Algorithms and Data Structures (II)

Gabriel Istrate

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Outline

- Wrap up hash tables.
- Skip lists.
- Binary search trees
- Randomized binary search trees

Where are we?

- A dictionary is an abstract data structure that represents a set of elements (or keys)
 - ► a dynamic set

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- Implementation (so far)
 - direct access tables. Linked lists. Hash tables.

Hashing with Chaining



Hashing with Open Addressing



Three versions: linear/quadratic/double probing.

Hash Tables: Scorecard

Algorithm	Average Complexity (Search successful/not)	
INSERT/SEARCH/DELETE, CHAINING:	$O(1+\alpha)$ \checkmark	
SEARCH, LINEAR PROBING:	$\frac{1}{2}(1+\frac{1}{1-\alpha}), \frac{1}{2}(1+\frac{1}{1-\alpha^2}) \qquad \checkmark$	
SEARCH, QUADRATIC PROBING:	$1 - \ln(1 - \alpha) - \frac{\alpha}{2}, \frac{1}{1 - \alpha} - \alpha - \ln(1 - \alpha)$	
SEARCH, DOUBLE HASHING:	$\frac{1}{\alpha}\ln(1-\alpha), \frac{1}{1-\alpha} \qquad \checkmark$	

Reference, probing complexities: Drozdek/Knuth.

- practical ! ✓
- Hard to analyze mathematically: those results under uniform hashing (not at all clear) ×
- Somewhat hard to engineer. ×.

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Perhaps $O(1(+\alpha))$ too ambitious ? Something, say O(logn)?

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In practice log(n) is a small number !

Caution

Topic **not** in Cormen. See Drozdek for details/C++ implementation.

- Problem with linked list: *search is slow !...* even when elements sorted.
- Solution: lists of ordered elements that allow skipping some elements to speed up search.
- Skip lists: variant of ordered linked lists that makes such search possible.

More advanced data structure (W. Pugh "Skip lists: a Probabilistic Alternative to Balanced Trees", Communication of the ACM 33(1990), pp. 668-676.) *If anyone curious/interested in data structures/algorithms, can give paper to read; taste how a research article looks like.*



Too theoretical?

Where does this ever get applied ? ...

Skip lists in real life

According to Wikipedia:

- MemSQL skip lists as prime indexing structure for its database technology.
- Cyrus IMAP server "skiplist" backend DB implementation
- Lucene uses skip lists to search delta-encoded posting lists in logarithmic time.
- QMap (up to Qt 4) template class of Qt that provides a dictionary.
- Redis, ANSI-C open-source persistent key/value store for Posix systems, skip lists in implementation of ordered sets.
- **nessDB**, a very fast key-value embedded Database Storage Engine.
- skipdb: open-source DB format using ordered key/value pairs.
- ConcurrentSkipListSet and ConcurrentSkipListMap in the Java 1.6 API.

Skip lists in real life (II)

According to Wikipedia:

- Speed Tables: fast key-value datastore for Tcl that use skiplists for indexes and lockless shared memory.
- leveldb, a fast key-value storage library written at Google that provides an ordered mapping from string keys to string values
- MuQSS Scheduler for the Linux kernel uses skip lists
- SkipMap uses skip lists as base data structure to build a more complex 3D Sparse Grid for Robot Mapping systems.

Skip lists: implementation

What we want

- $k = 1, \ldots, \lfloor \log_2(n) \rfloor, 1 \leq i \leq \lfloor n/2^{k-1} \rfloor 1.$
 - Item $2^{k-1} \cdot i$ points to item $2^{k-1} \cdot (i+1)$.
 - every second node points to positions two node ahead,
 - every fourth node points to positions four nodes ahead,
 - every eigth node points to positions eigth nodes ahead,
 -, and so on.

Different number of pointers in different nodes in the list !

- half the nodes only one pointer.
- a quarter of the nodes two pointers,
- an eigth of the nodes four pointers,
-, and so on.
- $n \log_2(n)/2$ pointers.

Search Algorithm

- First follow pointers on the highest level until a larger element is found or the list is exhausted.
- If a larger element is found, restart search from its predecessor, this time on a lower level.
- Continue doing this until element found, or you reach the first level and a larger element or the end of the list.

Inserting and deleting nodes

Major problem

- When inserting/deleting a node, pointers of prev/next nodes have to be restructured.
- Solution: rather than equal spacing, random spacing on a level.
- Invariant: Number of nodes on each level: equal, in expectation to what it would be under equal spacing

Principle

If you're traveling 10 meters in 10 steps, a step is on average one meter.

