

# Hashing Techniques

Hashing Functions & Collision Handling

# The Search Problem

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- ▶ Find items with **keys** matching a given **search key**
  - ▶ Given an array  $A$ , containing  $n$  keys, and a search key  $x$ , find the index  $i$  such as  $x=A[i]$
  - ▶ As in the case of sorting, a key could be part of a large record.

example of a record

<b>Key</b>	<b>other data</b>
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# Applications

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- ▶ Keeping track of customer account information at a bank
  - ▶ Search through records to check balances and perform transactions
- ▶ Keep track of reservations on flights
  - ▶ Search to find empty seats, cancel/modify reservations
- ▶ Search engine
  - ▶ Looks for all documents containing a given word



# Special Case: Dictionaries

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- ▶ **Dictionary** = data structure that supports mainly two basic operations: **insert** a new item and **return an item with a given key**
- ▶ Queries: return information about the set S:
  - ▶ Search (S, k)
  - ▶ Minimum (S), Maximum (S)
  - ▶ Successor (S, x), Predecessor (S, x)
- ▶ Modifying operations: change the set
  - ▶ Insert (S, k)
  - ▶ Delete (S, k) – **not very often**



# Direct Addressing

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## ► Assumptions:

- ▶ Key values are distinct
- ▶ Each key is drawn from a universe  $U = \{0, 1, \dots, m - 1\}$

## ► Idea:

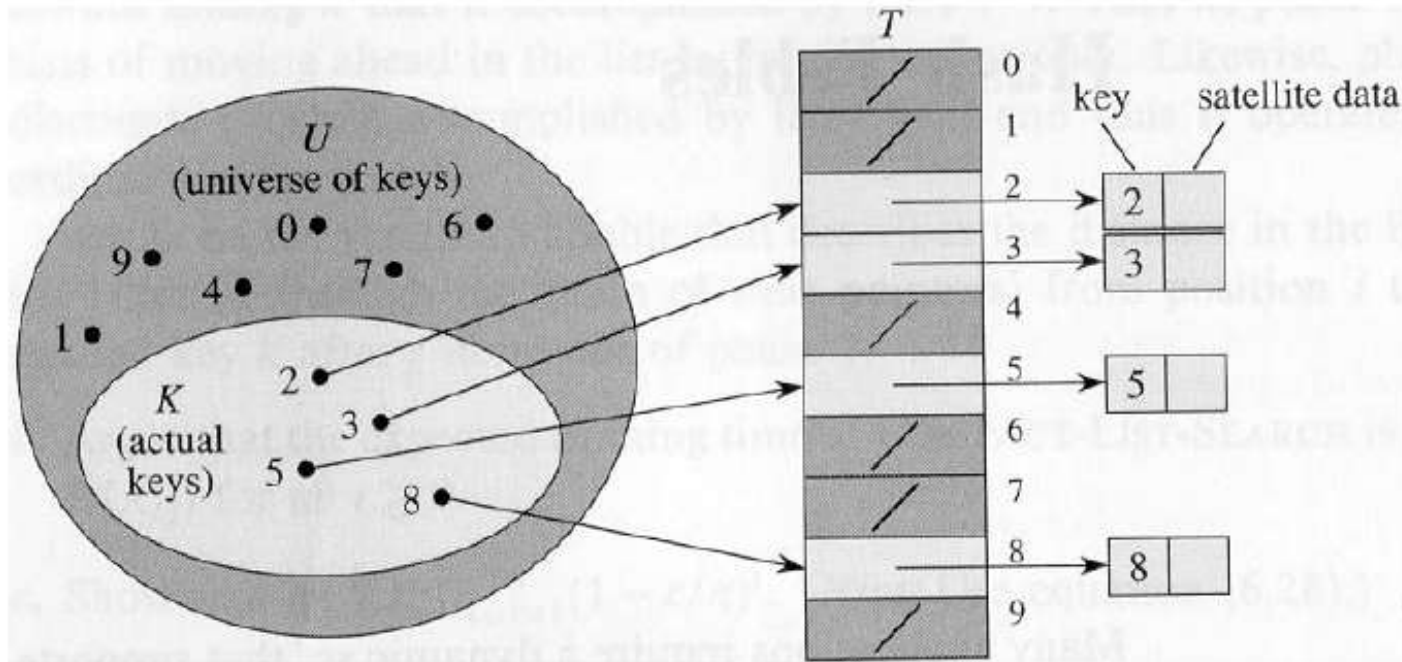
- ▶ Store the items in an array, indexed by keys

## • **Direct-address table** representation:

- An array  $T[0 \dots m - 1]$
- Each **slot**, or position, in  $T$  corresponds to a key in  $U$
- For an element  $x$  with key  $k$ , a pointer to  $x$  (or  $x$  itself) will be placed in location  $T[k]$
- If there are no elements with key  $k$  in the set,  $T[k]$  is empty, represented by **NIL**



