# Algorithms

#### Hash Tables

#### **Review: Skip Lists**





• Keep a doubly-linked list of elements

- Min, max, successor, predecessor: O(1) time
- Delete is O(1) time, Insert is O(1)+Search time
- During insert, add each level-*i* element to level *i*+1 with probability *p* (e.g., *p* = 1/2 or *p* = 1/4)

## Summary: Skip Lists

- O(1) expected time for most operations
- O(lg *n*) expected time for insert
- $O(n^2)$  time worst case
  - But random, so no particular order of insertion evokes worst-case behavior
- O(n) expected storage requirements
- Easy to code

#### **Review: Hash Tables**

• Hash table:

• Given a table *T* and a record *x*, with key (= symbol) and satellite data, we need to support:

- Insert (T, x)
- $\circ$  Delete (*T*, *x*)
- $\circ$  Search(*T*, *x*)
- We want these to be fast, but don't care about sorting the records
- In this discussion we consider all keys to be (possibly large) natural numbers

#### **Review: Direct Addressing**

- Suppose:
  - The range of keys is 0..*m*-1
  - Keys are distinct
- The idea:
  - Set up an array T[0..m-1] in which
    - T[i] = x if  $x \in T$  and key[x] = i
    - $\circ$  T[*i*] = NULL otherwise
  - This is called a *direct-address table*

• Operations take O(1) time!

## Review: The Problem With Direct Addressing

- Direct addressing works well when the range *m* of keys is relatively small
- But what if the keys are 32-bit integers?
  - Problem 1: direct-address table will have 2<sup>32</sup> entries, more than 4 billion
  - Problem 2: even if memory is not an issue, the time to initialize the elements to NULL may be
- Solution: map keys to smaller range 0..*m*-1
- This mapping is called a *hash function*

#### Hash Functions

#### • Next problem: *collision*



## **Resolving Collisions**

- *How can we solve the problem of collisions?*
- Solution 1: *chaining*
- Solution 2: open addressing

# **Open Addressing**

- Basic idea (details in Section 12.4):
  - To insert: if slot is full, try another slot, ..., until an open slot is found (*probing*)
  - To search, follow same sequence of probes as would be used when inserting the element

• If reach element with correct key, return it

- If reach a NULL pointer, element is not in table
- Good for fixed sets (adding but no deletion)

Example: spell checking

• Table needn't be much bigger than *n* 

• Chaining puts elements that hash to the same slot in a linked list:



• *How do we insert an element?* 



• *How do we delete an element?* 

Do we need a doubly-linked list for efficient delete?



• *How do we search for a element with a given key?* 



- Assume *simple uniform hashing*: each key in table is equally likely to be hashed to any slot
- Given *n* keys and *m* slots in the table: the load factor  $\alpha = n/m =$  average # keys per slot
- What will be the average cost of an unsuccessful search for a key?

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- What will be the average cost of a successful search?

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- What will be the average cost of a successful search? A:  $O(1 + \alpha/2) = O(1 + \alpha)$

## Analysis of Chaining Continued

- So the cost of searching =  $O(1 + \alpha)$
- If the number of keys n is proportional to the number of slots in the table, what is  $\alpha$ ?
- A:  $\alpha = O(1)$ 
  - In other words, we can make the expected cost of searching constant if we make α constant

## **Choosing A Hash Function**

- Clearly choosing the hash function well is crucial
  - What will a worst-case hash function do?
  - What will be the time to search in this case?
- What are desirable features of the hash function?
  - Should distribute keys uniformly into slots
  - Should not depend on patterns in the data

### Hash Functions: The Division Method

- $h(k) = k \mod m$ 
  - In words: hash k into a table with m slots using the slot given by the remainder of k divided by m
- What happens to elements with adjacent values of k?
- What happens if m is a power of 2 (say  $2^{P}$ )?
- What if m is a power of 10?
- Upshot: pick table size *m* = prime number not too close to a power of 2 (or 10)

## Hash Functions: The Multiplication Method

- For a constant A, 0 < A < 1:
- $h(k) = \lfloor m(kA \lfloor kA \rfloor) \rfloor$

What does this term represent?

## Hash Functions: The Multiplication Method

- For a constant A, 0 < A < 1:
- $h(k) = \lfloor m (kA \lfloor kA \rfloor) \rfloor$ Fractional part of kA
- Choose  $m = 2^P$
- Choose *A* not too close to 0 or 1
- Knuth: Good choice for  $A = (\sqrt{5} 1)/2$

## Hash Functions: Worst Case Scenario

- Scenario:
  - You are given an assignment to implement hashing
  - You will self-grade in pairs, testing and grading your partner's implementation
  - In a blatant violation of the honor code, your partner:
    - Analyzes your hash function
    - Picks a sequence of "worst-case" keys, causing your implementation to take O(*n*) time to search
- What's an honest CS student to do?

# Hash Functions: Universal Hashing

- As before, when attempting to foil an malicious adversary: randomize the algorithm
- *Universal hashing*: pick a hash function randomly in a way that is independent of the keys that are actually going to be stored
  - Guarantees good performance on average, no matter what keys adversary chooses