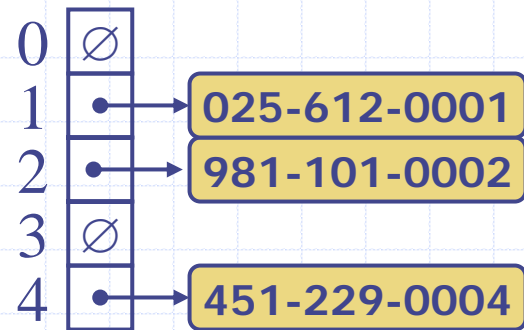
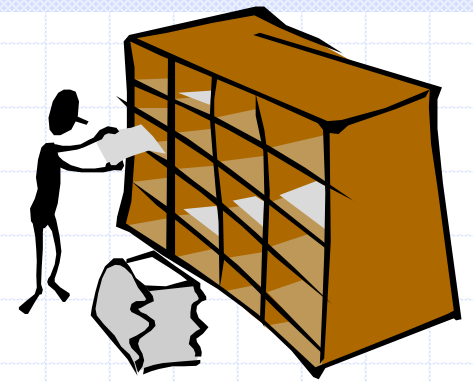


Hash Tables



Hash Functions and Hash Tables

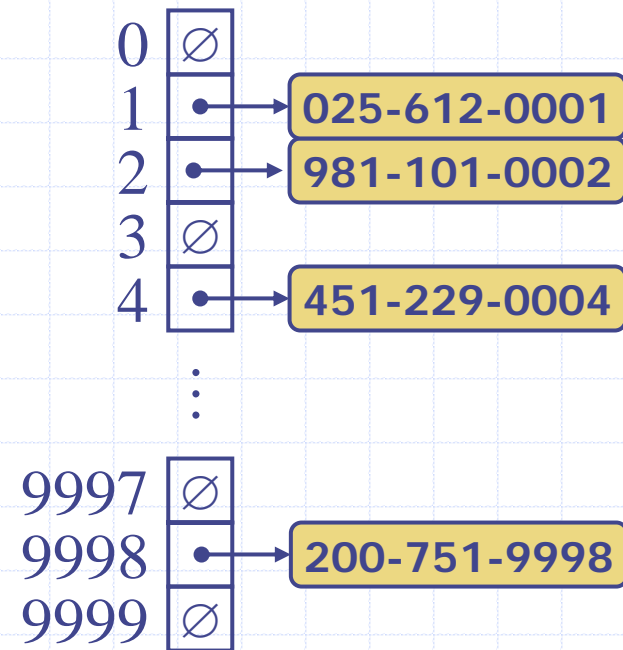


- ◆ A **hash function** h maps keys of a given type to integers in a fixed interval $[0, N - 1]$
- ◆ Example:
$$h(x) = x \bmod N$$

is a hash function for integer keys
- ◆ The integer $h(x)$ is called the **hash value** of key x
- ◆ A **hash table** for a given key type consists of
 - Hash function h
 - Array (called table) of size N
- ◆ When implementing a map with a hash table, the goal is to store item (k, o) at index $i = h(k)$

Example

- ◆ We design a hash table for a dictionary storing entries as (SSN, Name), where SSN (social security number) is a nine-digit positive integer
- ◆ Our hash table uses an array of size $N = 10,000$ and the hash function $h(x) = \text{last four digits of } x$



Hash Functions



- ◆ A hash function is usually specified as the composition of two functions:

Hash code:

$h_1: \text{keys} \rightarrow \text{integers}$

Compression function:

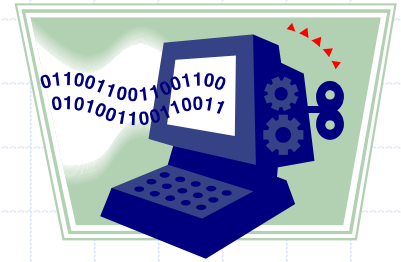
$h_2: \text{integers} \rightarrow [0, N - 1]$

- ◆ The hash code is applied first, and the compression function is applied next on the result, i.e.,

$$h(x) = h_2(h_1(x))$$

- ◆ The goal of the hash function is to “disperse” the keys in an apparently random way

Hash Codes



◆ Memory address:

- We reinterpret the memory address of the key object as an integer (default hash code of all Java objects)
- Good in general, except for numeric and string keys

◆ Integer cast:

- We reinterpret the bits of the key as an integer
- Suitable for keys of length less than or equal to the number of bits of the integer type (e.g., byte, short, int and float in Java)

◆ Component sum:

- We partition the bits of the key into components of fixed length (e.g., 16 or 32 bits) and we sum the components (ignoring overflows)
- Suitable for numeric keys of fixed length greater than or equal to the number of bits of the integer type (e.g., long and double in Java)

Hash Codes (cont.)

