\sqrt{N})$
- average depth for a binary search tree is $O(\log N)$
- depth can be as large as $N-1$



15

15

Binary Tree Implementation

- since a binary tree has two children at most, we can keep direct links to each of them
- element plus two pointers, left and right

```
struct BinaryNode
{
    Object element; // The data in the node
    BinaryNode *left; // Left child
    BinaryNode *right; // Right child
};
```

- drawn with circles and lines (graph)
- many applications, including compiler design

16

16

Tree Traversals

- easy to list all elements of a binary search tree in sorted order
- inorder traversal
- postorder traversal
- preorder traversal
- implemented with recursive functions
- all $O(N)$

17

17

Tree Traversals

- inorder traversal

```
1 /**
2  * Print the tree contents in sorted order.
3  */
4 void printTree( ostream & out = cout ) const
5 {
6     if( isEmpty( ) )
7         out << "Empty tree" << endl;
8     else
9         printTree( root, out );
10 }
11
12 /**
13  * Internal method to print a subtree rooted at t in sorted order.
14  */
15 void printTree( BinaryNode *t, ostream & out ) const
16 {
17     if( t != nullptr )
18     {
19         printTree( t->left, out );
20         out << t->element << endl;
21         printTree( t->right, out );
22     }
23 }
```

18

18

Tree Traversals

- preorder traversal
 - visit node first, then left subtree, then right subtree
- postorder traversal
 - visit left subtree, right subtree, then node
- graphic technique for traversals
- level-order traversal
 - all nodes at depth d are processed before any node at depth $d + 1$
 - not implemented with recursion
 - queue

19

Tree Traversals

- height method using postorder traversal

```

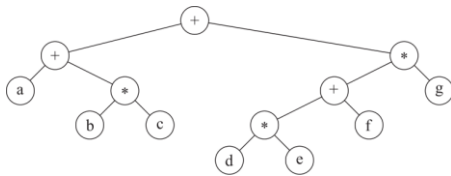
1  /**
2   * Internal method to compute the height of a subtree rooted at t.
3   */
4   int height( BinaryNode *t )
5   {
6       if( t == nullptr )
7           return -1;
8       else
9           return 1 + max( height( t->left ), height( t->right ) );
10  }

```

20

Binary Tree Example: Expression Trees

- expression tree
 - leaves represent operands (constants or variable names)
 - interior nodes represent operators
 - binary tree since most operators are binary, but not required
 - some operations are unary

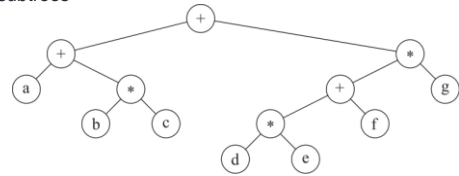


21

21

Binary Tree Example: Expression Trees

- evaluate expression tree, T , by applying operator at root to values obtained by recursively evaluating left and right subtrees



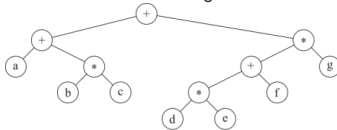
- left subtree: $a + (b * c)$
- right subtree: $((d * e) + f) * g$
- complete tree: $(a + (b * c)) + (((d * e) + f) * g)$

22

22

Binary Tree Example: Expression Trees

- inorder traversal
 - recursively produce left expression
 - print operator at root
 - recursively produce right expression
- postorder traversal
 - result: $a b c * + d e * f + g * +$
- preorder traversal
 - result: $++ a * b c * + * d e f g$



23

23

Binary Tree Example: Expression Trees

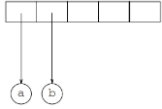
- goal: convert a postorder expression into an expression tree
 - read expression one symbol at a time
 - if operand, create node and push a pointer to it on the stack
 - if operator, pop pointers to two trees T_1 and T_2 from the stack
 - form new tree with operator as root
 - pointer to this tree is then pushed on the stack

24

24

Binary Tree Example: Expression Trees

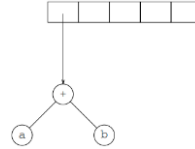
- example: $a b + c d e + * *$
- first two symbols are operands and are pushed on the stack



25

Binary Tree Example: Expression Trees

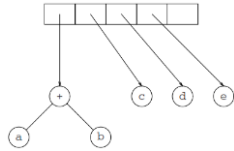
- example: $a b + c d e + * *$
- after + is read, two pointers are popped and new tree formed with a pointer pushed on the stack



26

Binary Tree Example: Expression Trees

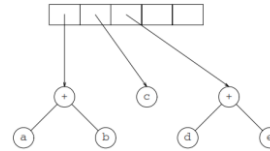
- example: $a b + c d e + * *$
- next, c, d, and e are read, with one-node tree created for each and pushed on the stack



27

Binary Tree Example: Expression Trees

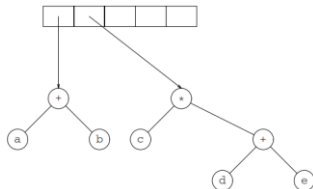
- example: $a b + c d e + * *$
- after + is read, two trees are merged



28

Binary Tree Example: Expression Trees

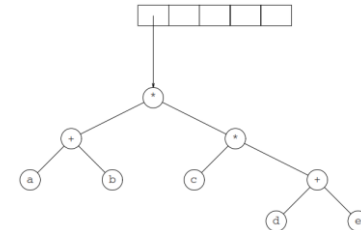
- example: $a b + c d e + * *$
- after * is read, two trees are popped to form a new tree with a * as root



29

Binary Tree Example: Expression Trees

- example: $a b + c d e + * *$
- finally * is read, two trees are popped to form a final tree, which is left on the stack