# Direct Addressing (cont'd)



(insert/delete in  $O(1)$  time)

# Operations

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*Alg.:* DIRECT-ADDRESS-SEARCH( $T, k$ )  
**return**  $T[k]$

*Alg.:* DIRECT-ADDRESS-INSERT( $T, x$ )  
 $T[\text{key}[x]] \leftarrow x$

*Alg.:* DIRECT-ADDRESS-DELETE( $T, x$ )  
 $T[\text{key}[x]] \leftarrow \text{NIL}$

► Running time for these operations:  $O(1)$



# Comparing Different Implementations

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- ▶ Implementing dictionaries using:
  - ▶ Direct addressing
  - ▶ Ordered/unordered arrays
  - ▶ Ordered/unordered linked lists

	Insert	Search
direct addressing	$O(1)$	$O(1)$
ordered array	$O(N)$	$O(\lg N)$
ordered list	$O(N)$	$O(N)$
unordered array	$O(1)$	$O(N)$
unordered list	$O(1)$	$O(N)$



# Examples Using Direct Addressing

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## Example 1:

- (i) Suppose that the keys are integers from 1 to 100 and that there are about 100 records
- (ii) Create an array  $A$  of 100 items and store the record whose key is equal to  $i$  in  $A[i]$

## Example 2:

- (i) Suppose that the keys are nine-digit social security numbers
- (ii) We can use the same strategy as before but it very inefficient now: an array of 1 billion items is needed to store 100 records !!

- $|U|$  can be very large
- $|K|$  can be much smaller than  $|U|$

# Hash Tables

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- ▶ When  $K$  is much smaller than  $U$ , a **hash table** requires much less space than a **direct-address table**
  - ▶ Can reduce storage requirements to  $|K|$
  - ▶ Can still get  $O(1)$  search time, but on the average case, not the worst case



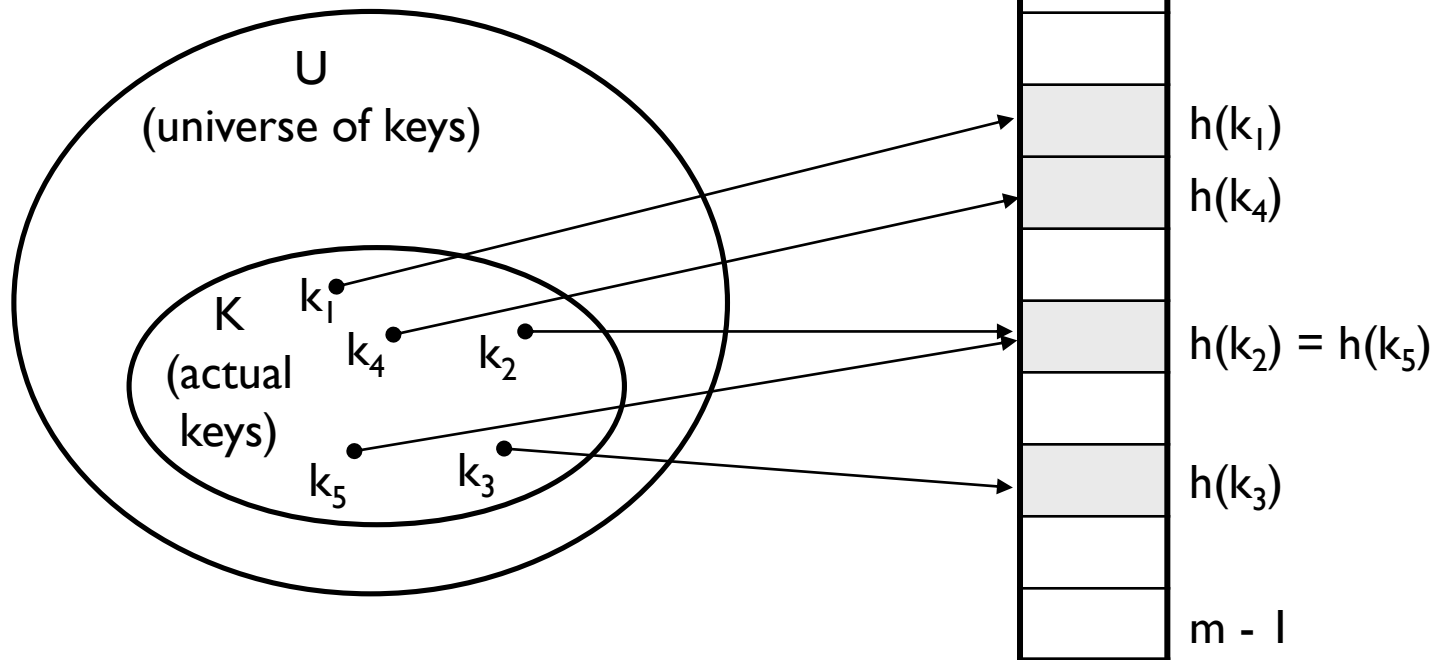
# Hash Tables

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## Idea:

