

CSE 326: Data Structures More Hashing Techniques

Hannah Tang and Brian Tjaden Summer Quarter 2002

Remember This List?

- How should we resolve collisions?
- What should the table size be?
- What should the hash function be?
- How well does hashing work in the real world? - We'll see a case study today!

Hashing Dilemma

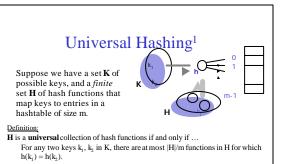
Suppose your **WorstEnemy** 1) knows your hash function; 2) gets to decide which keys to send you?

Faced with this enticing possibility, WorstEnemy decides to: a) Send you keys which maximize collisions for your hash function.b) Take a nap.

Moral: No single hash function can protect you!

Faced with this dilemma, you:

- a) Give up and use a linked list for your Dictionary.
- b) Drop out of software, and choose a career in fast foods.c) Run and hide.
- d) Proceed to the next slide, in hope of a better alternative.



• So ... if we randomly choose a hash function from H. our chances of collision are no more than if we get to choose hash table entries at random! Motivation: see previous slide (or visit http://www.burgerking.com/job:

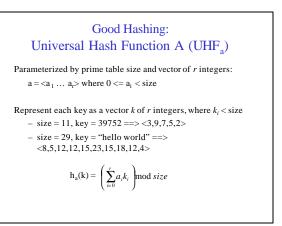
Random Hashing – Not!

How can we "randomly choose a hash function"? Certainly we cannot randomly choose hash functions at runtime, interspersed amongst the inserts, finds, deletes! Why not

· We can, however, randomly choose a hash function each time we initialize a new hashtable.

Conclusions

- WorstEnemy never knows which hash function we will choose neither do we!
- No single input (set of keys) can always evoke worst-case behavior



UHF_a: Example

• Context: hash strings of length 3 in a table of size 131

let a = <35, 100, 21> h_a("xyz") = (35*120 + 100*121 + 21*122) % 131 = 129

Let b = <25, 90, 83> h_b("xyz") = (25*120 + 90*121 + 83*122) % 131 = 43

Thinking about UHF_a

Strengths:

- Works on any type as long as you can map keys to vectors
- If we're building a static table, we can try many values of the hash vector <a>
- Random <a> has guaranteed good properties no matter what we're hashing

Weaknesses:

- Must choose prime table size larger than any ki

Good Hashing: Universal Hash Function B (UHF_b)

Parameterized by *j*, *a*, and *b*:

- -j* size should fit into an int
- -a and b must be less than size

 $h_{j,a,b}(k) = ((ak + b) \mod (j*size))/j$

UHF_b: Example

Context: hash integers in a table of size 160

```
Let j = 32, a = 13, b = 142

h_{j,a,b}(1000) = ((13*1000 + 142) \% (32*160)) / 32

= (13142 \% 5120) / 32

= 2902 / 32

= 90
```

```
Let j = 31, a = 82, b = 112

h_{j,a,b}(1000) = ((82*1000 + 112) \% (31*160)) / 31

= (82112 \% 4960) / 31

= 2752 / 31

= 89
```

Thinking about UHF_b

Strengths

- If we're building a static table, we can try many parameter values
- Randoma, b has guaranteed good properties no matter what we're hashing
- Can choose any size table
- Very efficient if *j* and *size* are powers of 2 why?

Weaknesses

- Need to turn non-integer keys into integers

Perfect Hashing

When we know the entire key set in advance ...

Examples: programming language keywords, CD-ROM file list, spelling dictionary, etc.

... then perfect hashing lets us achieve:

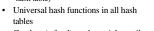
- Worst-case O(1) time complexity!
- Worst-case O(n) space complexity!