30

Binary Search Tree ADT

- binary trees often used for searching
- assume each node in the tree stores one element (integer)
- binary search tree
 - for every node X in the tree
 - all items in left subtree are smaller than X
 - all items in right subtree are greater than X
 - items in tree must be order-able



31

31

Binary Search Tree ADT

- common operations on binary search trees
 - often written recursively
 - since average depth is $O(\log N)$, no worry about stack space
- binary search tree interface
 - searching depends on $<$ operator, which must be defined for Comparable type
 - only data member is root pointer

32

32

Binary Search Tree ADT

```

1  template <typename Comparable>
2  class BinarySearchTree
3  {
4  public:
5      BinarySearchTree();
6      BinarySearchTree(const BinarySearchTree & rhs);
7      BinarySearchTree(BinarySearchTree && rhs);
8      ~BinarySearchTree();
9
10     const Comparable & findMin() const;
11     const Comparable & findMax() const;
12     bool contains(const Comparable & x) const;
13     bool isEmpty() const;
14     void printTree(ostream & out = cout) const;
15
16     void makeEmpty();
17     void insert(const Comparable & x);
18     void insert(Comparable && x);
19     void remove(const Comparable & x);
20
21     BinarySearchTree & operator=(const BinarySearchTree & rhs);
22     BinarySearchTree & operator=(BinarySearchTree && rhs);
  
```

33

33

Binary Search Tree ADT

```

24 private:
25     struct BinaryNode
26     {
27         Comparable element;
28         BinaryNode *left;
29         BinaryNode *right;
30
31         BinaryNode(const Comparable & theElement, BinaryNode *lt, BinaryNode *rt)
32             : element(theElement), left(lt), right(rt) {}
33
34         BinaryNode(Comparable && theElement, BinaryNode *lt, BinaryNode *rt)
35             : element(std::move(theElement)), left(lt), right(rt) {}
36     };
37
38     BinaryNode *root;
39
40     void insert(const Comparable & x, BinaryNode * & t);
41     void insert(Comparable && x, BinaryNode * & t);
42     void remove(const Comparable & x, BinaryNode * & t);
43     BinaryNode * findMin(BinaryNode * t) const;
44     BinaryNode * findMax(BinaryNode * t) const;
45     bool contains(const Comparable & x, BinaryNode * t) const;
46     void makeEmpty(BinaryNode * & t);
47     void printTree(BinaryNode * t, ostream & out) const;
48     BinaryNode * clone(BinaryNode * t) const;
49
  
```

34

34

Binary Search Tree ADT

- test for item in subtree

```

1  /**
2   * Internal method to test if an item is in a subtree.
3   * x is item to search for.
4   * t is the node that roots the subtree.
5   */
6  bool contains(const Comparable & x, BinaryNode * t) const
7  {
8      if(t == nullptr)
9          return false;
10     else if(x < t->element)
11         return contains(x, t->left);
12     else if(t->element < x)
13         return contains(x, t->right);
14     else
15         return true; // Match
16 }
  
```

35

35

Binary Search Tree ADT

- findMin** and **findMax**

- private methods return pointer to smallest/largest elements in the tree
- to find the minimum, start at the root and go left as long as possible
- similar for finding the maximum

36

36

Binary Search Tree ADT

-recursive version of `findMin`

```

1  /**
2   * Internal method to find the smallest item in a subtree t.
3   * Return node containing the smallest item.
4   */
5  BinaryNode * findMin( BinaryNode *t ) const
6  {
7      if( t == nullptr )
8          return nullptr;
9      if( t->left == nullptr )
10         return t;
11     return findMin( t->left );
12 }

```

37

Binary Search Tree ADT

-nonrecursive version of `findMax`

```

1  /**
2   * Internal method to find the largest item in a subtree t.
3   * Return node containing the largest item.
4   */
5  BinaryNode * findMax( BinaryNode *t ) const
6  {
7      if( t != nullptr )
8          while( t->right != nullptr )
9              t = t->right;
10     return t;
11 }

```

38

Binary Search Tree ADT

-insertion for binary search trees

- to insert X into tree T , proceed down the tree, as in the **contains** function
- if X is found, do nothing
- otherwise, insert X at the last spot on the path traversed
- example: insert 5 into binary search tree



39

Binary Search Tree ADT

- duplicates can be handled by adding a count to the node record
- better than inserting duplicates in tree
- may not work well if key is only small part of larger structure

40

Binary Search Tree ADT

- deletion in binary search tree may be difficult
- multiple cases
 - if node is leaf, it can be deleted immediately
 - if node has only one child, node can be deleted after its parent adjusts a link to bypass the node



41

Binary Search Tree ADT

- multiple cases (cont.)
 - complicated case: node with two children
 - replace data of this node with smallest data of right subtree and recursively delete the node
 - since smallest node in right subtree cannot have a left child, the second remove is easy



42

Binary Search Tree ADT

- if number of deletions small, lazy deletion may be used
 - node is marked deleted rather than actually being deleted
- especially popular if duplicates allowed
 - count of duplicates can be decremented
- incurs only small penalty on tree since height not affected greatly
- if deleted node reinstated, some benefits

43

43

Binary Search Tree Average-Case Analysis

- we expect most operations on binary search trees will have $O(\log N)$ time
 - average depth over all nodes can be shown to be $O(\log N)$
- all insertions and deletions must be equally likely
- sum of the depths of all nodes in a tree is known as internal path length