Inserting and deleting nodes (II)

- Level numbering: start with zero.
- New node inserted: probability 1/2 on first level, 1/4 second level, 1/8 third level, . . ., etc.
- Function *chooseLevel*: chooses randomly the level of the new node.
- Generate random number. If in [0,1/2] level 1, [1/2,3/4] level 2, etc.
- To delete node: have to update all links.

Computing the *i*'th element faster than in *O*(*i*)

- If we record "step sizes" in our lists we can even mimic indexing !
- Start on highest level.
- If step too big, restart search from predecessor, this time on a lower level.
- Continue doing this until element found.

Update "step sizes" by insertion/deletion

Easy if you have doubly linked lists.

- On deletion: *pred*[*i*].*size*+ = *deleted*.*size* on all levels *i*.
- On insertion: Simply keep track of predecessors and index of the inserteed sequence.

Skip Lists: Scorecard

Method Average Worst-Case

SPACE:	<i>O</i> (<i>n</i>)	O(nlog(n))
✓		
SEARCH:	$O(\log(n))$	<i>O</i> (<i>n</i>)
<u></u>		
INSERT: ($O(\log(n))$	O(n)
	$\mathbf{O}(\mathbf{I}_{1}, \mathbf{I}_{2})$	
DELETE: ($O(\log(n))$	O(n)
√		

- quite practical ! ✓
- Probabilistic, worst-case still bad. ×
- Not completely easy to implement. ×.

Compared to what?

Binary search trees. Will learn about them next.

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 - over a totally ordered domain

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 - iteration: TREE-SUCCESSOR(x) and TREE-PREDECESSOR(x) find the successor and predecessor, respectively, of an element x

Implementation

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Node *x*

- *x.parent* is the parent of node *x*
- x.key is the key stored in node x
- x.left is the left child of node x
- *x.right* is the right child of node x











Binary-search-tree property

- ► for all nodes *x*, *y*, and *z*
- $y \in left$ -subtree $(x) \Rightarrow y$.key $\leq x$.key
- $z \in right$ -subtree $(x) \Rightarrow z$.key $\geq x$.key

Inorder Tree Walk

■ We want to go through the set of keys *in order*



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2 4 5 9 12 13 15 17 18 19
Inorder Tree Walk (2)

A recursive algorithm

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A recursive algorithm

INORDER-TREE-WALK(<i>x</i>)1	if $x \neq \text{NIL}$
2	INORDER-TREE-WALK(<i>x</i> . <i>left</i>)
3	print x. key
4	INORDER-TREE-WALK (<i>x</i> . <i>right</i>)

Inorder Tree Walk (2)

A recursive algorithm

if $x \neq NIL$
INORDER-TREE-WALK(<i>x</i> . <i>left</i>)
print x. <i>key</i>
INORDER-TREE-WALK(x.right)

And then we need a "starter" procedure

INORDER-TREE-WALK-START(T)**1 INORDER-TREE-WALK**(T.root)











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Postorder-Tree-Walk (<i>x</i>)1	if $x \neq \text{NIL}$
2	Postorder-Tree-Walk (<i>x.left</i>)
3	POSTORDER-TREE-WALK (<i>x</i> . <i>right</i>)
4	print x. key









Reverse-Order-Tree-Walk (<i>x</i>)1	if $x \neq \text{NIL}$
2	Reverse-Order-Tree-Walk (<i>x.right</i>)
3	print <i>x. key</i>
4	Reverse-Order-Tree-Walk (<i>x.left</i>)

REVERSE-ORDER-TREE-WALK(x)1if $x \neq NIL$ 2REVERSE-ORDER-TREE-WALK(x.right)3print x.key4REVERSE-ORDER-TREE-WALK(x.left)



Reverse-Order-Tree-Walk(x)1if $x \neq NIL$ 2Reverse-Order-Tree-Walk(x.right)3print x.key4Reverse-Order-Tree-Walk(x.left)



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Application of postorder: Computing Arithmetic Expressions

- Arithmetic expressions can be represented by syntax trees.
- Given an expression represented by tree, compute its value !
- Each tree node: value field.
- Postorder traversal prints postfix notation/computes the value.

The general recurrence is

$$T(n) = T(n_L) + T(n - n_L - 1) + \Theta(1)$$

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R EVERSE-ORDER-TREE-WALK	$\Theta(n)$

We could prove this using the substitution method

Can we do better? No!

