

Dictionaries and Hashing

The Dictionary ADT

- a dictionary (table) is an abstract model of a database or lookup table
- like a priority queue, a dictionary stores keyelement pairs
- the main operation supported by a dictionary is searching by key

Examples

- Telephone directory
- Library catalogue
- Books in print: key ISBN
- FAT (File Allocation Table)

The Dictionary ADT

- simple container methods:
 - □ size()
 - □ isEmpty()
 - iterator()
- query methods:
 - □ get(key)
 - getAllElements(key)

The Dictionary ADT

- update methods:
 - insert(key, element)
 - □ remove(key)
 - removeAllElements(key)
- special element
 - NO_SUCH_KEY, returned by an unsuccessful search

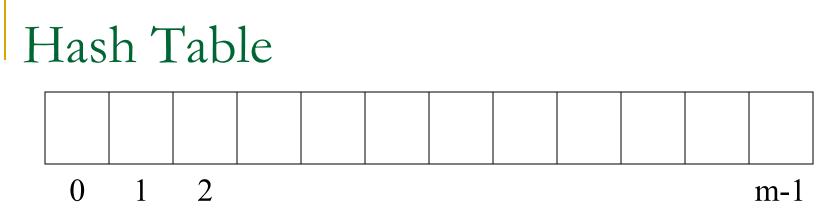
The Basic Problem

We have lots of data to store.

We desire efficient – O(1) – performance for insertion, deletion and searching.

Too much (wasted) memory is required if we use an array indexed by the data's key.

The solution is a "hash table".



Basic Idea

- The hash table is an array of size 'm'
- The storage index for an item determined by a hash function h(k): U → {0, 1, ..., m-1}

Desired Properties of h(k)

- easy to compute
- uniform distribution of keys over {0, 1, ..., m-1}

when $h(k_1) = h(k_2)$ for $k_1, k_2 \in U$, we have a collision

Division Method

- The hash function:
 - $h(k) = k \mod m$ where m is the table size.
- m must be chosen to spread keys evenly.
 - Poor choice: m = a power of 10
 - Poor choice: m = 2^b, b> 1
- A good choice of m is a prime number.
- Table should be no more than 80% full.
 - Choose m as smallest prime number greater than m_{min}, where

 m_{min} = (expected number of entries)/0.8

Multiplication Method

The hash function:

h(k) = [m(kA - [kA])]

where A is some real positive constant.

- A very good choice of A is the inverse of the "golden ratio."
- Given two positive numbers x and y, the ratio x/y is the "golden ratio" if φ = x/y = (x+y)/x
- The golden ratio:

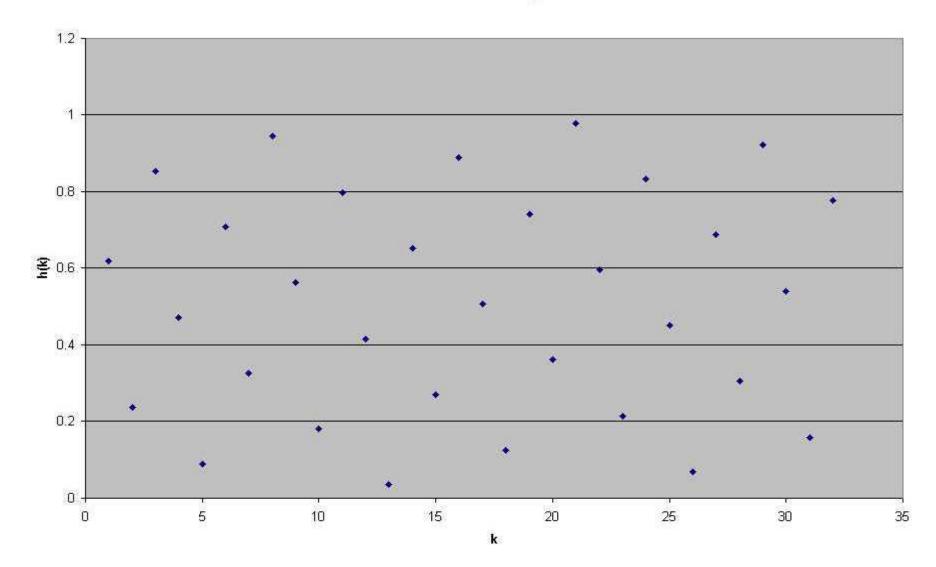
$$\begin{array}{ll} x^2 - xy - y^2 = 0 & \Rightarrow & \varphi^2 - \varphi - 1 = 0 \\ \varphi = (1 + \text{sqrt}(5))/2 & = & 1.618033989.. \\ \sim = \text{Fib}_i/\text{Fib}_{i-1} \end{array}$$

Multiplication Method (cont.)