44

44

Binary Search Tree Average-Case Analysis

- the run time of binary search trees depends on the depth of the tree, which in turn depends on the order that the keys are inserted
- let $D(N)$ be the internal path length for a tree of N nodes
- we know that $D(1) = 0$
- a tree of an i -node left subtree and an $(N - i - 1)$ -node right subtree, plus a root at depth zero for $0 \leq i \leq N$
- total number of nodes in tree = left subtree + right subtree + 1
- all nodes except the root are one level deeper,

$$D(N) = D(i) + D(N - i - 1) + N - 1$$

45

45

Binary Search Tree Average-Case Analysis

- if all subtree sizes are equally likely, then the average for each subtree is

$$\left(\frac{1}{N}\right) \sum_{j=1}^{N-1} D(j)$$

therefore, for the total number of nodes

$$D(N) = \left(\frac{2}{N}\right) \left[\sum_{j=1}^{N-1} D(j) \right] + N - 1$$

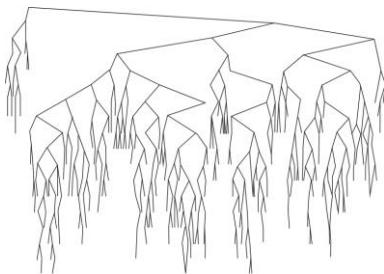
- once this recurrence relation is evaluated, the result is $D(N) = O(N \log N)$
- and the average number of nodes is $O(\log N)$

46

46

Binary Search Tree Average-Case Analysis

- example: randomly generated 500-node tree has expected depth of 9.98



47

47

Binary Search Tree Average-Case Analysis

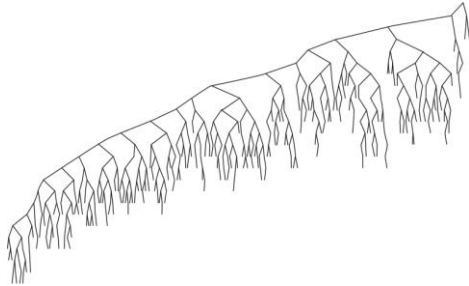
- deletions, however, bias the left subtrees to be longer because we always replace a deleted node with a node from the right subtree
- exact effect of deletions still unknown
- if insertions and deletions are alternated $\Theta(N^2)$ times, then expected depth is $\Theta(\sqrt{N})$

48

48

Binary Search Tree Average-Case Analysis

- after 250,000 random insert/delete pairs, tree becomes unbalanced, with depth = 12.51



49

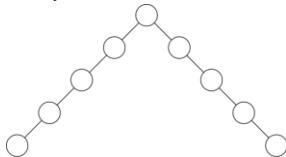
Binary Search Tree Average-Case Analysis

- could randomly choose between smallest element in the right subtree and largest element in the left subtree when replacing deleted element
 - should keep bias low, but not yet proven
- bias does not show up for small trees
- if $o(N^2)$ insert/remove pairs used, tree actually gains balance
- average case analysis extremely difficult
- two possible solutions
 - balanced trees
 - self-adjusting trees

50

AVL Trees

- Adelson-Velskii and Landis (AVL) tree is a binary search tree with a balance condition
- balance condition in general
 - must be easy to maintain
 - ensures depth of tree is $O(\log N)$
- simplest idea: left and right subtrees have the same height
 - does not always work



51

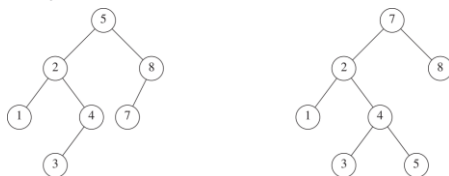
AVL Trees

- alternate balance condition: every node must have left and right subtrees of the same height
 - only perfectly balanced trees of $2^k - 1$ nodes would work
 - condition too rigid

52

AVL Trees

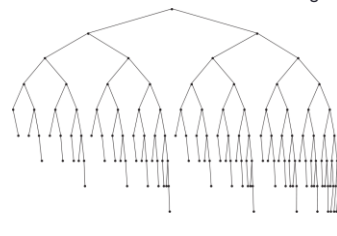
- AVL tree
 - for each node in the tree, height of left and right subtrees differ by at most 1
 - height balance = height of right subtree - height of left
 - height of an empty tree: -1
 - height information kept in the node structure



53

AVL Trees

- example AVL tree
 - fewest nodes for a tree of height 9
 - left subtree contains fewest nodes for height 7
 - right subtree contains fewest nodes for height 8



54

AVL Trees

- minimum number of nodes, $S(h)$, in an AVL tree of height h

$$S(h) = S(h-1) + S(h-2) + 1 \quad S(0) = 1, \quad S(1) = 2$$

- closely related to Fibonacci numbers
- all operations can be performed in $O(\log N)$ time, except insertion and deletion

55

55

AVL Trees

- insertion
 - update all balance information in the nodes on the path back to the root
 - could violate the balance condition
 - rotations used to restore the balance property
- deletion
 - perform same promotion as in a binary search tree, updating the balance information as necessary
 - same balancing operations for insertion can then be used

56

56

AVL Trees

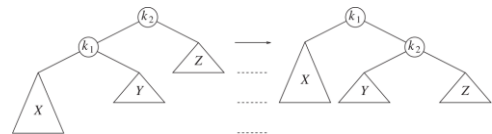
- if α is the node requiring rebalancing (the heights of its left and right subtrees differ by 2), the violation occurred in one of four cases
 - an insertion into the left subtree of the left child of α
 - an insertion into the right subtree of the left child of α
 - an insertion into the left subtree of the right child of α
 - an insertion into the right subtree of the right child of α
- cases 1 and 4 are mirror image symmetries with respect to α and can be resolved with a single rotation
- cases 2 and 3 are mirror image symmetries with respect to α and can be resolved with a double rotation

