\_\_\_\_ Disk Accesses \_\_\_\_\_ • Consider any search tree • The number of disk accesses per search will dominate the CS 361, Lecture 25 run time • Unless the entire tree is in memory, there will usually be a Jared Saia disk access every time an arbitrary node is examined University of New Mexico • The number of disk accesses for most operations on a B-tree is proportional to the height of the B-tree • I.e. The info on each node of a B-tree can be stored in main memorv 3 \_\_\_\_\_ B-Tree Properties \_\_\_\_\_ \_\_\_ Outline \_\_\_\_\_ The following is true for every node  $\boldsymbol{x}$ • B-Trees Skip Lists • x stores keys,  $key_1, \dots key_l(x)$  in sorted order (nondecreasing) • Graph Theory Intro • x contains pointers,  $c_1(x), \ldots, c_{l+1}(x)$  to its children • Let  $k_i$  be any key stored in the subtree rooted at the *i*-th child of x, then  $k_1 \leq key_1(x) \leq k_2 \leq key_2(x) \cdots \leq key_l(x) \leq k_{l+1}$ 1 4 B-Trees \_\_\_\_\_ B-Tree Properties \_\_\_\_\_

- B-Trees are balanced search trees designed to work well on disks
- B-Trees are *not* binary trees: each node can have many children
- Each node of a B-Tree potentially contains *several* keys, not just one
- When doing searches, we decide which child link to follow by finding the correct interval of our search key in the key set of the current node.

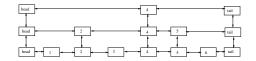
- All leaves have the same depth
- Lower and upper bounds on the number of keys a node can contain. Given as a function of a fixed integer t
  - If the tree is non-empty, the root must have at least one key, and 2 children
  - Every node other than the root must have at least (t-1) keys, and all internal nodes other than the root must have at least t children.
  - Every node can contain at most 2t 1 keys, and so any internal node can have at most 2t children

Note \_\_\_\_\_ • Technically, not a BST, but they implement all of the same • The above properties imply that the height of a B-tree is no operations more than  $\log_t \frac{n+1}{2}$ , for  $t \ge 2$ , where *n* is the number of keys. • Very elegant randomized data structure, simple to code but • If we make t, larger, we can save a larger (constant) fraction analysis is subtle over RB-trees in the number of nodes examined • They guarantee that, with high probability, all the major op-• A (2-3-4)-tree is just a *B*-tree with t = 2erations take  $O(\log n)$  time • We'll discuss them more next class 6 9 \_\_\_\_\_ High Level Analysis \_\_\_\_\_ In-Class Exercise \_\_\_\_\_ We will now show that for any B-Tree with height h and n keys, Comparison of various BSTs  $h \leq \log_t \frac{n+1}{2}$ , where  $t \geq 2$ . • RB-Trees: + guarantee  $O(\log n)$  time for each operation, easy to augment, - high constants Consider a B-Tree of height h > 1• AVL-Trees: + guarantee  $O(\log n)$  time for each operation, high constants • Q1: What is the minimum number of nodes at depth 1, 2, • B-Trees: + guarantee  $O(\log n)$  time for each operation, and 3 works well for trees that won't fit in memory, - inserts and • Q2: What is the minimum number of nodes at depth *i*? deletes are more complicated • Q3: Now give a lowerbound for the total number of keys • Splay Tress: + small constants, - amortized guarantees only (e.g.  $n \ge ???$ ) • Skip Lists: + easy to implement, - runtime guarantees are • Q4: Show how to solve for h in this inequality to get an probabilistic only upperbound on h7 10 Splay Trees \_\_\_\_\_ \_\_\_\_\_ Which Data Structure to use? \_\_\_\_\_ • A Splay Tree is a kind of BST where the standard operations • Splay trees work very well in practice, the "hidden constants" run in  $O(\log n)$  amortized time are small • This means that over *l* operations (e.g. Insert, Lookup, • Unfortunately, they can not guarantee that every operation Delete, etc), where l is sufficiently large, the total cost is takes  $O(\log n)$ • When this guarantee is required, B-Trees are best when the  $O(l * \log n)$ • In other words, the average cost per operation is  $O(\log n)$ entire tree will not be stored in memory • However a single operation could still take O(n) time • If the entire tree will be stored in memory, RB-Trees, AVL- In practice, they are very fast Trees, and Skip Lists are good