- Because of the relationship of the golden ratio to Fibonacci numbers, this particular value of A in the multiplication method is called "Fibonacci hashing."
- Some values of

 $h(k) = \lfloor m(k \phi^{-1} - \lfloor k \phi^{-1} \rfloor) \rfloor$ = 0 for k = 0 = 0.618m for k = 1 ($\phi^{-1} = 1/1.618... = 0.618...$) = 0.236m for k = 2 = 0.854m for k = 3 = 0.472m for k = 4 = 0.090m for k = 5 = 0.708m for k = 6 = 0.326m for k = 7 = ... = 0.777m for k = 32

Fibonacci Hashing



Non-integer Keys

In order to have a non-integer key, must first convert to a positive integer:

h(k) = g(f(k)) with f: U \rightarrow integer

g: I → {0 .. m-1}

- Suppose the keys are strings.
- How can we convert a string (or characters) into an integer value?

Horner's Rule

```
static int hash (String key, int tableSize)
  int hashVal = 0;
  for (int i = 0; i < key.length(); i++)</pre>
     hashVal = 37 * hashVal + key.charAt(i);
  hashVal %= tableSize;
  if(hashVal < 0)
     hashVal += tableSize;
```

return hashVal;

value = (s[i] + 31*value) % 101;

 A. Aho, J. Hopcroft, J. Ullman, "Data Structures and Algorithms", 1983, Addison-Wesley.

'A' = 65 'h' = 104 'o' = 111

value = (s[i] + 31*value) % 101;

| | Hash | |
|-----------|-------|-----------|
| Key | Value | |
| Aho | 49 | |
| Kruse | 95 | resulting |
| Standish | 60 | table is |
| Horowitz | 28 | "sparse" |
| Langsam | 21 | |
| Sedgewick | 24 | |
| Knuth | 44 | |

value = (s[i] + 1024*value) % 128;

| | Hash | |
|-----------|-------|--------------|
| Key | Value | |
| Aho | 111 | likely to |
| Kruse | 101 | result in |
| Standish | 104 | "clustering" |
| Horowitz | 122 | clustering |
| Langsam | 109 | |
| Sedgewick | 107 | |
| Knuth | 104 | |

value = (s[i] + 3*value) % 7;

| Kev | Hash Value |
|-----------|---------------|
| Aho | 0 |
| Kruse | 5 |
| Standish | 1 |
| Horowitz | 5 |
| Langsam | 5 |
| Sedgewick | 2 |
| Knuth | 1 |

"collisions"

HashTable Class

{

public class SeparateChainingHashTable<AnyType>

```
public SeparateChainingHashTable(){/* Later */}
public SeparateChainingHashTable(int size){/*Later*/}
public void insert( AnyType x ) { /*Later*/ }
public void remove( AnyType x ) { /*Later*/}
public boolean contains( AnyType x ) {/*Later */}
public void makeEmpty() { /* Later */ }
private static final int DEFAULT TABLE SIZE = 101;
private List<AnyType> [ ] theLists;
private int currentSize;
private void rehash() { /* Later */ }
private int myhash( AnyType x ) { /* Later */ }
private static int nextPrime( int n ) { /* Later */ }
private static boolean isPrime( int n ) { /* Later */ }
```

HashTable Ops

- boolean contains (AnyType x)
 Returns true if x is present in the table.
- void insert (AnyType x)
 - □ If x already in table, do nothing.
 - Otherwise, insert it, using the appropriate hash function.
- void remove (AnyType x)
 - Remove the instance of x, if x is present.
 - Ptherwise, does nothing
- void makeEmpty()

Hash Methods

```
private int myhash( AnyType x )
{
```

```
int hashVal = x.hashCode( );
```

return hashVal;

Handling Collisions

Collisions are inevitable. How to handle them?

Separate chaining hash tables

- Store colliding items in a list.
- If m is large enough, list lengths are small.

Insertion of key k

- hash(k) to find the proper list.
- If k is in that list, do nothing, else insert k on that list.

Asymptotic performance

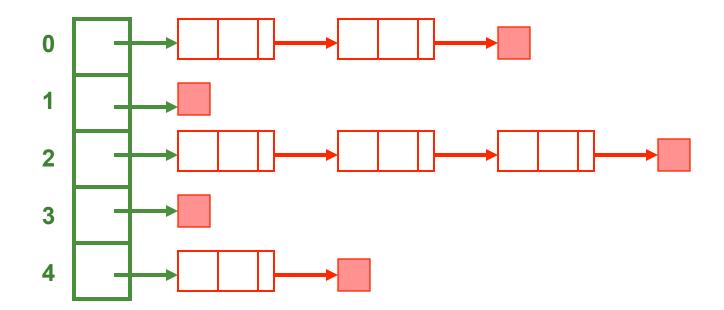
 If always inserted at head of list, and no duplicates, insert = O(1) for best, worst and average cases

Hash Class for Separate Chaining

 To implement separate chaining, the private data of the hash table is an array of Lists.
 The hash functions are written using List functions

private List<AnyType> [] theLists;

Chaining



Performance of contains()

contains

- □ Hash k to find the proper list.
- Call contains() on that list which returns a boolean.
- Performance
 - best:
 - worst:
 - average

Performance of remove()

Remove k from table

- Hash k to find proper list.
- Remove k from list.