- ▶ Use a function  $h$  to compute the slot for each key
- ▶ Store the element in slot  $h(k)$
- ▶ A **hash function**  $h$  transforms a key into an index in a hash table  $T[0..m-1]$ :
$$h : U \rightarrow \{0, 1, \dots, m - 1\}$$
- ▶ We say that  $k$  **hashes** to slot  $h(k)$
- ▶ Advantages:
  - ▶ Reduce the range of array indices handled:  **$m$  instead of  $|U|$**
  - ▶ Storage is also reduced

# Example: HASH TABLES





## Revisit Example 2

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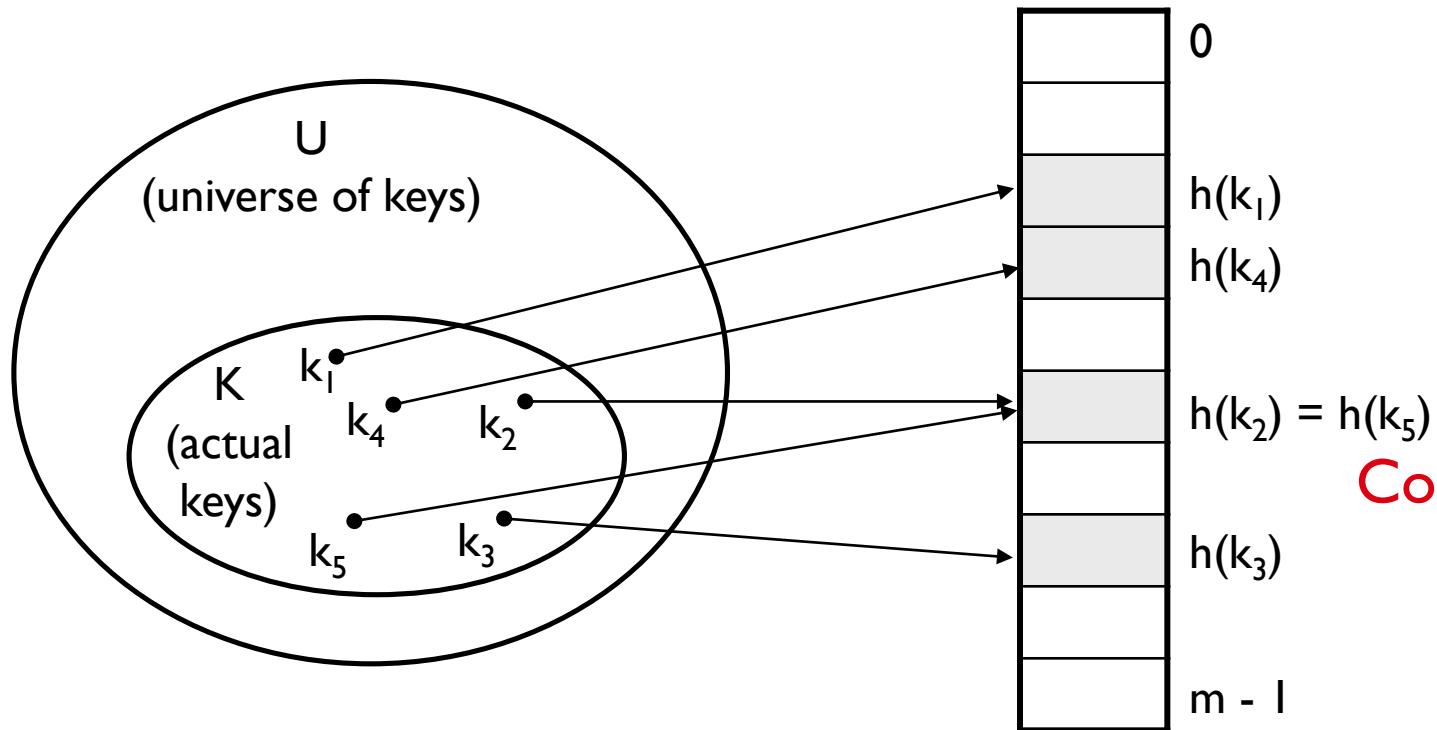
Suppose that the keys are nine-digit social security numbers

Possible hash function

$$h(ssn) = ssn \bmod 100 \text{ (last 2 digits of ssn)}$$

e.g., if  $ssn = 10123411$  then  $h(10123411) = 11$

Do you see any problems with this approach?



