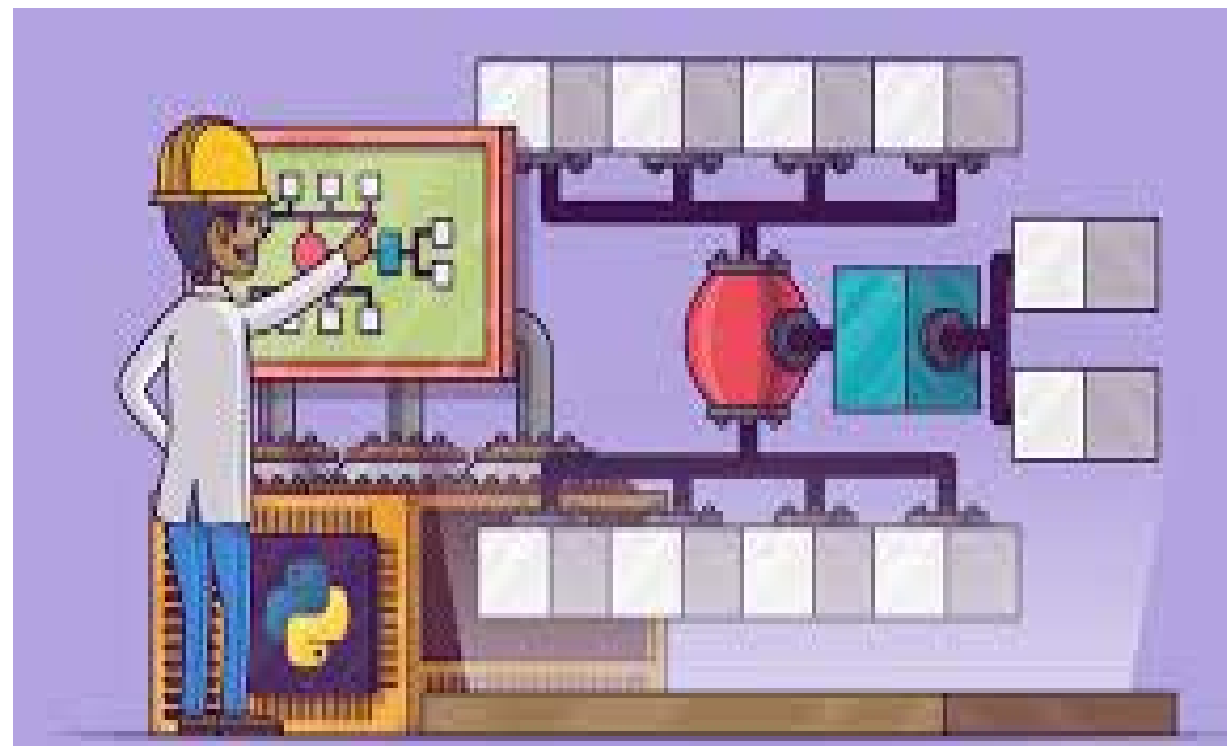




ساختمان داده ها

مدرس:
سمانه حسینی سمنانی

دانشگاه صنعتی اصفهان - دانشکده برق و
کامپیوتر





Hashing – درهم سازی

- Direct-address tables
- Hash tables
- Hash functions
- Open addressing



Hashing – درهم سازی

- Many applications require : INSERT, SEARCH, and DELETE.
- For example, a compiler that translates a programming language maintains a symbol table, in which the keys of elements are arbitrary character strings corresponding to identifiers in the language.
- A hash table is an effective data structure for implementing dictionaries.
- Under reasonable assumptions, the average time to search for an element in a hash table is $O(1)$

