CSCI 136 Data Structures & Advanced Programming

Hashing Loose Ends

Video Outline

- Choosing an appropriate hashtable size
- Growing hashtables
- Ideal hash function properties and examples
- Revisiting hashtable performance

Hashtable Size

- Vectors are useful because, when a Vector "runs out of space", the Vector grows
 - It's very clear when we need to grow a vector: excess capacity = 0
- What does it mean for a hashtable to "run out of space"?

External Chaining



 Even ignoring correctness, performance is slowed by "full" hashtables

Hashtable Size

- The right size for our hashtable will make a trade-off between space and performance
- In addition, we would like some flexibility in case we make a bad initial guess for our size

Hashtable Fullness: Load Factor

- Suppose a hashtable with M slots stores N elements
- Load factor is a measure of how full the hash table is
 - LF = (# elements) / (# slots) = N / M
- A smaller load factor means the hashtable is less full, which likely gives better performance

Calculating Load Factor

- To track a hashtable's load factor, we can keep a running count of its elements
 - Every successful remove() decrements the count
 - Some put() operations increment the count
 - Only increment when putting new keys: replacing the value associated with an existing key doesn't change the hashtable's count
- Load factor is then (count / table.length)

Using Load Factor

- Given a hashtable's load factor, what should we do?
 - If the load factor is low, nothing!
 - A low load factor should give good performance
 - If the load factor is high (.6?), grow our table
 - Increase the number of slots without changing the number of elements (LF = N / M)
- How to grow?
 - Vectors: ensureCapacity()
 - Allocate new Object array, then copy elements to same index within new (larger) array
 - Does this work for hashtables?

Doubling Array

- Cannot just copy values
 - Why?
 - Hash values may change
 - Example: suppose (key.hashCode() == 11)
 - **II** % 8 = 3;
 - **||** % **|6** = **||**;
- **Result**: to grow our array, we must recompute the hashcode for each item, then reinsert each item into new array

Good Hashing Functions

- Important point:
 - All of our performance hinges on using "good" hash functions that spread keys "evenly"
- Good hash functions:
 - Are fast to compute
 - Uniformly distribute keys across the range
- General rules of thumb?
 - Not really. We almost always have to test "goodness" empirically.

Example Hash Functions

- What are some feasible hash functions for ASCII TABLE Strings?
 - Use the first char's ASCII value?
 - 0-255 only



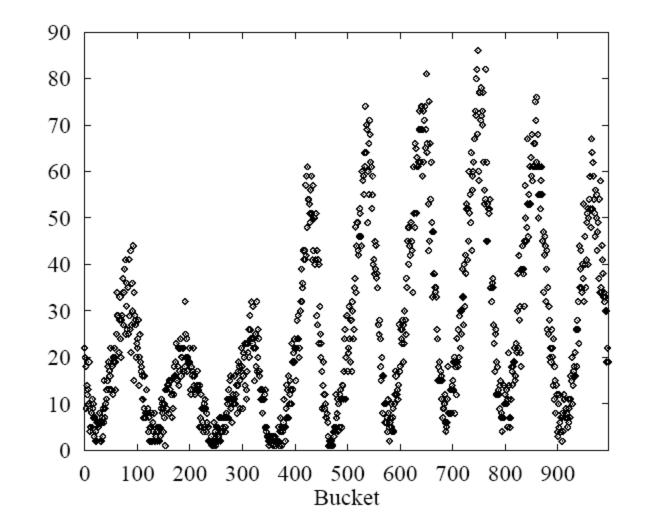
- Not uniform (some letters more popular than others)
- Sum of all characters' ASCII values?
 - Not uniform lots of small words
 - Doesn't give coverage over large array sizes
 - Not good at avoiding collisions e.g., smile, limes, miles, and slime are all the same