**Collisions!**

# Collisions

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- ▶ Two or more keys hash to the same slot!!
- ▶ For a given set  $K$  of keys
  - ▶ If  $|K| \leq m$ , collisions may or may not happen, depending on the hash function
  - ▶ If  $|K| > m$ , collisions will definitely happen (i.e., there must be at least two keys that have the same hash value)

Avoiding collisions completely is hard, even with a good hash function



# Handling Collisions

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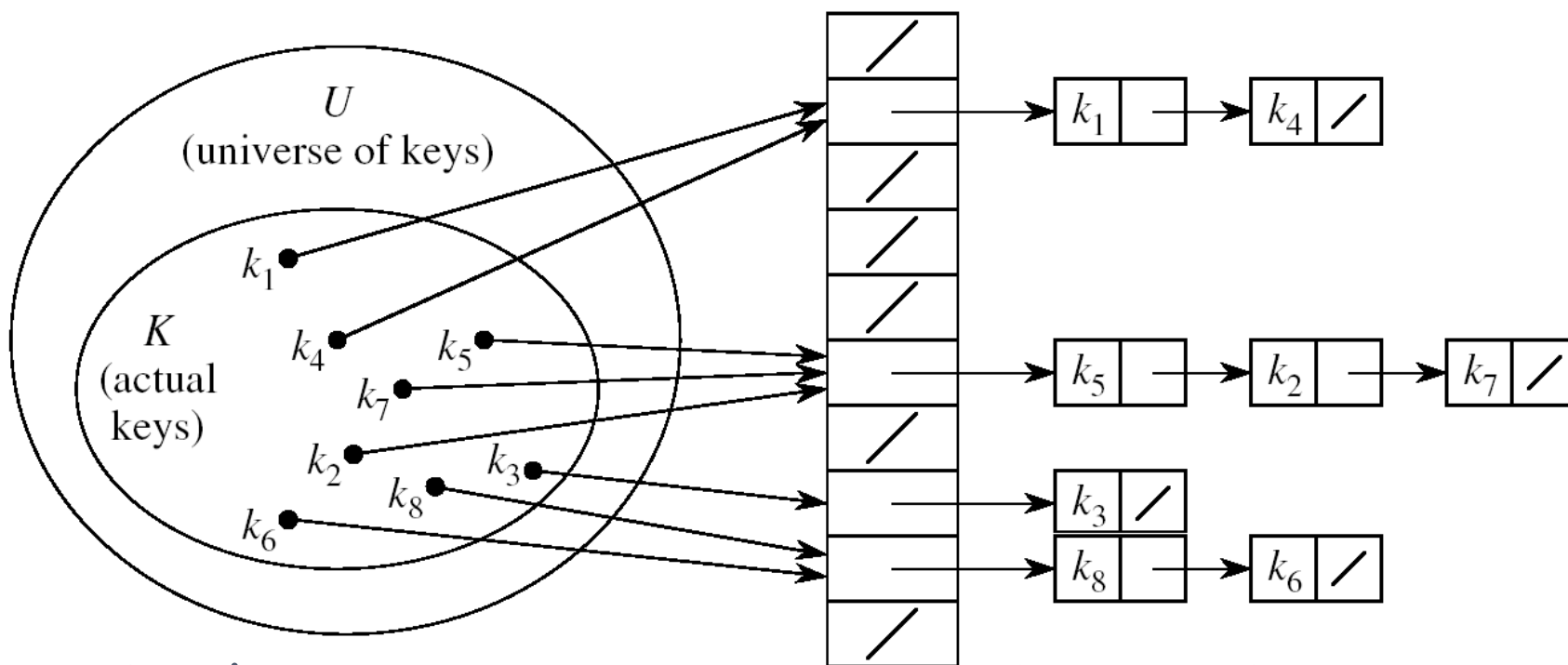
- ▶ We will review the following methods:
  - ▶ Chaining
  - ▶ Open addressing
    - ▶ Linear probing
    - ▶ Quadratic probing
    - ▶ Double hashing
- ▶ We will discuss **chaining** first, and ways to build “good” functions.



# Handling Collisions Using Chaining

## ► Idea:

- Put all elements that hash to the same slot into a linked list



- Slot  $j$  contains a pointer to the head of the list of all elements that hash to  $j$



# Collision with Chaining - Discussion

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- ▶ Choosing the size of the table
  - ▶ Small enough not to waste space
  - ▶ Large enough such that lists remain short
  - ▶ Typically  $1/5$  or  $1/10$  of the total number of elements
- ▶ How should we keep the lists: ordered or not?
  - ▶ Not ordered!
    - ▶ Insert is fast
    - ▶ Can easily remove the most recently inserted elements

# Insertion in Hash Tables

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*Alg.:* CHAINED-HASH-INSERT( $T, x$ )

insert  $x$  at the head of list  $T[h(\text{key}[x])]$

- ▶ Worst-case running time is  $O(1)$
- ▶ Assumes that the element being inserted isn't already in the list
- ▶ It would take an additional search to check if it was already inserted



# Deletion in Hash Tables

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*Alg.:* CHAINED-HASH-DELETE( $T, x$ )

delete  $x$  from the list  $T[h(\text{key}[x])]$

- ▶ Need to find the element to be deleted.
- ▶ Worst-case running time:
  - ▶ Deletion depends on searching the corresponding list