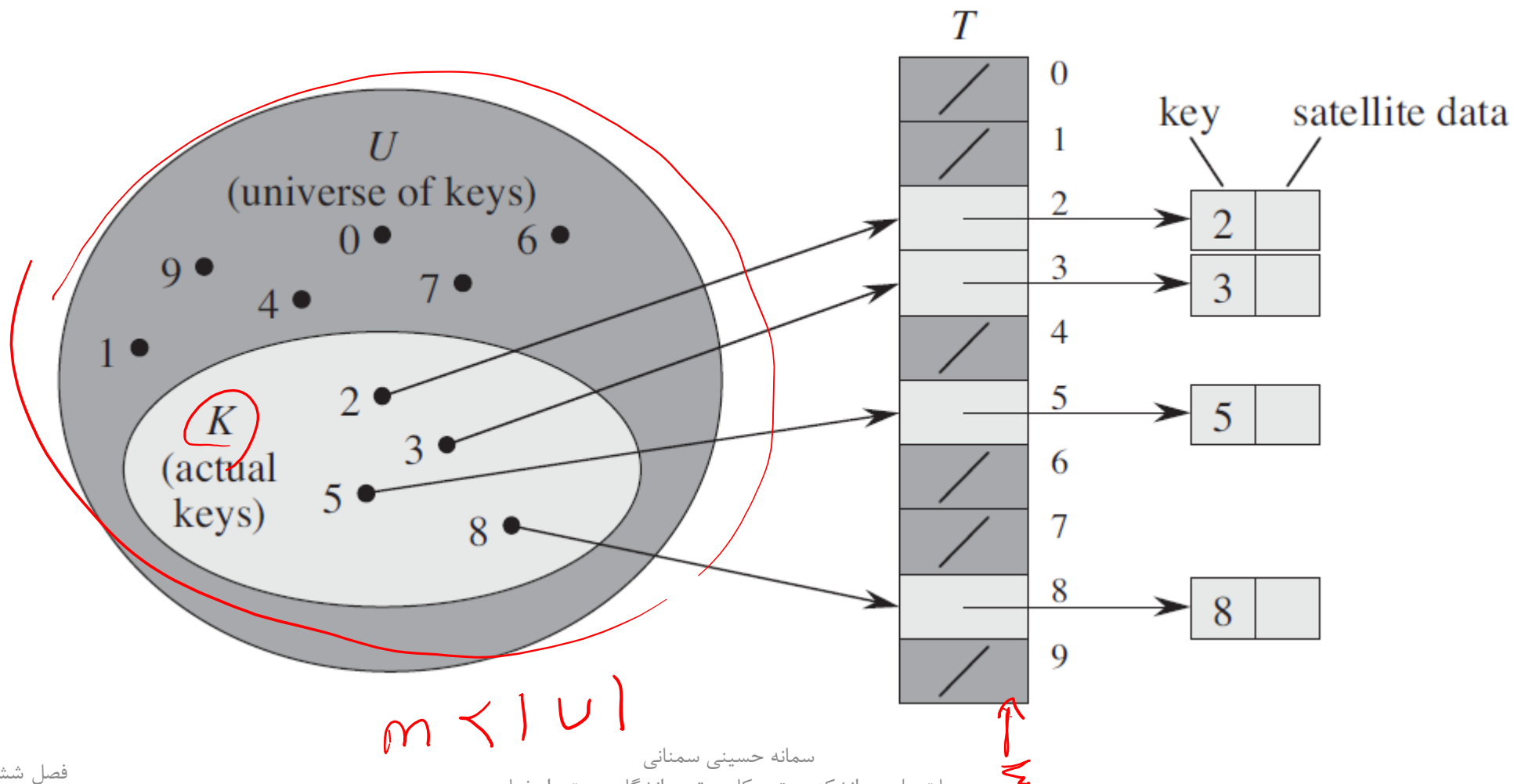


Direct-address tables

- Direct addressing is a simple technique that works well when the universe U of keys is reasonably small.
- Suppose that an application needs a dynamic set in which each element has a key drawn from the universe $U = \{0, 1, \dots, m - 1\}$, where m is not too large.
- We shall assume that no two elements have the same key.
- To represent the dynamic set, we use an array, or direct-address table, denoted $T[0..m - 1]$
- in which each position, or slot, corresponds to a key in the universe U



Direct-address tables





Direct-address tables

DIRECT-ADDRESS-SEARCH(T, k)

1 **return** $T[k]$

DIRECT-ADDRESS-INSERT(T, x)

1 $T[x.key] = x$

DIRECT-ADDRESS-DELETE(T, x)

1 $T[x.key] = \text{NIL}$

Each of these operations takes only $O(1)$ time.