57

57

AVL Trees

- single rotation
 - only possible case 1 scenario
 - to balance, imagine "picking up" tree by k_1



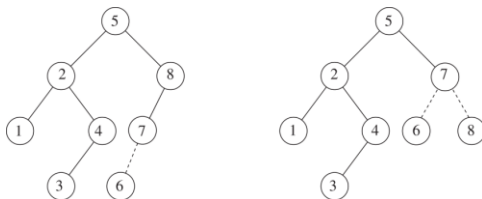
- new tree has same height as original tree

58

58

AVL Trees

- example tree
 - when adding 6, node 8 becomes unbalanced
 - to balance, perform single rotation between 7 and 8

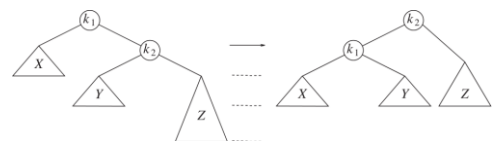


59

59

AVL Trees

- example tree
 - symmetric case for case 4



60

60

AVL Trees

- example
- insert 3, 2, and 1 into an empty tree



61

AVL Trees

- example
- insert 4 and 5



62

AVL Trees

- example
- insert 6



63

AVL Trees

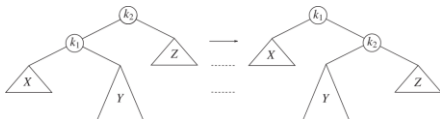
- example
- insert 7



64

AVL Trees

- double rotation
- for cases 2 and 3, a single rotation will not work



- tree Y can be expanded to a node with two subtrees

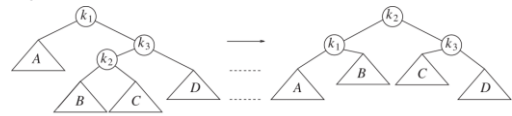
65

AVL Trees

- double rotation
- left-right double rotation for case 2



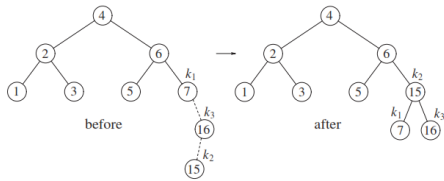
- right-left double rotation for case 3



66

AVL Trees

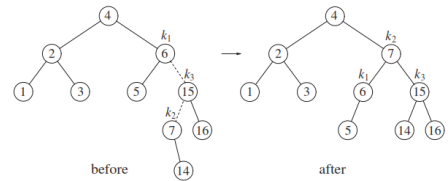
- example
- insert 16 and 15



67

AVL Trees

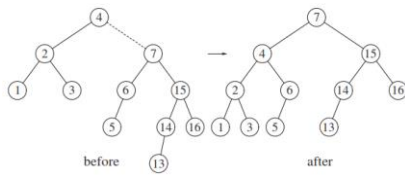
- example
- insert 14



68

AVL Trees

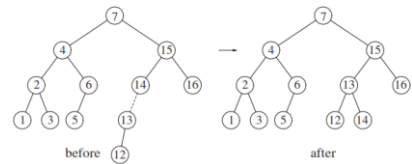
- example
- insert 13



69

AVL Trees

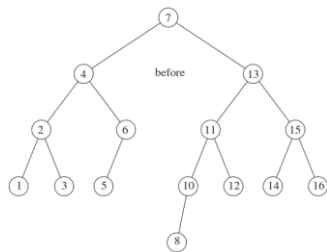
- example
- insert 12



70

AVL Trees

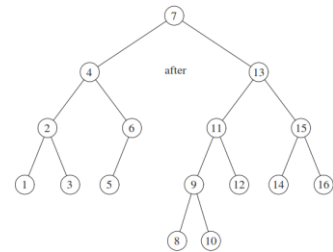
- example
- insert 11, 10, and 8



71

AVL Trees

- example
- insert 9



72

AVL Trees

- implementation
- node definition

```

1 struct AvlNode
2 {
3     Comparable element;
4     AvlNode *left;
5     AvlNode *right;
6     int height;
7
8     AvlNode( const Comparable &ele, AvlNode *lt, AvlNode *rt, int h = 0 )
9         : element( ele ), left( lt ), right( rt ), height( h ) {}
10
11     AvlNode( Comparable &&ele, AvlNode *lt, AvlNode *rt, int h = 0 )
12         : element( std::move( ele ) ), left( lt ), right( rt ), height( h ) {}
13 };

```

73

AVL Trees

- implementation
- function to compute height of AVL node

```

1 /**
2  * Return the height of node t or -1 if nullptr.
3  */
4 int height( AvlNode *t ) const
5 {
6     return t == nullptr ? -1 : t->height;
7 }

```

74

AVL Trees

- implementation
- insertion

```

1 /**
2  * Internal method to insert into a subtree.
3  * x is the item to insert.
4  * t is the node that roots the subtree.
5  * Set the new root of the subtree.
6  */
7 void insert( const Comparable &x, AvlNode * &t )
8 {
9     if( t == nullptr )
10         t = new AvlNode( x, nullptr, nullptr );
11     else if( x < t->element )
12         insert( x, t->left );
13     else if( t->element < x )
14         insert( x, t->right );
15     balance( t );
16 }

```

75

AVL Trees

- implementation

```

19 static const int ALLOWED_IMBALANCE = 1;
20
21 // Assume t is balanced or within one of being balanced
22 void balance( AvlNode * &t )
23 {
24     if( t == nullptr )
25         return;
26
27     if( height( t->left ) - height( t->right ) > ALLOWED_IMBALANCE )
28         if( height( t->left->left ) >= height( t->left->right ) )
29             rotateWithLeftChild( t );
30         else
31             doubleWithLeftChild( t );
32     else
33         if( height( t->right ) - height( t->left ) > ALLOWED_IMBALANCE )
34             if( height( t->right->right ) >= height( t->right->left ) )
35                 rotateWithRightChild( t );
36             else
37                 doubleWithRightChild( t );
38     t->height = max( height( t->left ), height( t->right ) ) + 1;
39 }
40

