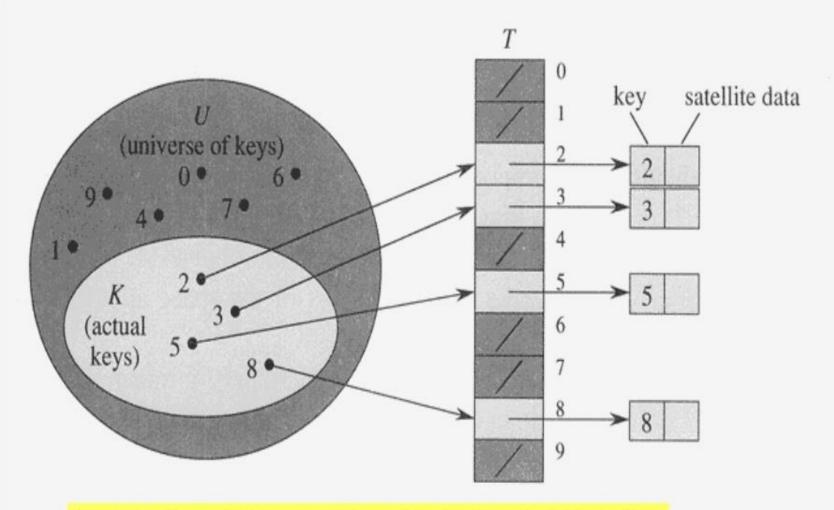
### Hash Table

#### Introduction

- Dictionary: a dynamic set that supports INSERT, SEARCH, and DELETE
- Hash table: an effective data structure for dictionaries
  - O(n) time for search in worst case
  - O(1) expected time for search

#### **Direct-Address Tables**

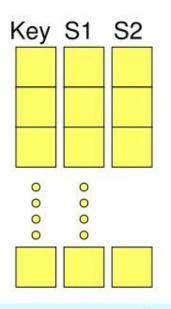
- An Array is an example of a direct-address table
  - Given a key k, the corresponding is stored in T[k]
    - Assume no two elements have the same key
  - Suitable when the universe U of keys is reasonably small
    - U={0, 1, 2,..., m-1} → T[0..m-1] → Θ(m)
    - What if the no. of keys actually occurred is small
- Operations of direct-address tables: all ⊕(1) time
  - DIRECT-ADDRESS-SEARCH(T, k): return T[k]
  - DIRECT-ADDRESS-INSERT(T, x): T[key[x]] ← x
  - DIRECT-ADDRESS-DELETE(T, x): T[key[x]] ← NIL



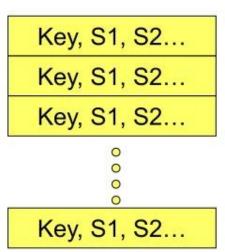
If the set contains no element with key k, T[k] = NIL

Implementing a dynamic set by a direct-address table T. Each key in the universe  $U = \{0, 1, ..., 9\}$  corresponds to an index in the table. The set  $K = \{2, 3, 5, 8\}$  of actual keys determines the slots in the table that contain pointers to elements. The other slots, heavily shaded, contain NIL.

# Alternative Illustration of Direct-Address Tables



Use separate arrays for key and satellite data if the programming language does not support objects (parallel array)