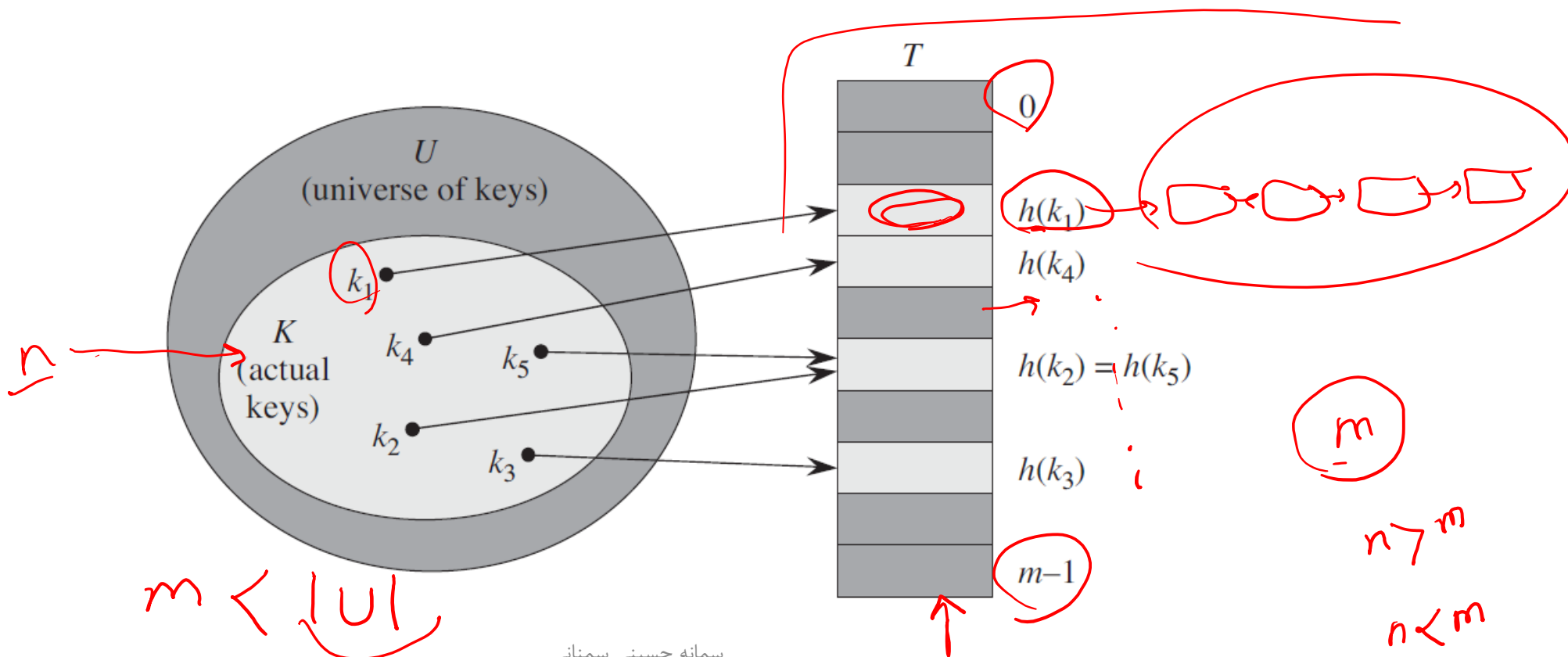
Hash table vs. Direct-address tables

- if the universe U is large, storing a table T of size $|U|$ may be impractical or even impossible, given the memory available on a typical computer.
- Furthermore, the set K of keys actually stored may be so small relative to U that most of the space allocated for T would be wasted.
- When the set K of keys stored in a dictionary is much smaller than the universe U of all possible keys, a hash table requires much less storage than a direct address table.
- we can reduce the storage requirement to $\Theta(|K|)$ while we maintain the benefit that searching for an element in the hash table still requires only $O(1)$ time.
- The catch is that this bound is for the average-case time, whereas for direct addressing it holds for the worst-case time.