# Searching in Hash Tables

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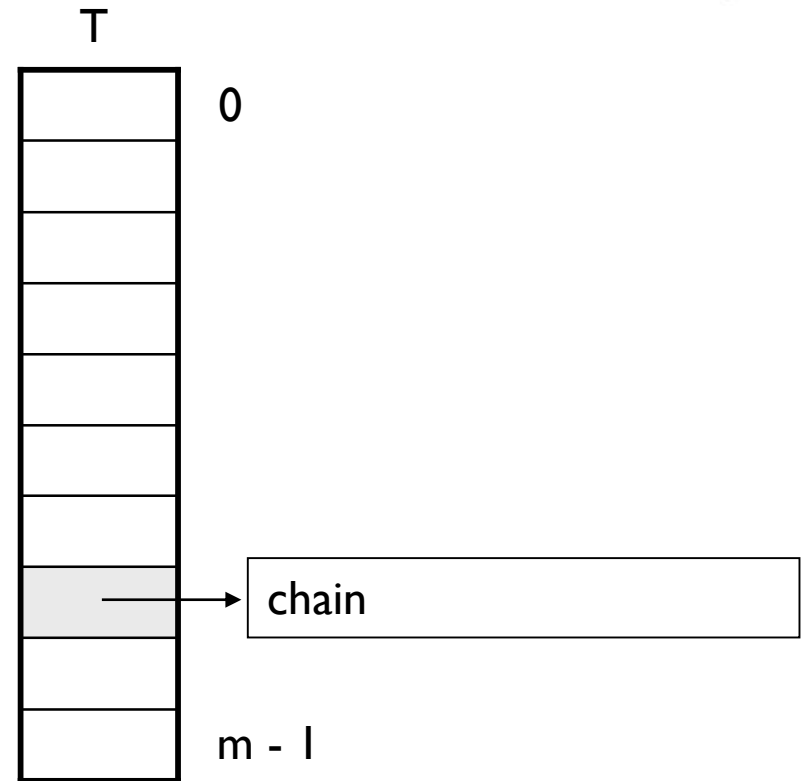
*Alg.:* CHAINED-HASH-SEARCH( $T, k$ )

search for an element with key  $k$  in list  $T[h(k)]$

- ▶ Running time is proportional to the length of the list of elements in slot  $h(k)$

# Analysis of Hashing with Chaining: **Worst Case**

- ▶ How long does it take to search for an element with a given key?
- ▶ Worst case:
  - ▶ All  $n$  keys hash to the same slot
  - ▶ Worst-case time to search is  $\Theta(n)$ , plus time to compute the hash function



# Analysis of Hashing with Chaining :Average Case

## ► Average case

- depends on how well the hash function distributes the  $n$  keys among the  $m$  slots

## ► **Simple uniform hashing** assumption:

- Any given element is equally likely to hash into any of the  $m$  slots (i.e., probability of collision  $\Pr(h(x)=h(y))$ , is  $1/m$ )

## ► Length of a list:

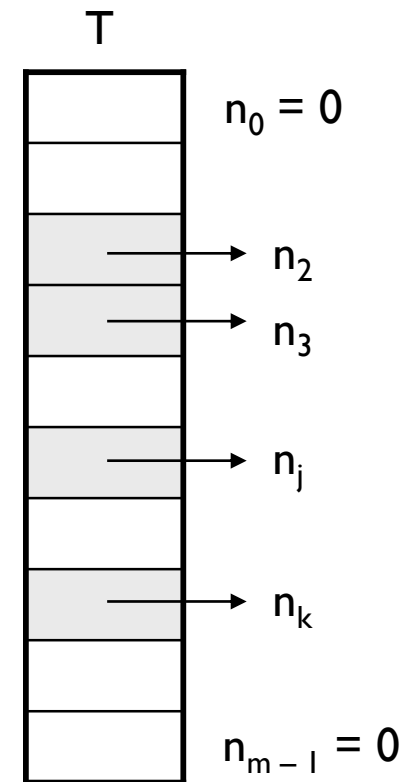
$$T[j] = n_j, \quad j = 0, 1, \dots, m-1$$

## ► Number of keys in the table:

$$n = n_0 + n_1 + \dots + n_{m-1}$$

## ► Expected value of $n_j$ :

$$E[n_j] = a = n/m$$

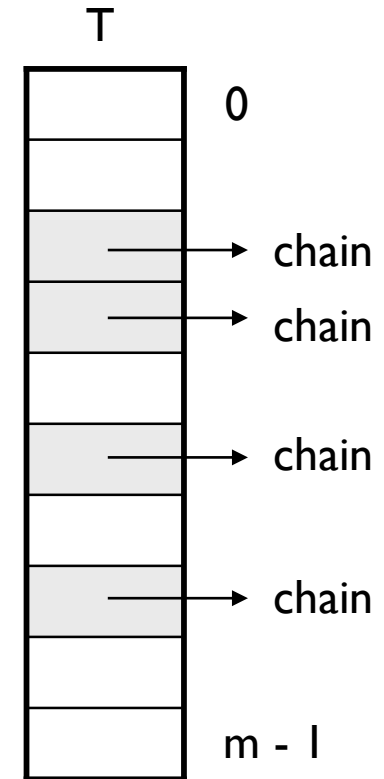


# Load Factor of a Hash Table

- ▶ Load factor of a hash table T:

$$\alpha = n/m$$

- ▶  $n$  = # of elements stored in the table
- ▶  $m$  = # of slots in the table = # of linked lists
- ▶  $\alpha$  encodes the average number of elements stored in a chain
- ▶  $\alpha$  can be  $<$ ,  $=$ ,  $> 1$





# Case 1: Unsuccessful Search (i.e., item not stored in the table)

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## Theorem

An unsuccessful search in a hash table takes expected time  $\Theta(1 + \alpha)$  under the assumption of simple uniform hashing

(i.e., probability of collision  $\Pr(h(x)=h(y))$ , is  $1/m$ )

## Proof

- ▶ Searching unsuccessfully for any key  $k$ 
  - ▶ need to search to the end of the list  $T[h(k)]$
- ▶ Expected length of the list:
  - ▶  $E[n_{h(k)}] = \alpha = n/m$
- ▶ Expected number of elements examined in an unsuccessful search is  $\alpha$
- ▶ Total time required is:
  - ▶  $O(1)$  (for computing the hash function) +  $\alpha \rightarrow \Theta(1 + \alpha)$





## Case 2: Successful Search

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Successful search:  $\Theta(1 + \frac{a}{2}) = \Theta(1 + a)$  time on the average

(search half of a list of length  $a$  plus  $O(1)$  time to compute  $h(k)$ )



# Analysis of Search in Hash Tables

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- ▶ If  $m$  (# of slots) is proportional to  $n$  (# of elements in the table):

- ▶  $n = O(m)$

- ▶  $\alpha = n/m = O(m)/m = O(1)$

⇒ Searching takes constant time on average



# Hash Functions

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- ▶ A hash function transforms a key into a table address
- ▶ **What makes a good hash function?**
  - (1) Easy to compute
  - (2) Approximates a random function: for every input, every output is equally likely (**simple uniform hashing**)
- ▶ In practice, it is very hard to satisfy the simple uniform hashing property
  - ▶ i.e., we don't know in advance the probability distribution that keys are drawn from

# Good Approaches for Hash Functions

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- ▶ Minimize the chance that closely related keys hash to the same slot
  - ▶ Strings such as **pt** and **pts** should hash to different slots
- ▶ **Derive a hash value that is independent from any patterns that may exist in the distribution of the keys**



# The Division Method

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## ► **Idea:**

- Map a key  $k$  into one of the  $m$  slots by taking the remainder of  $k$  divided by  $m$

$$h(k) = k \bmod m$$

## ► **Advantage:**

- fast, requires only one operation

## ► **Disadvantage:**

- Certain values of  $m$  are bad, e.g.,
  - power of 2
  - non-prime numbers

# Example - The Division Method

► If  $m = 2^p$ , then  $h(k)$  is just the least significant  $p$  bits of  $k$

►  $p = 1 \Rightarrow m = 2$

$\Rightarrow h(k) = \{0, 1\}$ , least significant 1 bit of  $k$

►  $p = 2 \Rightarrow m = 4$

$\Rightarrow h(k) = \{0, 1, 2, 3\}$ , least significant 2 bits of  $k$

• Choose  $m$  to be a prime, not close to a power of 2

• Column 2:  $k \bmod 97$

• Column 3:  $k \bmod 100$

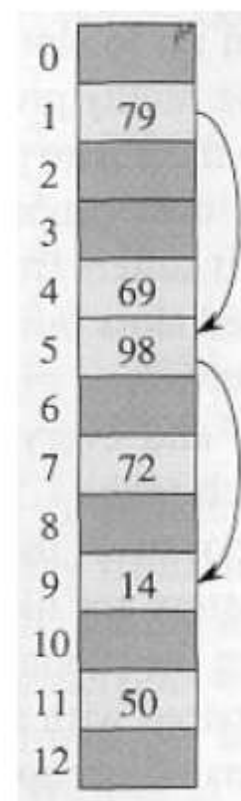
	m	m
	97	100
16838	57	38
5758	35	58
10113	25	13
17515	55	15
31051	11	51
5627	1	27
23010	21	10
7419	47	19
16212	13	12
4086	12	86
2749	33	49
12767	60	67
9084	63	84
12060	32	60
32225	21	25
17543	83	43
25089	63	89
21183	37	83
25137	14	37
25566	55	66
26966	0	66
4978	31	78
20495	28	95
10311	29	11
11367	18	67