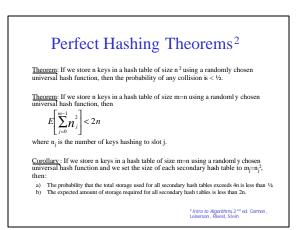
Secondary hash tables

Primary hash table

- Static set of *n* known keys
- Separate chaining, two-level hash
- Primary hash table size=n
 jth secondary hash table size=n_i²
- j⁻⁻ secondary nash table size=n_j² (where n_j keys hash to slot j in primary hash table)



• Conduct (a few!) random trials, until we get collision -free hash functions

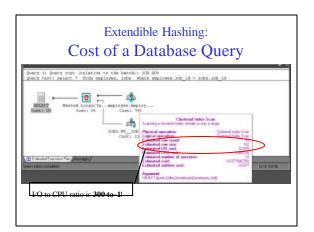


Perfect Hashing Conclusions

Perfect hashing theorems set tight expected bounds on sizes and collision behavior of all the hash tables (primary and all secondaries).

→ Conduct a few random trials of universal hash functions, by simply varying UHF parameters, until we get a set of UHFs and associated table sizes which deliver ...

- Worst-case O(1) time complexity!
- Worst-case O(n) space complexity!



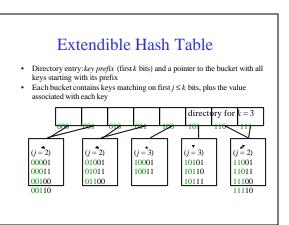


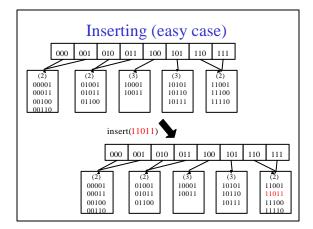
Hashing technique for huge data sets

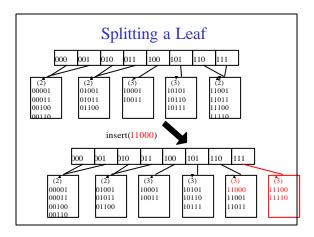
- Optimizes to reduce disk accesses
- Each hash bucket fits on one disk block
- Better than B-Trees if order is not important why?

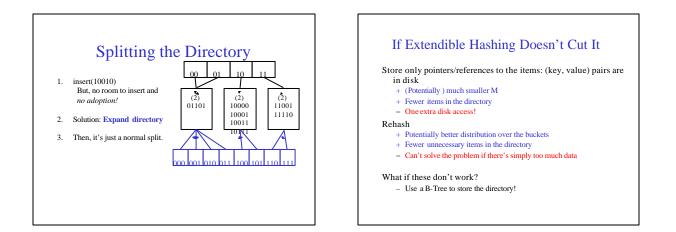
Table contains:

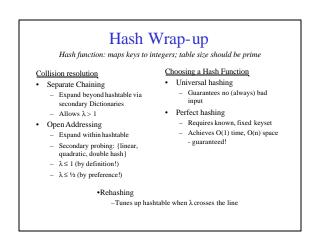
- Buckets, each fitting in one disk block, with the data
- A directory that fits in one disk block is used to hash to the correct bucket













Dictionary ADT Wrapup: Case Study

- Your company, Procrastinators Inc., will release its highly hyped word-processing program, *WordMaster2000* (yeah, they're a little behind the times), next month.
- Your highly successful alpha-test was marred by user requests for a spell-checker.
- Your mission: write and test a spell-checker module before *WordMaster2000* is released.
- For now, you only need to worry about the English language, although *WordMaster2000* is successful, you may need to port your spell-checker to other languages/character sets.

Case Study: Assumptions

You will be given a spelling dictionary of English words

- 30,000 words
- Static (ie, does not support adding user-supplied words yet)
- Arbitrary(ish) preprocessing time

Practical notes

- Almost all searches are successful Why?
- $-\,$ Words average about 8 characters in length
- 30,000 words at 8 bytes/word ~ .25 MB
- There are many regularities in the structure of English words

Case Study: Design Considerations

Issues:

- Which data structure should we use?
- What are our design goals?

Possible Solutions?