Hash tables

$$h : U \rightarrow \{0, 1, \dots, m-1\}$$



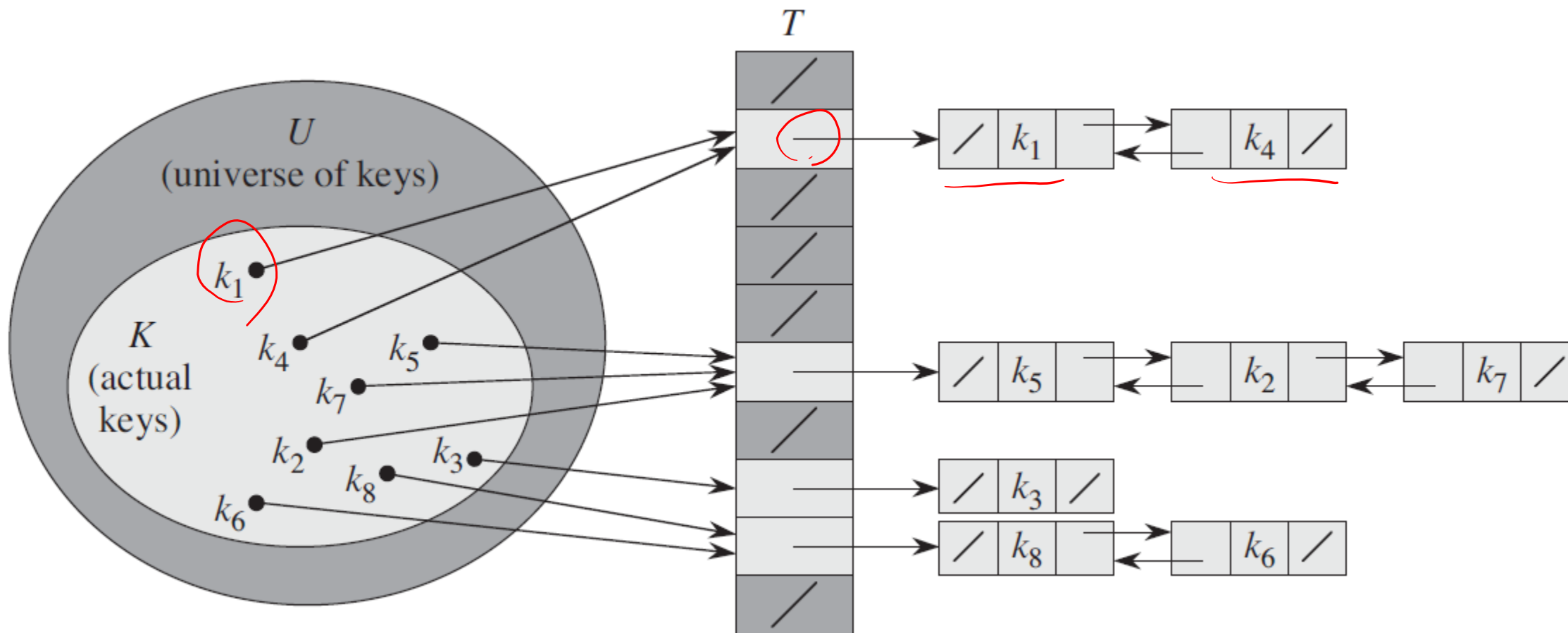


Hash tables

- Of course, the ideal solution would be to avoid collisions altogether.
- We might try to achieve this goal by choosing a suitable hash function h .
- “random”-looking hash function can minimize the number of collisions,
- we still need a method for resolving the collisions that do occur.
- collision resolution technique:
 - chaining.
 - open addressing.