If the programming language can support object-element arrays

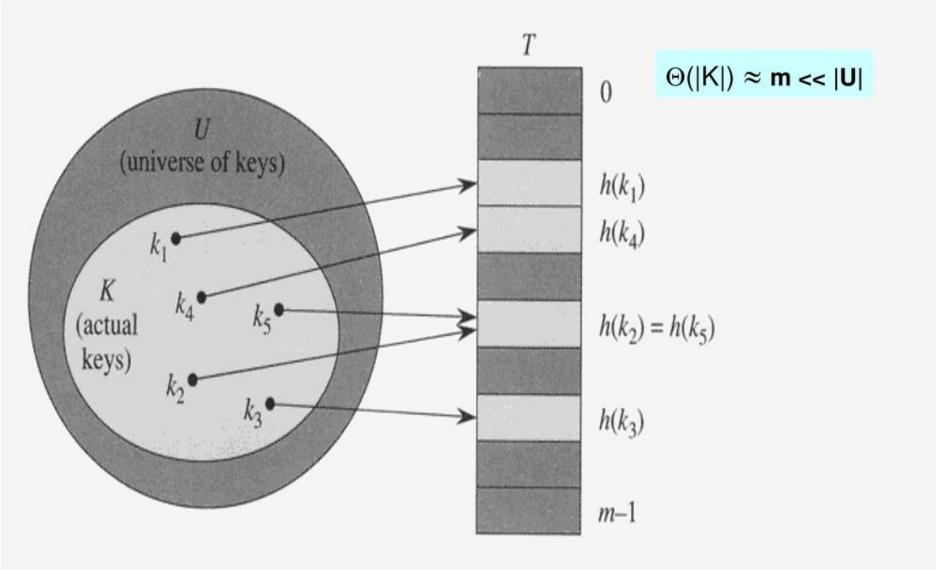
#### Hash Tables

#### Overview

- Difficulties with direct addressing
  - If U is large, storing a table T of size |U| may be impractical
  - The set K of keys actually stored may be small relative to U
- When the set K of keys stored in a dictionary is much smaller than U of all possible keys, a hash table requires much less storage than a direct-address table
  - Θ(|K|) storage
  - O(1) average time to search for an element
    - • Θ(1) worst case to search for an element in direct-address table

#### Hash Table

- Direct addressing: an element with key k → T[k]
- Hash table: an element with key k → T[h(k)]
  - h(k): hash function; usually can be computed in O(1) time
    - h: U  $\rightarrow$  {0, 1, ..., m-1}
    - An element with key k hashes to slot h(k)
    - h(k) is the hash function of key k
    - Instead of |U| values, we need to handle only m values
  - Collision may occur
    - Two different keys hash to the same slot (h(k<sub>1</sub>) = h(k<sub>2</sub>) = x)
    - Eliminating all collisions is impossible
    - But a well-designed "random"-looking hash table can minimize collisions



Using a hash function h to map keys to hash-table slots. Keys  $k_2$  and  $k_5$  map to the same slot, so they collide.

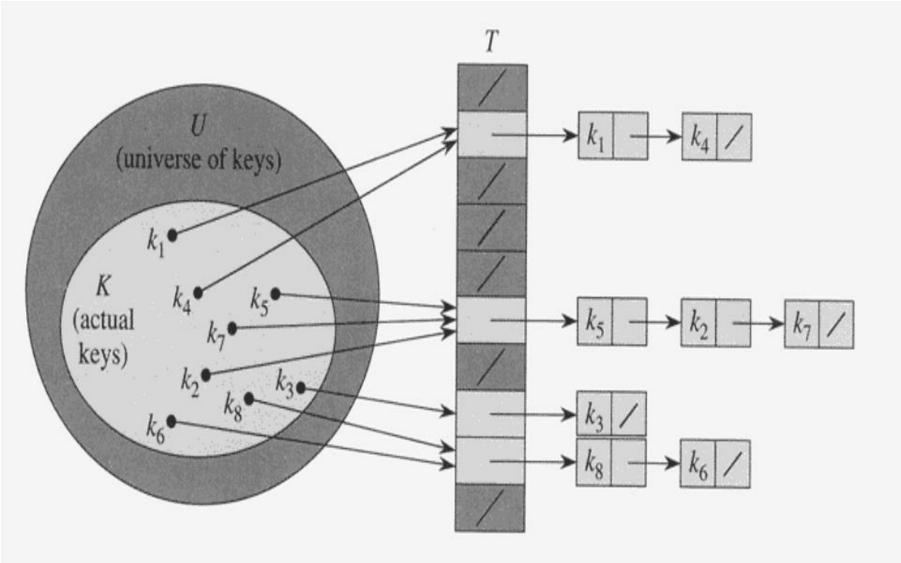
# Collision Resolution by Chaining

#### Idea

- Put all the elements that hash to the same slot in a linked list
- Slot j contains a pointer to the head of the list of all stored elements that hash to j

#### Operations

- CHAINED-HASH-INSERT(T, x): O(1) (no check for duplication)
  - Insert x at the head of list T[h(key[x])]
- CHAINED-HASH-SEARCH(T, k): proportional to the list length
  - Search for an element with key k in list T[h(k)]
- CHAINED-HASH-DELETE(T, x): O(1) for doubly linked list
  - Delete x from the list T[h(key[x])]



另一種方式:open addressing

Collision resolution by chaining. Each hash-table slot T[j] contains a linked list of all the keys whose hash value is j. For example,  $h(k_1) = h(k_4)$  and  $h(k_5) = h(k_2) = h(k_7)$ .