Performance

best

worst

average

Handling Collisions Revisited

Probing hash tables

- All elements stored in the table itself (so table should be large. Rule of thumb: m >= 2N)
- Upon collision, item is hashed to a new (open) slot.

Hash function

```
h: U x {0,1,2,....} → {0,1,...,m-1}
h( k, i ) = ( h' ( k ) + f( i ) ) mod m
for some h': U → { 0, 1,..., m-1}
and some f( i ) such that f(0) = 0
```

Each attempt to find an open slot (i.e. calculating h(k, i)) is called a *probe*

HashEntry Class for Probing Hash Tables

In this case, the hash table is just an array

```
private static class HashEntry<AnyType>{
   public AnyType element; // the element
   public boolean isActive; // false if deleted
   public HashEntry( AnyType e )
   { this(e, true); }
   public HashEntry ( AnyType e, boolean active )
   { element = e; isActive = active; }
}
// The array of elements
private HashEntry<AnyType> [ ] array;
// The number of occupied cells
private int currentSize;
```

Linear Probing

Use a linear function for f(i)
 f(i) = c * i

Example:

h' (k) = k mod 10 in a table of size 10 , f(i) = i So that

 $h(k, i) = (k \mod 10 + i) \mod 10$

Insert the values U={89,18,49,58,69} into the hash table

Linear Probing (cont.)

- Problem: Clustering
 - When the table starts to fill up, performance → O
 (N)
- Asymptotic Performance
 - Insertion and unsuccessful find, average
 - λ is the "load factor" what fraction of the table is used
 - Number of probes $\approx (\frac{1}{2})(1+1/(1-\lambda)^2)$
 - if $\lambda \cong 1$, the denominator goes to zero and the number of probes goes to infinity

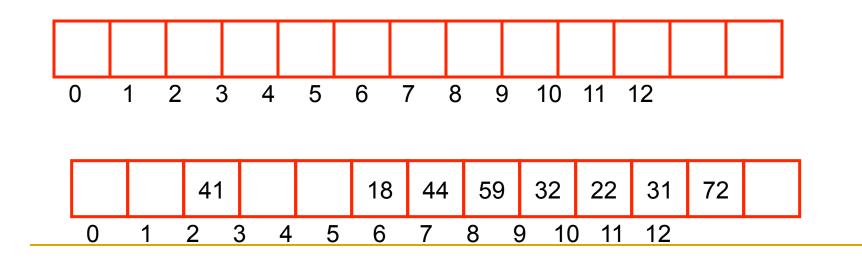
Linear Probing (cont.)

Remove

- Can't just use the hash function(s) to find the object and remove it, because objects that were inserted after X were hashed based on X's presence.
- Can just mark the cell as deleted so it won't be found anymore.
 - Other elements still in right cells
 - Table can fill with lots of deleted junk

Linear Probing Example

- h(k) = k mod 13
- Insert keys:
- 18 41 22 44 59 32 31 73



Quadratic Probing

Use a quadratic function for f(i)

$$f(i) = c_2 i^2 + c_1 i + c_0$$

The simplest quadratic function is $f(i) = i^2$

Example:

Let $f(i) = i^2$ and m = 10Let h' (k) = k mod 10

So that

h(k, i) = (k mod 10 + i^2) mod 10 Insert the value U={89, 18, 49, 58, 69} into an initially empty hash table

Quadratic Probing (cont.)

Advantage:

Reduced clustering problem

Disadvantages:

Reduced number of sequences

- No guarantee that empty slot will be found if λ ≥ 0.5, even if m is prime
- If m is not prime, may not find an empty slot even if λ < 0.5

Double Hashing

Let f(i) use another hash function

 $f(i) = i * h_2(k)$

Then h(k, I) = (h'(k) + i * h₂(k)) mod m And probes are performed at distances of h(k) 2 * h(k) 2 * h(k) 4 * h(k) ato

 $h_2(k), 2 * h_2(k), 3 * h_2(k), 4 * h_2(k), etc$

Choosing h₂(k)

- Don't allow $h_2(k) = 0$ for any k.
- A good choice:
 h₂(k) = R (k mod R) with R a prime smaller than m

Characteristics

- No clustering problem
- Requires a second hash function

Rehashing

- If the table gets too full, the running time of the basic operations starts to degrade.
- For hash tables with separate chaining, "too full" means more than one element per list (on average)
- For probing hash tables, "too full" is determined as an arbitrary value of the load factor.
- To rehash, make a copy of the hash table, double the table size, and insert all elements (from the copy) of the old table into the new table
- Rehashing is expensive, but occurs very infrequently.