Collision resolution by chaining





Collision resolution by chaining

CHAINED-HASH-INSERT(T, x)

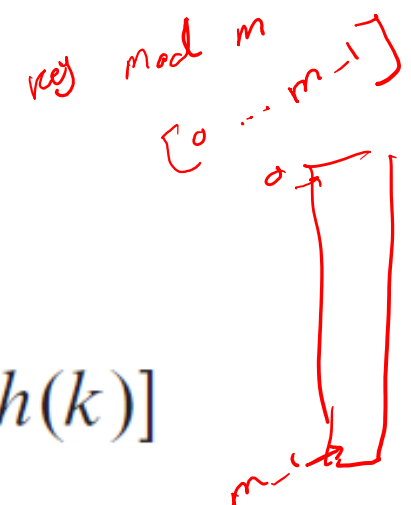
1 insert x at the head of list $T[h(x.key)]$

CHAINED-HASH-SEARCH(T, k)

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

CHAINED-HASH-DELETE(T, x)

1 delete x from the list $T[h(x.key)]$





Analysis of hashing with chaining

- Given a hash table T with m slots that stores n elements
- we define the load factor α for T as n/m
- the average number of elements stored in a chain.
- The worst-case behavior of hashing with chaining is terrible: all n keys hash to the same slot, creating a list of length n .
- The worst-case time for searching is thus $\theta(n)$ plus the time to compute the hash function.
- The average-case performance of hashing depends on how well the hash function h distributes the set of keys to be stored among the m slots, on the average.





Analysis of hashing with chaining

Theorem 11.1

In a hash table in which collisions are resolved by chaining, an unsuccessful search takes average-case time $\Theta(1 + \alpha)$, under the assumption of simple uniform hashing.



Analysis of hashing with chaining

Theorem 11.2

In a hash table in which collisions are resolved by chaining, a successful search takes average-case time $\Theta(1 + \alpha)$, under the assumption of simple uniform hashing.



Hash functions

- design of good hash functions:
 - hashing by division
 - hashing by multiplication
 - universal hashing



What makes a good hash function?

- A good hash function satisfies (approximately) the assumption of simple uniform hashing
- Occasionally we do know the distribution.
- For example: if we know that the keys are random real numbers k independently and uniformly distributed in the range $0 \leq k < 1$, then the hash function:

$$h(k) = \lfloor km \rfloor$$

- satisfies the condition of simple uniform hashing.



What makes a good hash function?

- Qualitative information about the distribution of keys may be useful in this design process.
- For example, consider a compiler's symbol table, in which the keys are character strings representing identifiers in a program.
- Closely related symbols, such as pt and pts, often occur in the same program.
- A good hash function would minimize the chance that such variants hash to the same slot.



The division method

In the *division method* for creating hash functions, we map a key k into one of m slots by taking the remainder of k divided by m . That is, the hash function is

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

For example, if the hash table has size $m = 12$ and the key is $k = 100$, then $h(k) = 4$. Since it requires only a single division operation, hashing by division is quite fast.



The division method

- When using the division method, we usually avoid certain values of m .
- m should not be a power of 2,
- since if $m = 2^p$, then $h(k)$ is just the p lowest-order bits of k .
- we are better off designing the hash function to depend on all the bits of the key.
- A prime not too close to an exact power of 2 is often a good choice for m .



The division method

- e.g. suppose we wish to allocate a hash table, with collisions resolved by chaining, to hold roughly $n = 2000$ character strings, where a character has 8 bits.
- We don't mind examining an average of 3 elements in an unsuccessful search, and so we allocate a hash table of size $m = 701$.
- We could choose $m = 701$ because it is a prime near $2000/3$ but not near any power of 2.
- Treating each key k as an integer, our hash function would be

$$h(k) = k \bmod 701$$



The multiplication method

$$0 < A < 1$$

$$h(k) = \lfloor m (kA \bmod 1) \rfloor$$

“ $kA \bmod 1$ ” means the fractional part of kA , that is, $kA - \lfloor kA \rfloor$