# Open Addressing

- ▶ If we have enough contiguous memory to store all the keys ( $m \geq N$ )  $\Rightarrow$  **store the keys in the table itself**
- ▶ No need to use linked lists anymore
- ▶ Basic idea:
  - ▶ Insertion: if a slot is full, try another one, until you find an empty one
  - ▶ Search: follow the same sequence of probes
  - ▶ Deletion: more difficult ... (we'll see why)
- ▶ Search time depends on the length of the probe sequence!

e.g., insert 14





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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
```

# Generalize hash function notation:

- ▶ A hash function contains two arguments now:

(i) Key value, and (ii) Probe number

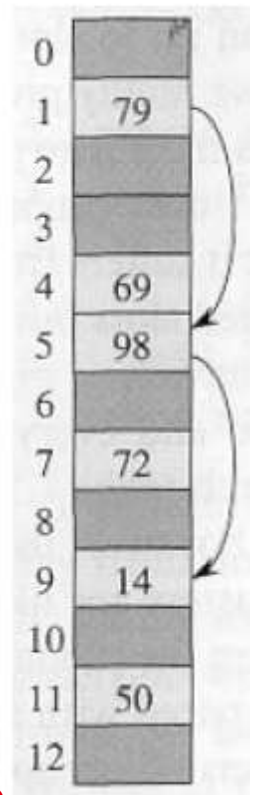
$$h(k,p), \quad p=0,1,\dots,m-1$$

- ▶ Probe sequences

$$\langle h(k,0), h(k,1), \dots, h(k,m-1) \rangle$$

- ▶ Must be a permutation of  $\langle 0,1,\dots,m-1 \rangle$
- ▶ There are  $m!$  possible permutations
- ▶ Good hash functions should be able to produce all  $m!$  probe sequences

insert 14

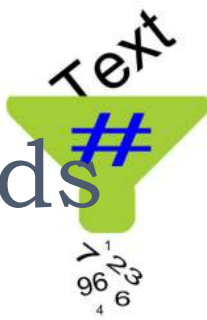


Example

$\langle 1, 5, 9 \rangle$

# Common Open Addressing Methods

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- ▶ Linear probing
- ▶ Quadratic probing
- ▶ Double hashing

# Linear probing: Inserting a key

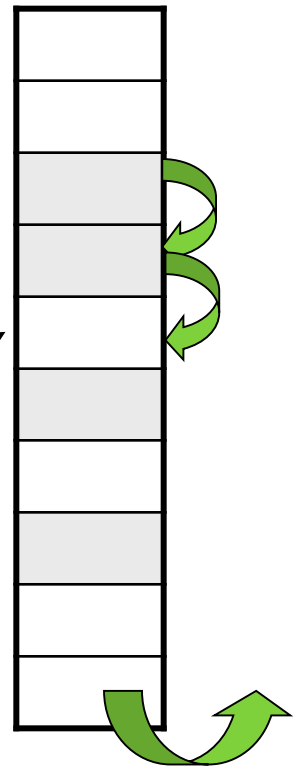
- ▶ Idea: when there is a collision, check the next available position in the table (i.e., probing)

$$h(k,i) = (h^l(k) + i) \bmod m$$

$$i=0,1,2,\dots$$

- ▶ First slot probed:  $h^l(k)$
- ▶ Second slot probed:  $h^l(k) + 1$
- ▶ Third slot probed:  $h^l(k)+2$ , and so on

probe sequence:  $\langle h^l(k), h^l(k)+1, h^l(k)+2, \dots \rangle$



wrap around



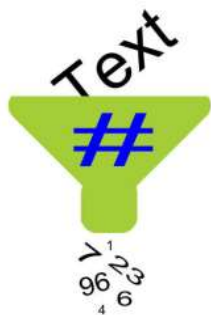
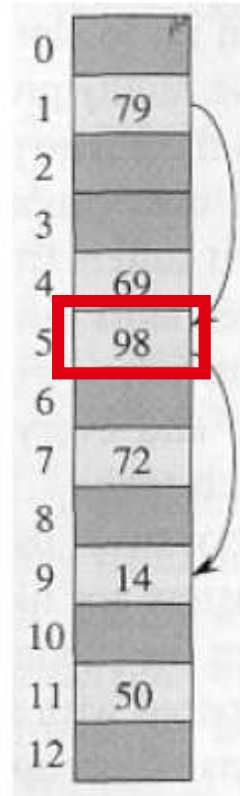
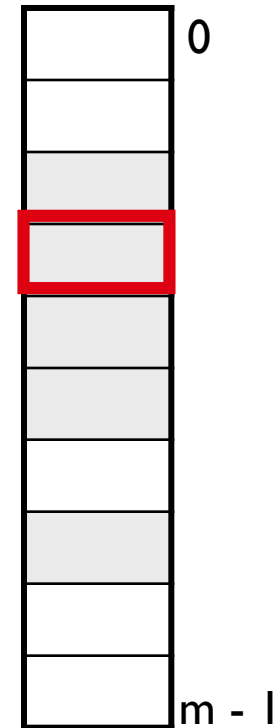