```

76

AVL Trees

- implementation
- single rotation

```

1 /**
2  * Rotate binary tree node with left child.
3  * For AVL trees, this is a single rotation for case 1.
4  * Update heights, then set new root.
5  */
6 void rotateWithLeftChild( AvlNode * &k2 )
7 {
8     AvlNode *k1 = k2->left;
9     k2->left = k1->right;
10    k1->right = k2;
11    k2->height = max( height( k2->left ), height( k2->right ) ) + 1;
12    k1->height = max( height( k1->left ), k2->height ) + 1;
13    k2 = k1;
14 }

```



77

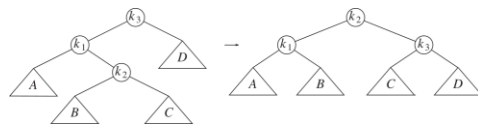
AVL Trees

- implementation
- double rotation

```

1 /**
2  * Double rotate binary tree node: first left child
3  * with its right child; then node k3 with new left child.
4  * For AVL trees, this is a double rotation for case 2.
5  * Update heights, then set new root.
6  */
7 void doubleWithLeftChild( AvlNode * &k3 )
8 {
9     rotateWithRightChild( k3->left );
10    rotateWithLeftChild( k3 );
11 }

```



78

AVL Trees

- implementation
- deletion

```

1 //
2 // Internal method to remove from a subtree.
3 // x is the item to remove.
4 // t is the node that roots the subtree.
5 // Set the new root of the subtree.
6 //
7 void remove(const Comparable & x, AvlNode * & t)
8 {
9     if (t == nullptr)
10         return; // item not found, do nothing
11
12     if (x < t->element)
13         remove(x, t->left);
14     else if (t->element < x)
15         remove(x, t->right);
16     else if (t->left != nullptr && t->right != nullptr) // Two children
17     {
18         t->element = findMin(t->right, t->element);
19         remove(t->element, t->right);
20     }
21     else
22     {
23         AvlNode *oldNode = t;
24         t = (t->left != nullptr) ? t->left : t->right;
25         delete oldNode;
26     }
27     balance(t);
28 }

```

79

Splay Trees

- different approach to ensuring $O(\log N)$ behavior for tree operations (searches, insertions, and deletions).
- worst case
 - splay trees operations may take N time
- however, splay trees make slow operations infrequently
 - guarantee that M consecutive operations (insertions or deletions) requires at most $O(M \log N)$, so, on average, operations are $O(\log N)$
 - $O(\log N)$ is an amortized complexity
 - derivation is complex
- common for binary search trees to have a sustained sequence of bad accesses

80

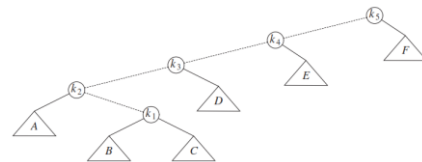
Splay Trees

- basic idea: when a node is accessed, it is moved to the top of the tree, with the thought that we might want to revisit recently accessed nodes more frequently
- use double rotations similar to AVL to move nodes to top of tree
- along the way, more branching is introduced in the tree, which reduces the height of the tree and thus the cost of tree operations

81

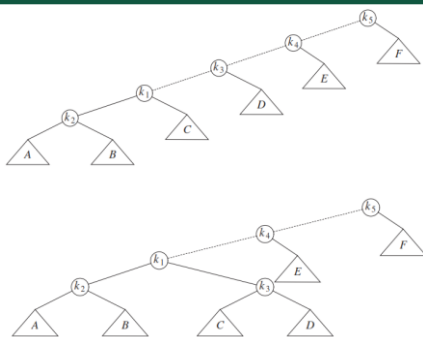
Splay Trees

- single rotations don't work
- access k_1



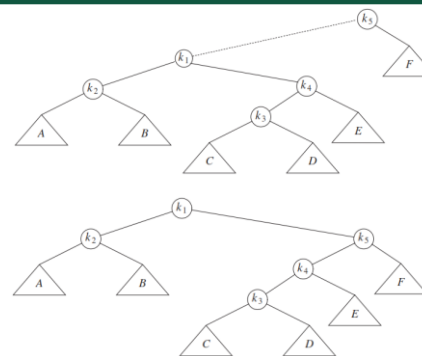
82

Splay Trees



83

Splay Trees



84

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84

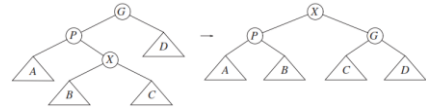
Splay Trees

- double rotations consider parent and grandparent of accessed node
- zig: single branch (in one direction)
- zag: secondary branch (in opposite direction)
- when the parent node is the root, a single rotation for the zig is sufficient

85

Splay Trees

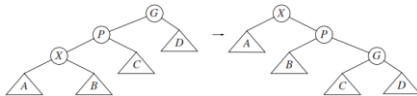
- access X
- zig-zag



86

Splay Trees

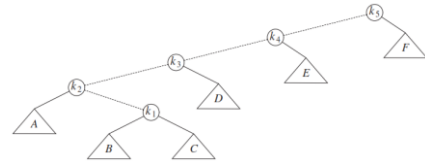
- access X
- zig-zig



87

Splay Trees

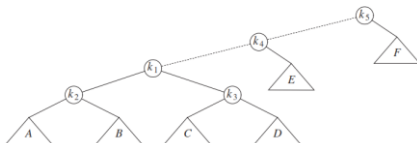
- consider tree from previous example
- access k_1
- zig-zag