• the length of the output is $\Theta(n)$

Minimum and Maximum Keys

Minimum and Maximum Keys

- Recall the binary-search-tree property
 - for all nodes x, y, and z
 - $y \in left$ -subtree $(x) \Rightarrow y$.key $\leq x$.key
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Minimum and Maximum Keys

- Recall the binary-search-tree property
 - for all nodes x, y, and z
 - $y \in left$ -subtree $(x) \Rightarrow y$.key $\leq x$.key
 - ▶ $z \in right$ -subtree $(x) \Rightarrow z$.key $\ge x$.key
- So, the minimum key is in all the way to the left
 - similarly, the maximum key is all the way to the right

TREE-MINIMUM(x)1	while x.left \neq NIL x = x.left
3 TREE-MAXIMUM(<i>x</i>)1 2 3	while x.right \neq NIL x = x.right return x











Given a node *x*, find the node containing the next key value



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■ The successor of *x* is the *minimum* of the *right* subtree of *x*, if that exists

• Otherwise it is the *first ancestor a* of *x* such that *x* falls in the *left* subtree of *a*























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$$T(n) = \Theta(depth of the tree)$$
$$T(n) = O(n)$$

Search (2)

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■ Iterative *binary search*

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■ Iterative *binary* search

ITERATIVE-TREE-SEARCH(T, k)1 x = T.root2 while $x \neq \text{NIL} \land k \neq x.key$ 3 if k < x.key4 x = x.left5 else x = x.right6 return x

Insertion

Insertion



Insertion



Idea

- in order to insert x, we search for x (more precisely x.key)
- if we don't find it, we add it where the search stopped

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.parent = y
9	if $y = NIL$
10	T.root = z
11	else if <i>z</i> . <i>key</i> < <i>y</i> . <i>key</i>
12	y.left = z
13	else y.right = z





























$T(n) = \Theta(h)$

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 - the problem is that the "worst" case is not that uncommon
- *Idea:* use randomization to turn all cases in the average case

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 - problem: A is not necessarily known in advance
- Idea 2: we can obtain a random permutation of the input sequence by randomly alternating two insertion procedures
 - tail insertion: this is what TREE-INSERT does
 - head insertion: for this we need a new procedure TREE-ROOT-INSERT
 - inserts *n* in *T* as if *n* was inserted as the first element

TREE-RANDOMIZED-INSERT1(T, z)1r = uniformly random value from $\{1, \ldots, t.size + 1$ 2if r = 13TREE-ROOT-INSERT(T, z)4else TREE-INSERT(T, z)

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Does this really simulate a random permutation?

i.e., with all permutations being equally likely?

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It is true that any node has the same probability of being inserted at the top

this suggests a recursive application of this same procedure

TREE-RANDOMIZED-INSERT(t, z) 1 **if** t = NIL2 return z 3 r = uniformly random value from $\{1, \ldots, t.size + 1\}$ $\# \Pr[r = 1] = 1/(t.size + 1)$ 4 **if** r = 15 $z_size = t_size + 1$ 6 return Tree-Root-Insert(t, z)7 if z.key < t.key8 t.left = Tree-RANDOMIZED-INSERT(t.left, z)9 else t.right = TREE-RANDOMIZED-INSERT(t.right, z)10 t.size = t.size + 111 return t

```
TREE-RANDOMIZED-INSERT(t, z) 1 if t = NIL
                   2
                           return z
                   3
                     r = uniformly random value from \{1, \ldots, t.size + 1\}
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                   4 if r = 1
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                     else t.right = Tree-RANDOMIZED-INSERT(t.right, z)
                  10 t.size = t.size + 1
                  11
                      return t
```

Looks like this one really simulates a random permutation...





• x = Right-Rotate(x)



- **•** x = Right-Rotate(x)
- **•** x = Left-Rotate(x)









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Root Insertion (2)

TREE-ROOT-INSERT(x, z)1if x = NIL2return z3if z.key < x.key4x.left = TREE-ROOT-INSERT(x.left, z)5return RIGHT-ROTATE(x)6else x.right = TREE-ROOT-INSERT(x.right, z)7return LEFT-ROTATE(x)

General strategies to deal with complexity in the worst case

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 - relatively expensive but "amortized" operations
 - optimized data structures: a self-balanced data structure
 - guaranteed $O(\log n)$ complexity bounds

Deletion

Deletion



Deletion



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 - simply remove z





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- 2. z has one child
 - remove z
 - connect z. parent to z. right
- 3. z has two children



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- 3. z has two children
 - replace z with
 y = TREE-SUCCESSOR(z)
 - remove y (1 child!)



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 - simply remove z
- 2. z has one child
 - remove z
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 - connect y. parent to y. right

Deletion (2)

TREE-DELETE(T, z) 1 if z.left = NIL or z.right = NIL 2 V = Z3 else y = TREE-SUCCESSOR(z)4 **if** y.left \neq NIL 5 x = y.left6 else x = y.right 7 **if** $x \neq \text{NIL}$ 8 x.parent = y.parent9 if y.parent == NIL 10 T.root = x**else if** y = y. parent. left 11 12 y.parent.left = x13 **else** y.parent.right = x14 if $y \neq z$ 15 z.key = y.key16 copy any other data from y into z