# Analysis of Hashing with Chaining for Searching

- Load factor α = n elements/m slots (負載因子)
  - Average number of elements stored in a chain
- Worst case: ⊕(n) + time to compute the hash function
  - all n keys hash into the same slot
- Average performance
  - How well the hash function h distributes the set of keys to be stored among the m slots, on the average
  - Assume simple uniform hashing
    - Any given element is equally likely to hash into any of the m slots, independently of where any other elements has hashed to
  - The time required for a successful or unsuccessful search is  $\Theta(1+\alpha)$ 
    - $n = O(m) \rightarrow \alpha = n/m = O(m)/m = O(1)$

#### **Hash Functions**

#### Overview

- Interpreting keys as natural numbers N={0, 1, 2,...}
  - If the keys are not natural number → convert
     For example "pt": p=112,t=116,using radix-128 integer→pt becomes 112\*128+116=14452
- What makes a good hash function?
  - Each key is equally likely to hash to any of the m slots, independently of where any other key has hashed to
    - Rarely knows the probability distribution
    - Keys may not be drawn independently
  - Hash by division: heuristic
  - Hash by multiplication: heuristic
  - Universal hashing: randomization to provide provably good performance

#### Hash By Division

- h(k) = k mod m
  - $m = 12 \text{ and } k=100 \rightarrow h(k) = 4$
  - Must avoid certain values of m → rely on characteristics of k values
    - m should not be a power of 2 → h(k) is the lowest-order bits of k
  - Good values for m are primes not too close to an 2<sup>P</sup>
  - For example: n=2000 character string, we don'n mind examing 3 elements in unsuccessful search → allocate a hash table of size 701 701 is prime near 2000/3, not near any power of 2 h(k)=k mod 701

### Hash By Multiplication

- $h(k) = \lfloor m^*(k^*A \mod 1) \rfloor$ 
  - multiply the key k by a  $A \in (0, 1)$  and extract the fractional part of kA
  - multiply the fractional part of kA by m and take the floor of the result
- The value of m is not critical
  - Typically, m = 2<sup>P</sup> (for easy implementation on computers)
  - (Knuth)  $A \approx (\sqrt{5} 1)/2 = 0.6180339887...$
  - For example:k=123456,m=10000,a=0.618
    H(k)=floor(10000\*(123456\*0.618... mod 1))
    =floor(10000\*(76300.004151...mod 1))
    =floor(10000\*0.0041151....)=41.

### Universal Hashing

#### Idea:

- Choose the hash function randomly in a way that is independent of the keys that are actually going to be stored
- Select the hash function at random from a carefully designed class of functions at the beginning of execution
  - The algorithm can behave differently on each execution, even for the same input

## Universal Hashing (Cont.)

- Let H be a finite collection of hash functions that map a given universe U of keys into the range {0, 1,..., m-1}.
- H is universal if for each pair of distinct keys k, l ∈ U, the number of hash functions h ∈H for which k(h) = k(l) is at most |H|/m
  - Collision chance = 1/m
- Theorem 11.3.
  - Suppose that a hash function h is chosen from a universal collection of hash function and is used to hash n keys into a table T of size m, using chaining to resolve collisions. If key k is not in the table, then the expected length  $E[n_{h(k)}]$  of the list that key k hashes to is at most  $\alpha$ . If key k is in the table, then the expected length  $E[n_{h(k)}]$  of the list containing key k is at most 1+  $\alpha$

# Designing A Universal Class of Hash Functions

- Steps to design a universal class of hash functions
  - Choose a prime number p, so that k ∈ [0, p-1] and p > m
  - $-Z_p=\{0, 1, ..., p-1\}$
  - $-Z_{p}^{*}=\{1, 2, ..., p-1\}$
  - $h_{a,b}(k)$  = ((ak+b) mod p) mod m, for any a ∈ $Z_p^*$  and b ∈ $Z_p$
- The family of all such hash functions is
  - $H_{p,m} = \{h_{a,b} : a \in Z_p^* \text{ and } b \in Z_p\}$
  - Total: p(p-1) hash functions in  $H_{p,m}$

## Other hash function(1)

 Mid-square: 先將數值平方在取中 間部分位元

$$(39)^2 = (100111)^2 = (10111110001)_2$$
  
 $F(39) = (11110)_2 = (29)_{10}$ 

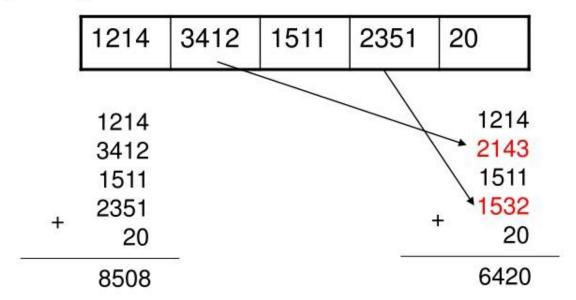
 Digital analysis:位數分析,對資料 每一位數加以分析,剔除不均勻分 布之位數,剩下位數作為hash位址