◆ Polynomial accumulation:

- We partition the bits of the key into a sequence of components of fixed length (e.g., 8, 16 or 32 bits)

$$a_0 a_1 \dots a_{n-1}$$

- We evaluate the polynomial

$$p(z) = a_0 + a_1 z + a_2 z^2 + \dots \\ \dots + a_{n-1} z^{n-1}$$

at a fixed value z , ignoring overflows

- Especially suitable for strings (e.g., the choice $z = 33$ gives few collisions on an application using English words)

◆ Polynomial $p(z)$ can be evaluated in $O(n)$ time using Horner's rule:

- The following polynomials are successively computed, each from the previous one in $O(1)$ time

$$p_0(z) = a_{n-1}$$

$$p_i(z) = a_{n-i-1} + z p_{i-1}(z) \\ (i = 1, 2, \dots, n-1)$$

◆ We have $p(z) = p_{n-1}(z)$



Compression Functions

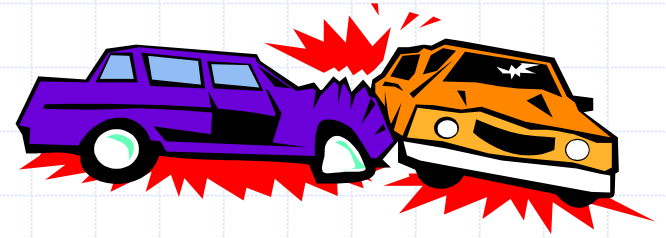
◆ Division:

- $h_2(y) = y \bmod N$
- The size N of the hash table is usually chosen to be a prime
- The reason has to do with number theory and is beyond the scope of this course

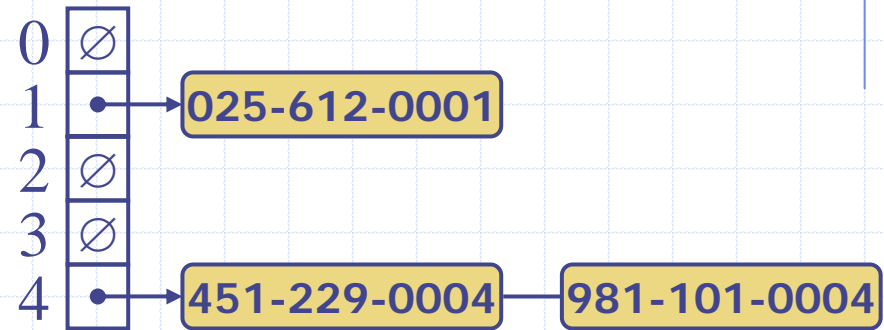
◆ Multiply, Add and Divide (MAD):

- $h_2(y) = (ay + b) \bmod N$
- a and b are nonnegative integers such that
$$a \bmod N \neq 0$$
- Otherwise, every integer would map to the same value b

Collision Handling



◆ Collisions occur when different elements are mapped to the same cell



◆ **Separate Chaining:**
let each cell in the table point to a linked list of entries that map there

◆ Separate chaining is simple, but requires additional memory outside the table

Dictionary Methods with Separate Chaining used for Collisions

◆ Delegate operations to a list-based dictionary at each cell:

Algorithm find(k):

Output: The value associated with the key k in the dictionary, or **null** if there is no entry with key equal to k in the dictionary

return $A[h(k)].get(k)$ {delegate the get to the list-based map at $A[h(k)]$ }

Algorithm insert(k, v):

Output: If there is an existing entry in our dictionary with key equal to k , then we return its value (replacing it with v); otherwise, we return **null**

$t = A[h(k)].put(k, v)$ {delegate the put to the list-based map at $A[h(k)]$ }

if $t = \text{null}$ **then** { k is a new key}

$n = n + 1$

return t

Algorithm remove(k):

Output: The (removed) value associated with key k in the dictionary, or **null** if there is no entry with key equal to k in the dictionary

$t = A[h(k)].remove(k)$ {delegate the remove to the list-based map at $A[h(k)]$ }

if $t \neq \text{null}$ **then** { k was found}

$n = n - 1$

return t

Linear Probing

- ◆ **Open addressing**: the colliding item is placed in a different cell of the table
- ◆ **Linear probing** handles collisions by placing the colliding item in the next (circularly) available table cell
- ◆ Each table cell inspected is referred to as a “probe”
- ◆ Colliding items lump together, causing future collisions to cause a longer sequence of probes

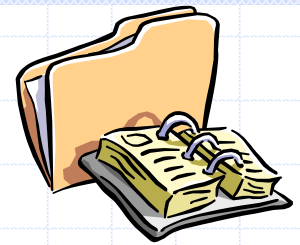
◆ Example:

- $h(x) = x \bmod 13$
- Insert keys 18, 41, 22, 44, 59, 32, 31, 73, in this order

0	1	2	3	4	5	6	7	8	9	10	11	12

↓

		41			18	44	59	32	22	31	73	
0	1	2	3	4	5	6	7	8	9	10	11	12



Search with Linear Probing

◆ Consider a hash table A that uses linear probing

◆ **find(k)**

- We start at cell $h(k)$
- We probe consecutive locations until one of the following occurs
 - ◆ An item with key k is found, or
 - ◆ An empty cell is found, or
 - ◆ N cells have been unsuccessfully probed

Algorithm *get(k)*

$i \leftarrow h(k)$

$p \leftarrow 0$

repeat

$c \leftarrow A[i]$

if $c = \emptyset$

return *null*

else if $c.key() = k$

return $c.element()$

else

$i \leftarrow (i + 1) \bmod N$

$p \leftarrow p + 1$

until $p = N$

return *null*

Updates with Linear Probing

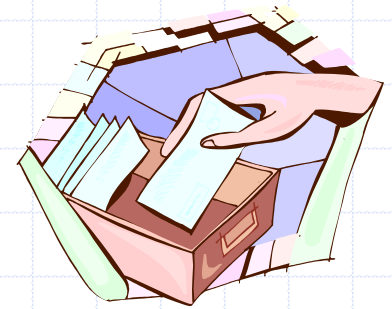
- ◆ To handle insertions and deletions, we introduce a special object, called *AVAILABLE*, which replaces deleted elements

- ◆ *remove*(k)

- We search for an entry with key k
- If such an entry (k, o) is found, we replace it with the special item *AVAILABLE* and we return element o
- Else, we return *null*

- ◆ *insert*(k, o)

- We throw an exception if the table is full
- We start at cell $h(k)$
- We probe consecutive cells until one of the following occurs
 - ◆ A cell i is found that is either empty or stores *AVAILABLE*, or
 - ◆ N cells have been unsuccessfully probed
- We store entry (k, o) in cell i



Double Hashing

- ◆ Double hashing uses a secondary hash function $d(k)$ and handles collisions by placing an item in the first available cell of the series $(i + jd(k)) \bmod N$ for $j = 0, 1, \dots, N - 1$
- ◆ The secondary hash function $d(k)$ cannot have zero values
- ◆ The table size N must be a prime to allow probing of all the cells
- ◆ Common choice of compression function for the secondary hash function:
$$d_2(k) = q - k \bmod q$$

where

 - $q < N$
 - q is a prime
- ◆ The possible values for $d_2(k)$ are $1, 2, \dots, q$

Example of Double Hashing

- Consider a hash table storing integer keys that handles collision with double hashing

- $N = 13$
 - $h(k) = k \bmod 13$
 - $d(k) = 7 - k \bmod 7$

- Insert keys 18, 41, 22, 44, 59, 32, 31, 73, in this order

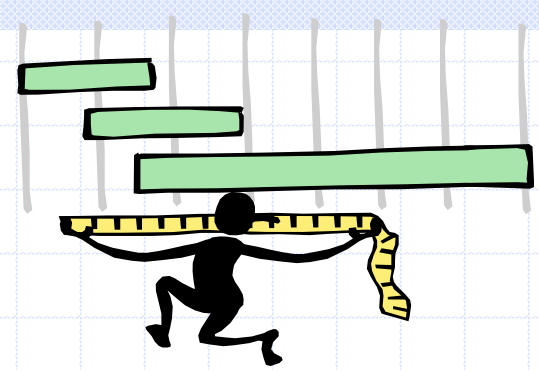
k	$h(k)$	$d(k)$	Probes	
18	5	3	5	
41	2	1	2	
22	9	6	9	
44	5	5	5	10
59	7	4	7	
32	6	3	6	
31	5	4	5	9 0
73	8	4	8	

0	1	2	3	4	5	6	7	8	9	10	11	12



31		41			18	32	59	73	22	44		
0	1	2	3	4	5	6	7	8	9	10	11	12

Performance of Hashing



- ◆ In the worst case, searches, insertions and removals on a hash table take $O(n)$ time
- ◆ The worst case occurs when all the keys inserted into the map collide
- ◆ The load factor $\alpha = n/N$ affects the performance of a hash table
- ◆ Assuming that the hash values are like random numbers, it can be shown that the expected number of probes for an insertion with open addressing is
$$1 / (1 - \alpha)$$
- ◆ The expected running time of all the dictionary ADT operations in a hash table is $O(1)$
- ◆ In practice, hashing is very fast provided the load factor is not close to 100%
- ◆ Applications of hash tables:
 - small databases
 - compilers
 - browser caches