88

Splay Trees

- consider tree from previous example
- access k_1
- zig-zig



89

Splay Trees

- consider tree from previous example
- access k_1

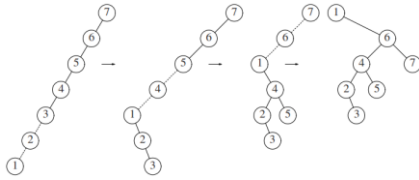


- k_1 is now at the root
- final tree has halved the distance of most nodes on the access path to the root

90

Splay Trees

- example 2
- access 1



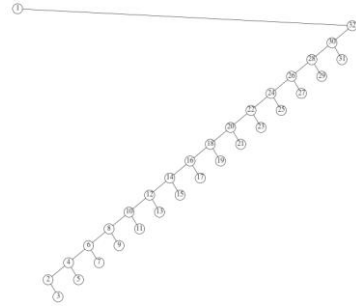
-tree starts as worst case and results in much better structure for performance

91

91

Splay Trees

- example 3: tree with only left children – access 1

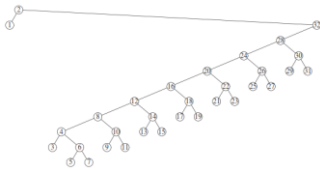


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Splay Trees

- example 3: tree with only left children – access 2

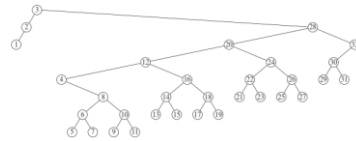


93

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Splay Trees

- example 3: tree with only left children – access 3



94

94

Splay Trees

- example 3: tree with only left children – access 4

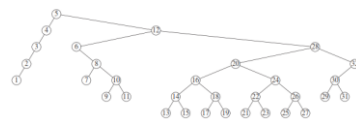


95

95

Splay Trees

- example 3: tree with only left children – access 5

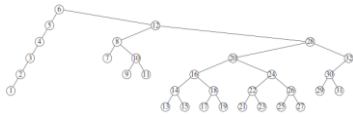


96

96

Splay Trees

-example 3: tree with only left children – access 6

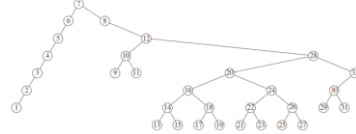


97

97

Splay Trees

-example 3: tree with only left children – access 7

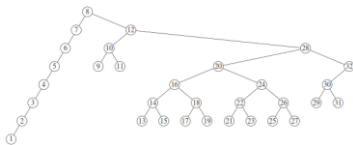


98

98

Splay Trees

-example 3: tree with only left children – access 8

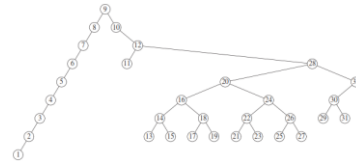


99

99

Splay Trees

-example 3: tree with only left children – access 9



100

100

Splay Trees

- deleting nodes
 - first, access the node, which moves it to the root of the tree
 - let T_L and T_R be the left and right subtrees of the new root
 - find e , the largest element of T_L
 - rotate e to the root of T_L
 - since e is the largest element of T_L , it will have no right child, so we can attach T_R there
 - rather than the largest element of T_L , we could use the smallest element of T_R and modify T_R

101

101

Top-Down Splay Trees

- previous method requires traversal from root down to node, then a bottom-up traversal to implement the splaying
 - can be accomplished by maintaining parent links
 - or by storing access path on the stack
 - both methods require substantial overhead
 - both must handle a variety of special cases

102

102

Top-Down Splay Trees

- instead, perform rotations on initial access path
- result is faster
- uses extra space $O(1)$
- retains amortized time bound of $O(\log N)$

103

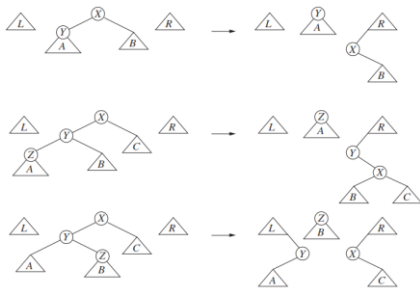
Top-Down Splay Trees

- suppose we wish to access key i
- during the access and concurrent splaying operation, the tree is broken into three parts
 - a left tree, which contains all the keys from the original tree known at the time to be less than i
 - a right tree, which contains all the keys from the original tree known at the time to be greater than i
 - a middle tree, which consists of the subtree of the original tree rooted at the current node on the access path
- initially, the left and right trees are empty and the middle tree is the entire tree
- at each step we tack bits of the middle tree onto the left and right trees

104

Top-Down Splay Trees

- rotations for zig, zig-zig, and zig-zag cases

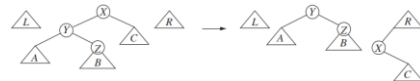


105

105

Top-Down Splay Trees

- zig-zag case can be simplified to just a zig since no rotations are performed
- instead of making Z the root, we make Y the root



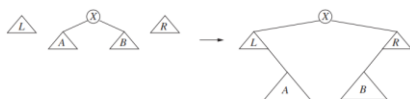
- simplifies coding, but only descends one level
- requires more iterations