> 適合static data 且位數相同 For example

Х	F(x)
1 <mark>510</mark> 327	5107
1857384	8574
2621439	6219
1796333	7963
1038420	0380
7142481	1421
7 <mark>203</mark> 326	2036
2385425	3855

## Other hash function(2)

- Folding(摺疊法):
- (1)shift folding
- (2)folding at boundaries



# Open Addressing

#### Overview

- All elements are stored in the hash table itself
  - Each table entry contains either an set element or NIL
  - Search: systematically examine table slots until the desired element is found or it is clear that the element is not in the table
  - No lists and no elements are stored outside the table
    - Load factor α ≤ 1
  - At most m elements can be stored in the hash table
    - The extra memory freed by not storing pointers provides the hash table with a larger number of slots for the same amount of memory, potentially yielding fewer collisions and faster retrieval

#### Probe

- Insertion: successively examine (probe) the hash table until
  we find an empty slot in which to put the key
  - The sequence of positions probed relies on the key being inserted
    - Probe sequence for every key k must be a permutation of <0,1,..,m-1>
  - Extended hash function → h: U \* {0,1,...,m-1} → {0,1,..., m-1}
    - $\langle h(k,0), h(k,1),...,h(k,m-1) \rangle$
- Assume uniform hashing
  - Each key is equally likely to have any of the m! permutations of <0,1,...,m-1> as its probe sequence

# INSERT and SEARCH in Open Addressing

```
HASH-INSERT(T, k)

1 i \leftarrow 0

2 repeat j \leftarrow h(k, i)

3 if T[j] = \text{NIL}

4 then T[j] \leftarrow k

5 return j

6 else i \leftarrow i + 1

7 until i = m

8 error "hash table overflow"
```

```
HASH-SEARCH(T, k)

1 i \leftarrow 0

2 repeat j \leftarrow h(k, i)

3 if T[j] = k

4 then return j

5 i \leftarrow i + 1

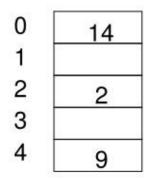
6 until T[j] = \text{NIL} or i = m

7 return NIL
```

Assume keys are not deleted from the table

### Example

h'(k)=k mod 5  
h(k,i)=(h'(k) + i) mod 5  
For k=9 
$$\rightarrow$$
 h(k,i) = <4, 0, 1, 2, 3>  
For k=3  $\rightarrow$  h(k,i) = <3, 4, 0, 1, 2>



HASH-INSERT(T,2) HASH-INSERT(T,9) HASH-INSERT(T,14)

0	14
1	
2	2
3	
4	DELETED

We have to mark a deleted slot, instead of letting it be NIL

HASH-DELETE(T, 9)

HASH-SEARCH(T, 14)

## DELETE in Open Addressing

#### Difficult !!

- Cannot simply mark the deleted slot i as empty by storing NIL
  - Impossible to retrieve any key k during whose insertion we had probed slot i and found it occupied
- Solution: mark the slot by storing in it a special value DELETED
  - Modify HASH-INSERT: treat a DELETED slot as a NIL slot
  - No modification of HASH-SEARCH is needed
    - Search times are no longer dependent on  $\alpha$
    - Therefore, chaining is more commonly selected as a collision resolution technique when keys must be deleted

### Linear Probing

- $h(k,i) = (h'(k) + i) \mod m (i = 0, 1,..., m-1)$ 
  - h': U → {0,1,...,m-1} is called an auxiliary hash function
  - $T[h'(k)] \rightarrow T[h'(k)+1] \rightarrow ... \rightarrow T[m-1] \rightarrow T[0] \rightarrow ... \rightarrow T[h'(k)-1]$
  - Only m distinct probe sequences
  - Easy to implement
  - Problem: Primary clustering
    - Long runs of occupied slots tend to get longer, and the average search time increases
  - Secondary clustering
    - If two keys have the same initial probe position, then their probe sequences are the same
  - Example:f(x)=x, table size=19(0..18), data={1.0.5.1.18.3.8.9.14.7.5.5.1.13.12.5}