$$A \approx (\sqrt{5} - 1)/2 = 0.6180339887 \dots$$

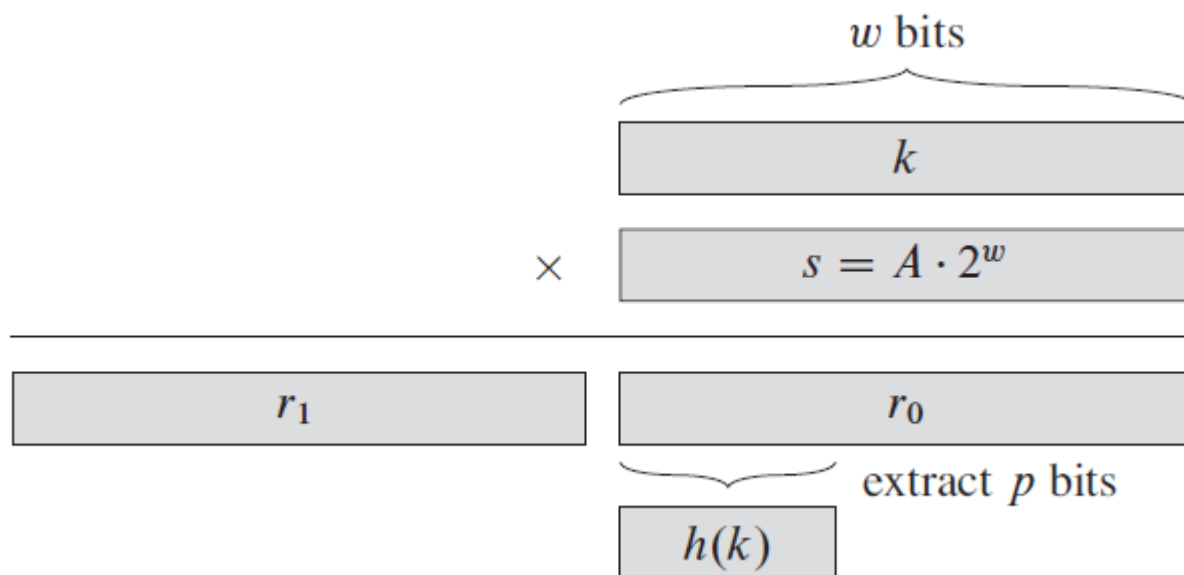
We typically choose it to be a power of 2 ($m = 2^p$ for some integer p).





The multiplication method

$$h(k) = \lfloor m (kA \bmod 1) \rfloor$$





Universal hashing

- If a malicious adversary chooses the keys to be hashed by some fixed hash function,
- then the adversary can choose n keys that all hash to the same slot,
- Average retrieval time of $\theta(n)$.
- Any fixed hash function is vulnerable to such terrible worst-case behavior
- the only effective way to improve the situation is to choose the hash function randomly in a way that is independent of the keys that are actually going to be stored.
- This approach, called universal hashing, can yield provably good performance on average, no matter which keys the adversary chooses.



Universal hashing

- at the beginning of execution we select the hash function at random from a carefully designed class of functions
- Because we randomly select the hash function, the algorithm can behave differently on each execution, even for the same input, guaranteeing good average-case performance for any input.
- compiler's symbol table, we find that the programmer's choice of identifiers cannot now cause consistently poor hashing performance.
- Poor performance occurs only when the compiler chooses a random hash function that causes the set of identifiers to hash poorly,
- but the probability of this situation occurring is small and is the same for any set of identifiers of the same size.



Universal hashing

- Let \mathcal{H} be a finite collection of hash functions that map a given universe U of keys into the range

$$\{0, 1, \dots, m - 1\}$$

$$\mathcal{H} = \{h_1, h_2, h_3, \dots, h_l\}$$



Designing a universal class of hash functions

$$\mathcal{H} = \{h_1, h_2, h_3, \dots, h_l\}$$

- برای هر k و t تعداد توابعی که k و t را به یک خانه map می کنند حداکثر $\frac{|\mathcal{H}|}{m}$
- میشود نشان داد که می توانیم چنین مجموعه تابعی تعریف کنیم.