Example Hash Functions

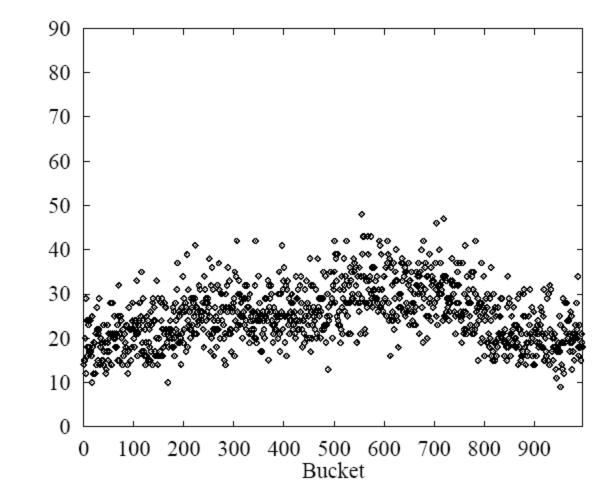
- Question: what does Java use?
 - java.lang.String uses a weighted sum
 - Small words get bigger codes
 - Distributes keys better than non-weighted sum
 - Let's look at different weights...



Hash of all words in UNIX spelling dictionary (997 buckets)

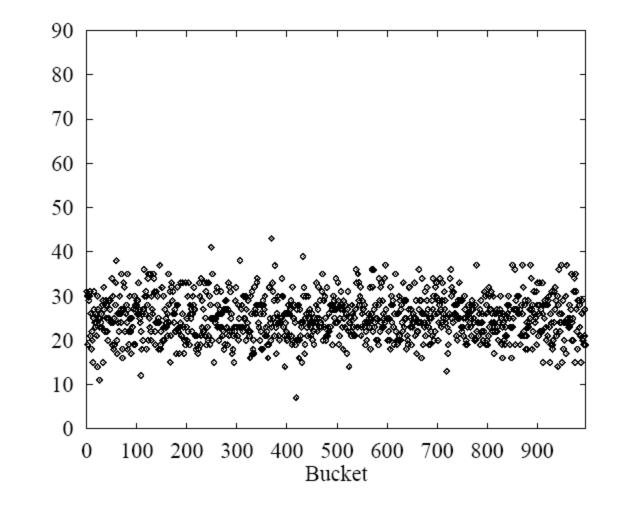






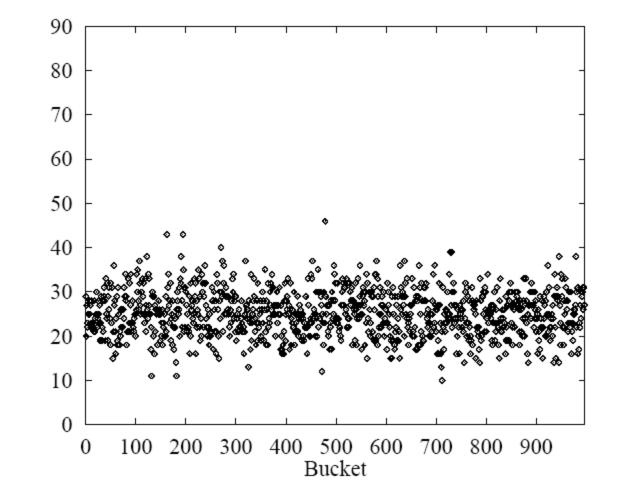


This looks pretty good, but 256ⁱ is big...





Java uses: n $\sum_{i=1}^{n} s.charAt(i) * 31^{(n-i-1)}$ i=0



Hashtables: O(I) operations?

- How long does it take to compute a String's hashCode?
 - O(s.length())
- Given an object's hash code, how long does it take to find that object?
 - O(run length) or O(chain length) PLUS cost of .equals() method to compare keys
- Conclusion: for a good hash function (fast, uniformly distributed) and a low load factor (short runs/chains), we say hashtables are O(1)

Summary

	put	get	space
unsorted vector	O(n)	O(n)	O(n)
unsorted list	O(n)	O(n)	O(n)
sorted vector	O(n)	O(log n)	O(n)
balanced BST	O(log n)	O(log n)	O(n)
hashtable	O(I)*	O(I)*	O(n)*