106

106

Top-Down Splay Trees

- after final splaying

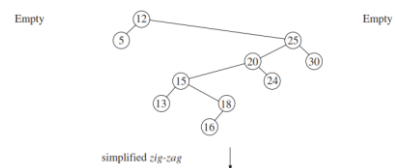


107

107

Top-Down Splay Trees

- example: access 19

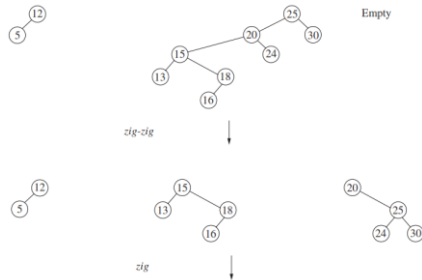


108

108

Top-Down Splay Trees

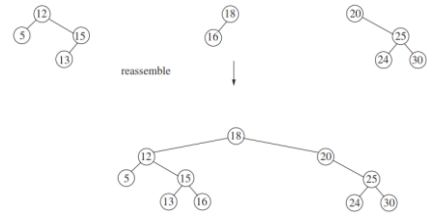
-example: access 19



109

Top-Down Splay Trees

-example: access 19



110

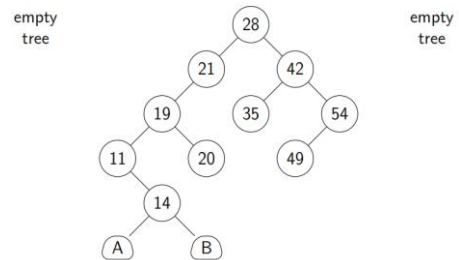
Top-Down Splay Trees

- use a header to hold the roots of the left and right subtrees
- left pointer will contain root of right subtree
- right pointer will contain root of left subtree
- easy to reconstruct at end of splaying

111

Top-Down Splay Trees

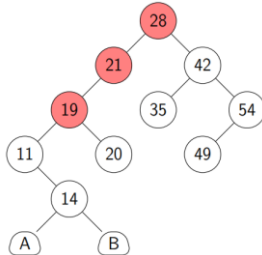
-example 2: access 14



112

Top-Down Splay Trees

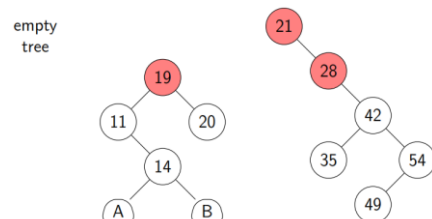
- example 2: access 14
- start at root and look down two nodes along path to 14



113

Top-Down Splay Trees

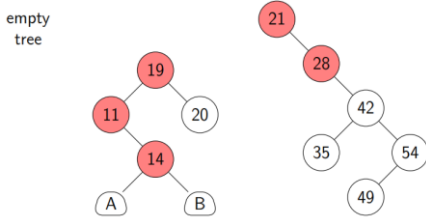
-example 2: access 14



114

Top-Down Splay Trees

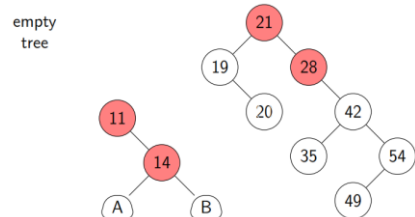
- example 2: access 14
- continuing down the tree, this is a zig-zag condition



115

Top-Down Splay Trees

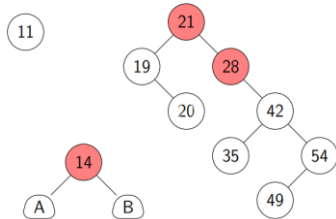
- example 2: access 14
- tree is reconfigured



116

Top-Down Splay Trees

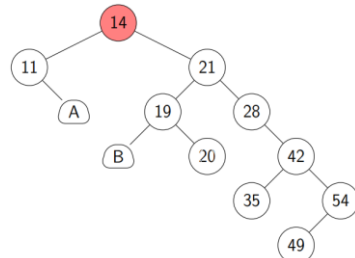
- example 2: access 14
- simple zig



117

Top-Down Splay Trees

- example 2: access 14
- move accessed node to root and reassemble tree



118

B-Trees

- B-trees were developed in the late 1960s by Rudolf Bayer and Edward McCreight:

R. Bayer and E. McCreight, *Organization and maintenance of large ordered indexes*, Acta Informatica vol. 1, no. 3 (1972), pp. 173-189.

- originally motivated by applications in databases
- B-trees shown here really B+ tree

119

B-Trees

- thus far, we have frequently treated the key as if it were the data being stored, but that is rarely the case
- example: student records in Banner
 - most effective search key is W&M ID (e.g., 930...) since it is unique
 - the record (value) associated with each key contains much more information
 - Student Information
 - Student Academic Transcript
 - Student Active Registrations
 - Student Schedule
 - Student E-mail Address
 - Student Address and Phones ...

120

B-Trees

- B-trees are particularly useful when we cannot fit all of our data in memory, but have to perform reads and writes from secondary storage (e.g., disk drives)
 - disk accesses incredibly expensive, relatively speaking
 - consider a disk drive that rotates at 7200 rpm
 - the rotational speed plays a role in retrieval time; for a 7200 rpm disk, each revolution takes $60/7200 = 1/120$ s, or about 8.3 ms
 - a typical seek time (the time for the disk head to move to the location where data will be read or written) for 7200 rpm disks is around 9 ms
 - this means we can perform 100-120 random disk accesses per second

121

121

B-Trees

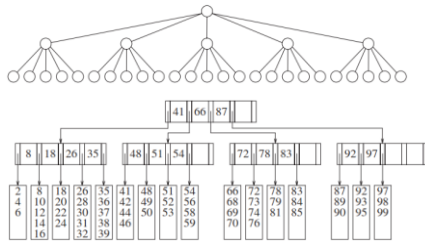
- meanwhile, our CPU can perform $> 1,000,000,000$ operations per second
- suppose we have a database with $N = 10,000,000$ entries that we organize in a tree
 - in an AVL tree, a worst-case search requires $1.44 \lg N \approx 33$ disk accesses
 - at 9 ms per access, this requires about 300 ms, so on average we can perform less than 4 searches per second
 - we would expect 1000 worst-case searches to take $300,000 \text{ ms} = 300 \text{ s}$, or about 5 minutes
 - in this application, search trees with height $\lg N$ are still too high!

