Algorithms and Data Structures (II)

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Where are we:

The last topics of this course

- Data Structures for external memory: B-Trees.
- A bit of data compression
- Tries.

Next: B-Trees

Outline:

- Search in secondary storage
- B-Trees
 - properties
 - search
 - insertion

Basic assumption so far: data structures fit completely in main memory (RAM)

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Disk is 10,000–100,000 times slower than RAM

Idea

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- In practice we *increase the degree* (or *branching factor*) of each node up to *d* > 2, so *h* = ⌊log_{*d*} *n*⌋
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E.g., if d = 1000, then only three accesses (h = 2) cover up to one billion keys





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- x.leaf is a Boolean flag that is TRUE if x is a leaf node or FALSE if x is an internal node
- x.c[1], x.c[2], ..., x.c[x.n + 1] are the x.n + 1 pointers to its children, if x is an *internal node*





■ The keys *x*. *key*[*i*] delimit the ranges of keys stored in each subtree



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- every node must contain *at most* 2t 1 *keys*
 - ▶ a node is *full* when it contains exactly 2*t* − 1 keys
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Example



Search in B-Trees

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```
B-TREE-SEARCH(x, k)
1 \quad i = 1
2 while i \le 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
  if x.leaf
7
        return NIL
8
  else DISK-READ(x.c[i])
        return B-TREE-SEARCH(x.c[i], k)
9
```

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- each subtree contains $1 + t + t^2 \cdots + t^{h-1}$ nodes, each one containing t 1 keys, so

$$n \ge 1 + 2(t^h - 1)$$











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

1 $z = ALLOCATE-NODE()$
2 $z.leaf = y.leaf$
3 $z.n = t - 1$
4 **for** $j = 1$ **to** $t - 1$
5 $z.key[j] = y.key[j + t]$
6 **if not** $y.leaf$
7 **for** $j = 1$ **to** t
8 $z.c[j] = y.c[j + t]$
9 $y.n = t - 1$
10 **for** $j = x.n + 1$ **downto** $i + 1$
11 $x.c[j + 1] = x.c[j]$
12 **for** $j = x.n + 0$ **downto** i
13 $x.key[j + 1] = x.key[j]$
14 $x.key[i] = y.key[t]$
15 $x.n = x.n + 1$
16 **DISK-WRITE** (y)
17 **DISK-WRITE** (z)
18 **DISK-WRITE** (x)

Complexity of **B-TREE-SPLIT-CHILD**

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- \blacksquare $\Theta(t)$ basic CPU operations
- **3 DISK-WRITE** operations

```
B-TREE-SPLIT-CHILD(x, i, y)
    z = Allocate-Node()
 2 z.leaf = y.leaf
 3 z.n = t - 1
 4 for i = 1 to t - 1
        x.key[j] = x.key[j+t]
 5
 6 if not x.leaf
 7
        for j = 1 to t
 8
             z.c[j] = y.c[j + t]
 9 y.n = t - 1
10 for j = x.n + 1 downto i + 1
11
        x.c[i + 1] = x.c[i]
12 for i = x \cdot n downto i
13
        x.key[i+1] = x.key[i]
14 x.key[i] = y.key[t]
15 x.n = x.n + 1
16 DISK-WRITE(y)
17
    DISK-WRITE(z)
18
    DISK-WRITE(x)
```

Insertion Under Non-Full Node

Insertion Under Non-Full Node

```
B-TREE-INSERT-NONFULL(x, k)
    i = x.n
                                        II assume x is not full
     if x.leaf
 2
 3
          while i \ge 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 \ge 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 // child x.c[i] is full
14
              B-TREE-SPLIT-CHILD(x, i, x, c[i])
15
               if k > x. key[i]
16
                    i = i + 1
17
          B-TREE-INSERT-NONFULL(x, c[i], k)
```

Insertion Procedure

Insertion Procedure

B-Tree-Insert(T, k)

r = T.root1 2 **if** r, n = 2t - 13 s = Allocate-Node()4 T.root = s5 s.leaf = FALSE 6 s.*n* = 0 7 s.c[1] = r8 **B-TREE-SPLIT-CHILD**(s, 1, r)9 **B-Tree-Insert-Nonfull**(*s*, *k*) **else B-TrEe-INSERTHUL**(*r*, *k*) 10

Insertion Procedure

B-Tree-Insert(T, k)

- 1 r = T.root2 **if** r.n = 2t - 1
- 3 s = Allocate-Node()
- 4 T.root = s

7
$$s.c[1] = r$$

8 **B-TREE-SPLIT-CHILD**
$$(s, 1, r)$$

- 9 **B-Tree-Insert-Nonfull**(*s*, *k*)
- 10 else B-Tree-Insert-Nonfull(r, k)



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- $O(th) = O(t \log_t n)$ basic CPU steps operations
- $O(h) = O(\log_t n)$ disk-access operations
- The best value for *t* can be determined according to
 - the ratio between CPU (RAM) speed and disk-access time
 - the block-size of the disk, which determines the maximum size of an object that can be accessed (read/write) in one shot

Deletion

Mirror image of insertion.

- In insertion: key always goes to leaf. Before inserting new key check if node to insert is full
- If so: first split node, to make it non-full.
- : Deletion: want to delete from leaf. But key may not be in leaf.
- **B**efore deleting key check if node to delete from is minimal (t 1 keys).

Deletion

- Case 1: key is in a leaf, delete key.
- Case 2: key is **not** in a leaf. Then its predecessor/successor are in leaf. Delete key, promote pred/succ.
- Cases 1/2 may cause leaf node to become defficient (too few keys). Have to make it nonminimal.
- Look at the immediately adjacent siblings of this node. Several cases:
 - If there is a non-minimal sibling, then take a key/child pointer from that sibling to the parent, and one key/child from parent to defficient leaf.
 - If both siblings are minimal: merge node with one sibling (doesn't matter which) and one node from parent. If this makes the parent have too few nodes repeat recursively.