11

\_\_\_ Example \_\_\_\_

- Technically, not a BST, but they implement all of the same operations
- Very elegant randomized data structure, simple to code but analysis is subtle
- They guarantee that, with high probability, all the major operations take  $O(\log n)$  time



15

16

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___ Skip List _____
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- A skip list is basically a collection of doubly-linked lists,  $L_1, L_2, \ldots, L_x$ , for some integer x
- Each list has a special head and tail node, the keys of these nodes are assumed to be -MAXINT and +MAXINT respectively
- The keys in each list are in sorted order (non-decreasing)

\_\_\_\_ Search \_\_\_\_\_

Search(k){
 pLeft = L\_x.head;
 for (i=x;i>=0;i--){
 Search from pLeft in L\_i to get the rightmost elem, r,
 with value <= k;
 pLeft = pointer to r in L\_(i-1);
 }
 if (pLeft==k)
 return pLeft
 else
 return nil
 }
}</pre>

13

12

\_ Skip List \_\_\_\_

## \_\_\_\_ Insert \_\_\_\_

• Every key is in the list  $L_1$ .

- For all i > 2, if a key x is in the list  $L_i$ , it is also in  $L_{i-1}$ . Further there are up and down pointers between the x in  $L_i$  and the x in  $L_{i-1}$ .
- All the head(tail) nodes from neighboring lists are interconnected

 $\boldsymbol{p}$  is a constant between 0 and 1, typically  $\boldsymbol{p}=1/2$ 

```
Insert(k){
First call Search(k), let pLeft be the leftmost elem <= k in L_1
Insert k in L_1, to the right of pLeft
i = 2;
while (rand()<p){
    insert k in the appropriate place in L_i;
}</pre>
```

Q: How much memory do we expect a skip list to use up? • Deletion is very simple • First do a search for the key to be deleted • Let  $X_i$  be the number of lists elem i is inserted in • Then delete that key from all the lists it appears in from • Q: What is  $P(X_i \ge 1)$ ,  $P(X_i \ge 2)$ ,  $P(X_i \ge 3)$ ? the bottom up, making sure to "zip up" the lists after the • Q: What is  $P(X_i \ge k)$  for general k? • Q: What is  $E(X_i)$ ? deletion • Q: Let  $X = \sum_{i=1}^{n} X_i$ . What is E(X)? 18 21 \_\_\_\_\_ Height of Skip List \_\_\_\_\_ In-Class Exercise \_\_\_\_\_ A trick for computing expectations of discrete positive random variables: • Assume there are *n* nodes in the list • Q: What is the probability that a particular key *i* achieves • Let X be a discrete r.v., that takes on values from 1 to nheight  $k \log n$  for some constant k? • A: If p = 1/2,  $P(X_i \ge k \log n) = \frac{1}{n^k}$  $E(X) = \sum_{i=1}^{n} P(X \ge i)$ 19 22 \_\_\_\_ Why? \_\_\_\_\_ \_\_\_\_\_ Height of Skip List \_\_\_\_\_ • Q: What is the probability that any of the nodes achieve height higher than  $k \log n$ ? • A: We want  $\sum_{i=1}^{n} P(X \ge i) = 1 * P(X = 1) + 2 * P(X = 2) + \dots \quad (1)$  $= E(X) \quad (2)$ 

 $P(X_1 \ge k \log n \text{ or } X_2 \ge k \log n \text{ or } \dots \text{ or } X_n \ge k \log n)$ 

• By a Union Bound, this probability is no more than

$$P(X_1 \ge k \log n) + P(X_2 \ge k \log n) + \dots + P(X_n \ge k \log n)$$
• Which equals  $\frac{n}{n^k} = n^{1-k}$ 

- If we choose k to be, say 10, this probability gets very small as n gets large
- In particular, the probability of having a skip list of size exceeding  $k \log n$  is o(1)
- So we say that the height of the skip list is  $O(\log n)$  with high probability

	24

- Note that the expected number of "siblings" of a node, x, at any level i is 2
- Why? Because for a node to be a sibling of x at level *i*, it must have failed to advance to the next level
- The first node that advances to the next level ends the possibility of further siblings.
- This is the same as asking expected number of times we need to flip a coin to get a heads the answer is 2

25

\_\_\_ Search Time \_\_\_\_

Search Time

- The expected number of "siblings" of a node, x, at any level i is 2
- The number of levels is  $O(\log n)$  with high probability
- From these two facts, we can argue that the expected search time is  $O(\log n)$
- (Warning: The argument is not as simple as multiplying these two values. We can't do this since the two random variables are not independent.)