122

122

B-Trees

- height can be reduced if we allow more branching
- binary search trees only allow 2-way branching
- example: 5-ary 31-node tree with height 3



123

123

B-Trees

- B-tree of order M is an M -ary tree with the following properties
 1. data items are stored at leaves
 2. nonleaf nodes (internal nodes) store up to $M - 1$ keys to guide the searching: key i represents the smallest key in subtree $i + 1$
 3. root is either a leaf or has between two and M children
 4. all nonleaf nodes (except the root) have between $\lceil M/2 \rceil$ and M children
 5. all leaves are at the same depth and have between $\lceil L/2 \rceil$ and L data items, for some L

124

124

B-Trees

- examples
 - for $M = 2$, there are between $\lceil 2/2 \rceil = 1$ to 2 children
 - for $M = 3$, there are between $\lceil 3/2 \rceil = 2$ to 3 children
 - for $M = 4$, there are between $\lceil 4/2 \rceil = 2$ to 4 children
 - for $M = 5$, there are between $\lceil 5/2 \rceil = 3$ to 5 children
 - for $M = 42$, there are between $\lceil 42/2 \rceil = 21$ to 42 children
- requiring nodes to be half full guarantees that the tree will not degenerate into a simple binary search tree

125

125

B-Trees

- examples: $M = 5$



- all nonleaf nodes have between 3 and 5 children (and thus between 2 and 4 keys)
- root could have just 2 children
- here L is also 5: each leaf has between 3 and 5 data items

126

126

B-Trees

- choosing M and L
 - each node will occupy a disk block, say 8192 bytes, so we choose M and L based on the size of the items being stored
 - suppose each key uses 32 bytes and a link to another node uses 8 bytes
 - a node in a B-tree of order M has $M - 1$ keys and M links, so a node requires

$$32(M - 1) + 8M = 40M - 32 \text{ bytes}$$
 - we choose the largest M that will allow a node to fit in a block

$$M = \left\lfloor \frac{8192 + 32}{40} \right\rfloor = 205$$

127

B-Trees

- choosing M and L (cont.)
 - if the values are each 256 bytes, then we can fit

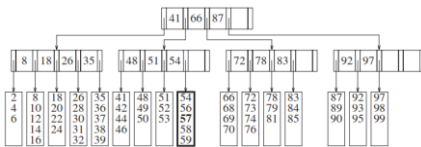
$$L = \left\lfloor \frac{8192}{256} \right\rfloor = 32$$
 in a single block
 - each leaf has between 16 and 32 values, and each internal node branches in at least 103 ways
 - if there are 1,000,000,000 values to store, there are at most 62,500,000 leaves
 - the leaves would be, in the worst case, on level

$$1 + \log_{103} 62,500,000 = 5$$
 so we can find data in at most 5 disk access
 - a BST would have at least $1 + \log_2 62,500,000 = 27$ levels!

128

B-Trees

- insertion: easy case – insert 57
 - first, follow the search tree to the correct leaf (external node)
 - if there are fewer than L items in the leaf, insert in the correct location
 - cost: 1 disk access
- insert 55?



129

B-Trees

- insertion: splitting a leaf – insert 55
 - if there are already L items in the leaf
 - add the new item, split the node in two, and update the links in the parent node
 - cost: 3 disk accesses (one for each new node and one for the update of the parent node)



130

B-Trees

- insertion: splitting a leaf – insert 55 (cont.)
 - the splitting rule ensures we still have a B-tree: each new node has at least $\lceil L/2 \rceil$ values (e.g., if $L = 3$, there are 2 values in one node and 1 in the other, and if $L = 4$, each new node has 2 keys)



131

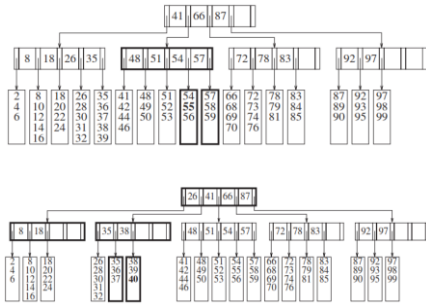
B-Trees

- insertion: splitting a parent – insert 40
 - what if the parent node already has all the child nodes it can possibly have?
 - split the parent node, and update its parent
 - repeat until we arrive at the root
 - if necessary, split the root into two nodes and create a new root with the two nodes as children
 - this is why the root is allowed as few as 2 children
 - thus, a B-tree grows at the root

132

B-Trees

-insertion: splitting a parent – insert 40 (cont.)



133

B-Trees

-insertion: other techniques – insert 29

- put a child up for adoption if a neighbor has room
- here, move 32 to the next leaf
- modifies parent, but keeps nodes fuller and saves space in the long run



134

B-Trees

-deletion: delete 99

- could bring leaf below minimum number of data items
- adopt neighboring item if neighbor not at minimum
- otherwise, combine with neighbor to form a full leaf
- process could make its way up to the root
 - if root left with 1 child, remove root and make its child the new root of the tree



135

B-Trees

-deletion: delete 99 (cont.)



136