# Linear probing: Deleting a key

## ► Problems

- ▶ Cannot mark the slot as empty
- ▶ Impossible to retrieve keys inserted after that slot was occupied

## ► Solution

- ▶ Mark the slot with a sentinel value DELETED
- ▶ The deleted slot can later be used for insertion
- ▶ Searching will be able to find all the keys

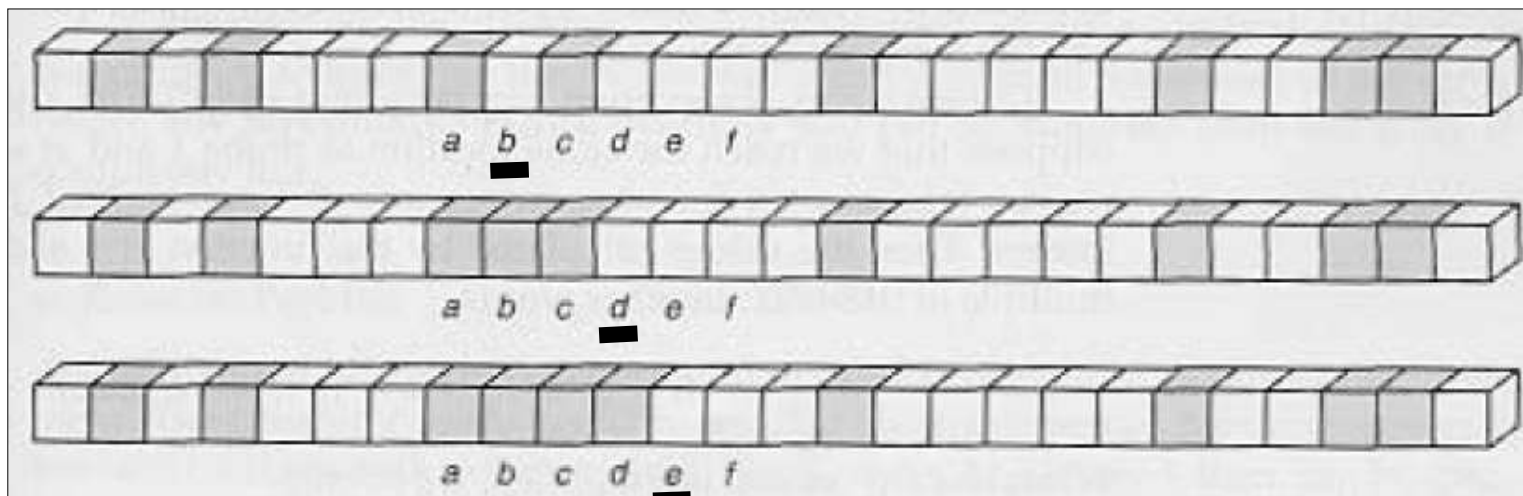


# Primary Clustering Problem

- ▶ Some slots become more likely than others
- ▶ Long chunks of occupied slots are created

⇒ search time increases!!

initially, all slots have probability  $1/m$



Slot b:  
 $2/m$

Slot d:  
 $4/m$

Slot e:  
 $5/m$

# Quadratic probing

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$h(k, i) = (h'(k) + c_1 i + c_2 i^2) \bmod m$ , where  $h': U \rightarrow (0, 1, \dots, m-1)$

$i=0, 1, 2, \dots$

- Clustering problem is less serious but still an issue (*secondary clustering*)
- How many probe sequences quadratic probing generate ?  $m$   
(the initial probe position determines the probe sequence)



# Double Hashing

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- (1) Use one hash function to determine the first slot
- (2) Use a second hash function to determine the increment for the probe sequence

$$h(k,i) = (h_1(k) + i h_2(k) ) \bmod m, \quad i=0,1,\dots$$

- ▶ Initial probe:  $h_1(k)$
- ▶ Second probe is offset by  $h_2(k) \bmod m$ , so on ...
- ▶ **Advantage:** avoids clustering

# Double Hashing: Example

$$h_1(k) = k \bmod 13$$

$$h_2(k) = 1 + (k \bmod 11)$$

$$h(k,i) = (h_1(k) + i h_2(k)) \bmod 13$$

► Insert key 14:

$$h_1(14,0) = 14 \bmod 13 = 1$$

$$\begin{aligned} h(14,1) &= (h_1(14) + h_2(14)) \bmod 13 \\ &= (1 + 4) \bmod 13 = 5 \end{aligned}$$

$$\begin{aligned} h(14,2) &= (h_1(14) + 2 h_2(14)) \bmod 13 \\ &= (1 + 8) \bmod 13 = 9 \end{aligned}$$

0	
1	79
2	
3	
4	69
5	98
6	
7	72
8	
9	14
10	
11	50
12	

Thank you!