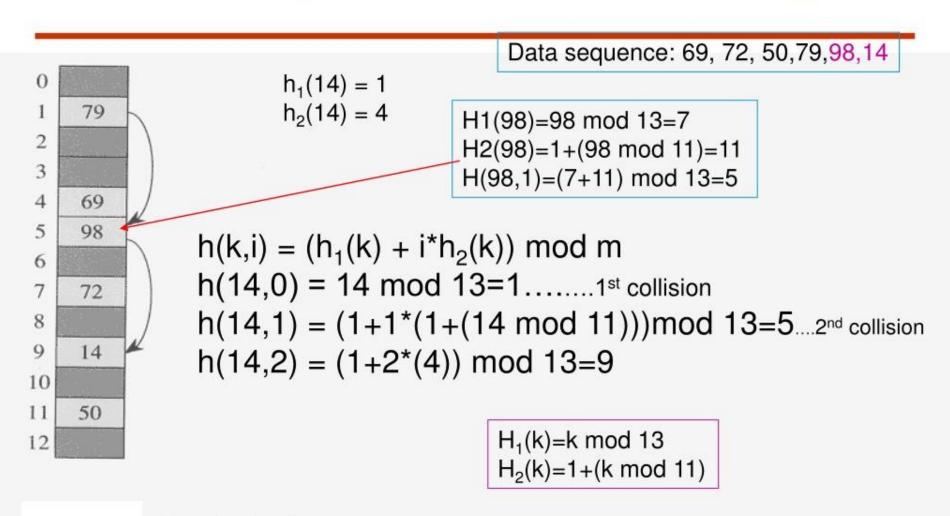
### Quadratic Probing

- $h(k,i) = (h'(k) + c_1i + c_2i^2) \mod m \ (i = 0, 1,..., m-1)$ 
  - h': auxiliary hash function;  $c_1$ ,  $c_2 \neq 0$  (auxiliary constants)
  - The values of c<sub>1</sub>, c<sub>2</sub>, m are constrained (See Problem 11-3)
  - Work much better than linear probing (alleviate primary clustering)
  - Secondary clustering
    - If two keys have the same initial probe position, then their probe sequences are the same
  - Only m distinct probe sequences

## Double Hashing: the best

- $h(k,i) = (h_1(k) + i*h_2(k)) \mod m \ (i = 0, 1,..., m-1)$ 
  - h<sub>1</sub> and h<sub>2</sub> are auxiliary hash functions
  - Depends in two ways upon the key i, since the initial probe position, the offset, or both, may vary → ⊕(m²) distinct probing sequences
  - The value of h<sub>2</sub>(k) must be relative prime to the hash-table size m for the entire hash table to be searched (Exercise 11.4-3)
    - Let m = 2<sup>P</sup> and design h<sub>2</sub> so that it always produces an odd number
    - Let m be prime and design h<sub>2</sub> so that it always produces a positive integer less than m
      - $-h_1(k) = k \mod 701$
      - $-h_2(k) = 1 + (k \mod 700)$

## Example of Double Hashing



Insertion by double hashing. Here we have a hash table of size 13 with  $h_1(k) = k \mod 13$  and  $h_2(k) = 1 + (k \mod 11)$ . Since  $14 \equiv 1 \pmod 13$  and  $14 \equiv 3 \pmod 11$ , the key 14 is inserted into empty slot 9, after slots 1 and 5 are examined and found to be occupied.

# Analysis of Open-Address Hashing

#### Theorem 11.6

- Given an open-address hash table with load factor  $\alpha = n/m < 1$ , the expected number of probes in an unsuccessful search is at most  $1/(1-\alpha)$ , assuming uniform hashing
  - If  $\alpha$  is a constant  $\rightarrow$  an unsuccessful search runs in O(1) time

#### Corollary 11.7

- Inserting an element into an open-address hash table with load factor  $\alpha$  requires at most  $1/(1-\alpha)$  probes on average, assuming uniform hashing
- Theorem 11.8
  - Given an open-address hash table with load factor  $\alpha = n/m < 1$ , the expected number of probes in an successful search is at most  $1/\alpha * \ln(1/(1-\alpha))$ , assuming uniform hashing and that each key in the table is equally likely to be searched for
    - If  $\alpha$  is a constant  $\rightarrow$  an unsuccessful search runs in O(1) time

#### Self-Study

- Proof of Theorems
- Section 11.5 Perfect Hashing
  - The worst-case number of memory accesses required to perform a search is O(1)

### example

- A hashed table is constructed using the division hash algorithm function with 5 buckets (a bucket at most 4 records). If the following key field values are to be placed in buckets: 3,5,24,22,109,10,8,6,23,28, 100, 103, 9, 39, 27, 0.
  - Identify the number of records in each bucket.
  - Which bucket overflows?