Designing a universal class of hash functions

- $P =$ a prime number larger than all the numbers in the domain

مثال

- $h_{ab}(k) = ((ak + b) \bmod p) \bmod m$.

- $a \in \{1, \dots, p-1\}$

- $b \in \{0, \dots, p-1\}$

- $\mathcal{H}_{pm} = \{h_{ab} : a \in \mathbb{Z}_p^* \text{ and } b \in \mathbb{Z}_p\}$

متوسط زمان جستجو $O(\alpha)$

Theorem 11.5

The class \mathcal{H}_{pm} of hash functions defined by equations (11.3) and (11.4) is universal.

برای هر k و t تعداد توابعی که k و t را به یک خانه map می کنند حداکثر $\frac{|\mathcal{H}|}{m}$



- self.
- c set or ~~NIL~~
- ide the table.
- an be made.

$$\alpha = \left(\frac{n}{m} \right) \text{ K}$$

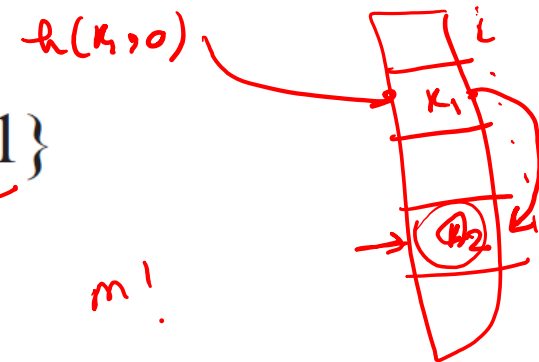


Insertion using open addressing

k_1
 k_2
 k_3

- we successively examine, or probe, the hash table until we find an empty slot in which to put the key.
- the sequence of positions probed depends upon the key being inserted.
- extend the hash function to include the probe number as a second input

$$h : U \times \{0, 1, \dots, m-1\} \rightarrow \{0, 1, \dots, m-1\}$$



probe sequence

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

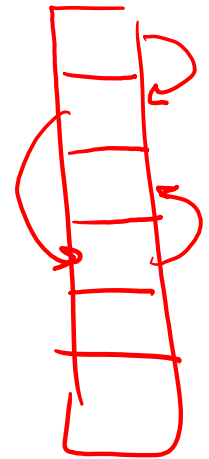


Insertion using open addressing

HASH-INSERT (T, k)

```
1   $i$  = 0
2  repeat
3       $j$  =  $h(k, i)$ 
4      if  $T[j] == \text{NIL}$ 
5           $T[j] = k$ 
6          return  $j$ 
7      else  $i = i + 1$ 
8  until  $i == m$ 
9  error "hash table overflow"
```

j
 i





Search using open addressing

HASH-SEARCH(T, k)

```
1   $i = 0$ 
2  repeat
3       $j = h(k, i)$ 
4      if  $T[j] == k$ 
5          return  $j$ 
6       $i = i + 1$ 
7  until  $T[j] == \text{NIL}$  or  $i == m$ 
8  return  $\text{NIL}$ 
```




Example: Insert and Search

→	30	←
→	20	
→	40	←
→	17	
	18	

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

$$m = 5$$

$$k \in \{30, 20, 40, 17, 18\}$$

40

$$\begin{aligned} h(40, 0) &= 40 + 0 \bmod 5 = 0 \\ h(40, 1) &= 41 \bmod 5 = 1 \\ h(40, 2) &= 42 \bmod 5 = 2 \end{aligned}$$

$$\begin{aligned} h(k, 0) &= k + 0 \bmod m \\ &= k \bmod m \\ &= k \bmod 5 \end{aligned}$$

$$0 = 3$$



Deletion using open addressing

- When we delete a key from slot i , we cannot simply mark that slot as empty by storing **NIL** in it.
- If we did, we might be unable to retrieve any key k during whose insertion we had probed slot i and found it occupied.
- Example: delete 30, search 40 in previous example.



Deletion using open addressing

→	30 Deleted	0
→	20	
	40	
	17	
	18	

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

$$m = 5$$

$$k \in \{30, 20, 40, 17, 18\}$$

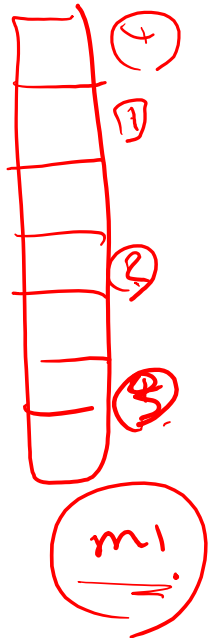
- We can solve this problem by marking the slot, storing in it the special value DELETED instead of NIL.



Uniform hashing

- Uniform hashing: the probe sequence of each key is equally likely to be any of the $m!$ permutations of $\langle 0, 1, \dots, m-1 \rangle$.
- generalizes the notion of simple uniform hashing
- True uniform hashing is difficult to implement, however, and in practice suitable approximations (such as double hashing, defined below) are used.

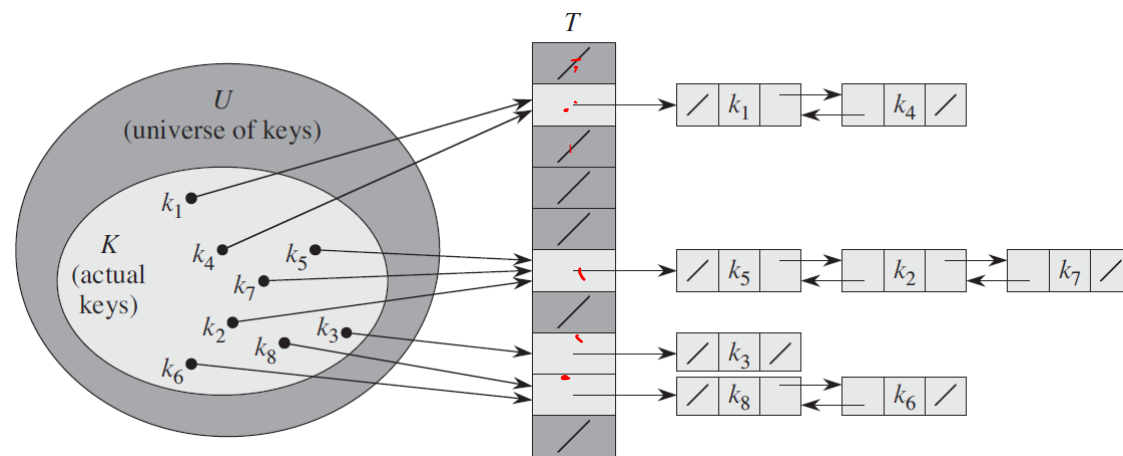
$h(k, i)$ \dots $0 \dots m-1$





Simple uniform hashing

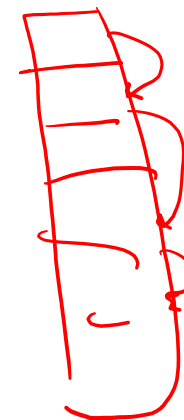
any given element is equally likely to hash into any of the m slots,
independently of where any other element has hashed to.





Techniques to compute the probe sequences

- linear probing
- quadratic probing
- double hashing



All guarantee :

0, 5, 3, 8, ...

0, 1, ..., m-1

$\langle h(k, 0), h(k, 1), \dots, h(k, m-1) \rangle$ is a permutation of $\langle 0, 1, \dots, m-1 \rangle$ for any key k



Linear probing

Given an ordinary hash function $h' : U \rightarrow \{0, 1, \dots, m - 1\}$,

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



Because the initial probe determines the entire probe sequence, there are only m distinct probe sequences.



Linear probing

- Easy to implement
- Problem:

30
20
40
50
60

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

$$m = 5$$

$$k \in \{30, 20, 40, 50, 60\}$$