

# Lecture 13. Hash Tables

Introduction to Algorithms  
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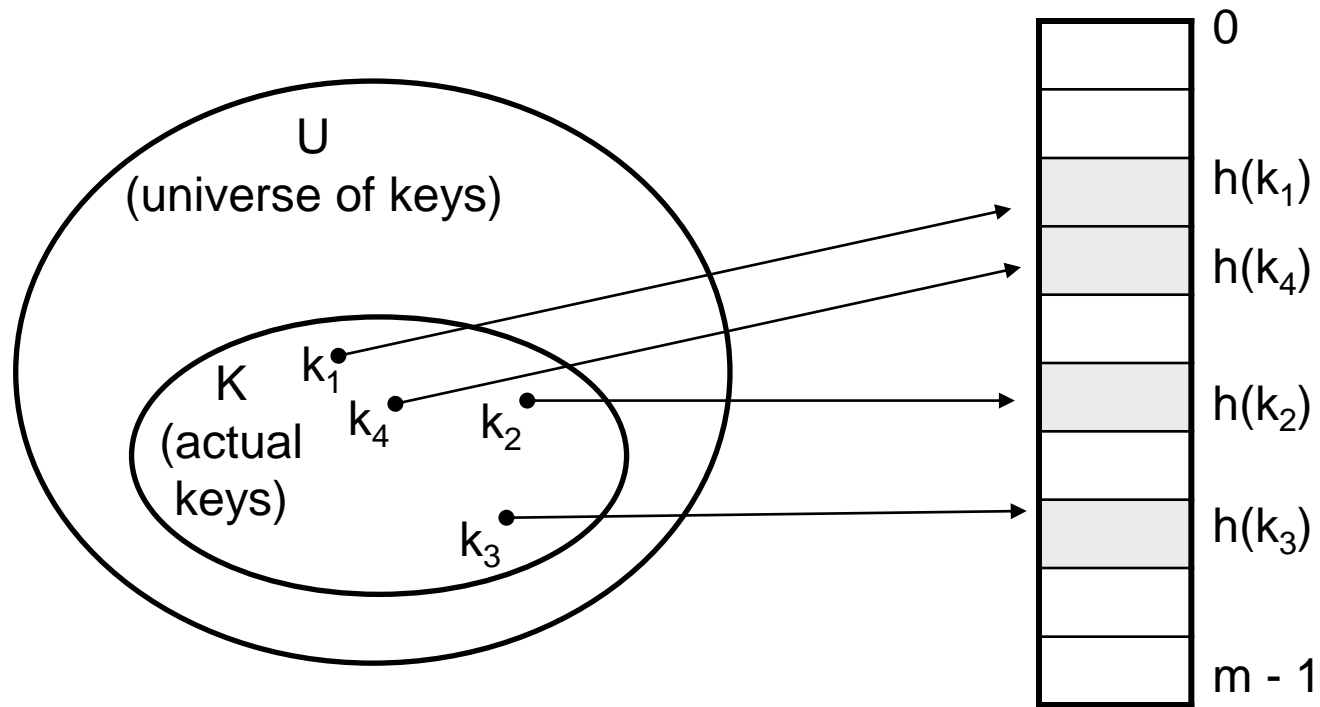
# Hash Tables



## ■ Motivation: symbol tables

- ▶ A compiler uses a *symbol table* to relate symbols to associated data
  - ★ Symbols: variable names, procedure names, etc.
  - ★ Associated data: memory location, call graph, etc.
- ▶ For a symbol table (also called a *dictionary*), we care about search, insertion, and deletion
- ▶ We typically don't care about sorted order

# Hash Tables



# Hash Tables

## ■ 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 correspondingly reduced

# Hash Tables

## ■ More formally:

- ▶ Given a table  $T$  and a record  $x$ , with key (= symbol) and satellite data, we need to support:
  - ★ Insert ( $T, x$ )
  - ★ Delete ( $T, x$ )
  - ★ Search( $T, x$ )
- ▶ We want these to be fast, but don't care about sorting the records

## ■ The structure we will use is a *hash table*

- ▶ Supports all the above in  $O(1)$  expected time !

# Direct-Address Tables

## ■ Suppose:

- ▶ The range of keys is  $0..m-1$
- ▶ Keys are distinct

## ■ The idea:

- ▶ Set up an array  $T[0..m-1]$  in which
  - ★  $T[i] = x$  if  $x \in T$  and  $\text{key}[x] = i$
  - ★  $T[i] = \text{NULL}$  otherwise
- ▶ This is called a *direct-address table*
  - ★ Operations take  $O(1)$  time!
  - ★ *So what's the problem?*

# Direct-Address Tables

- Dictionary operations are trivial and take  $O(1)$  time each:

- ▶ **DIRECT-ADDRESS-SEARCH**( $T, k$ )  
return  $T[k]$

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

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

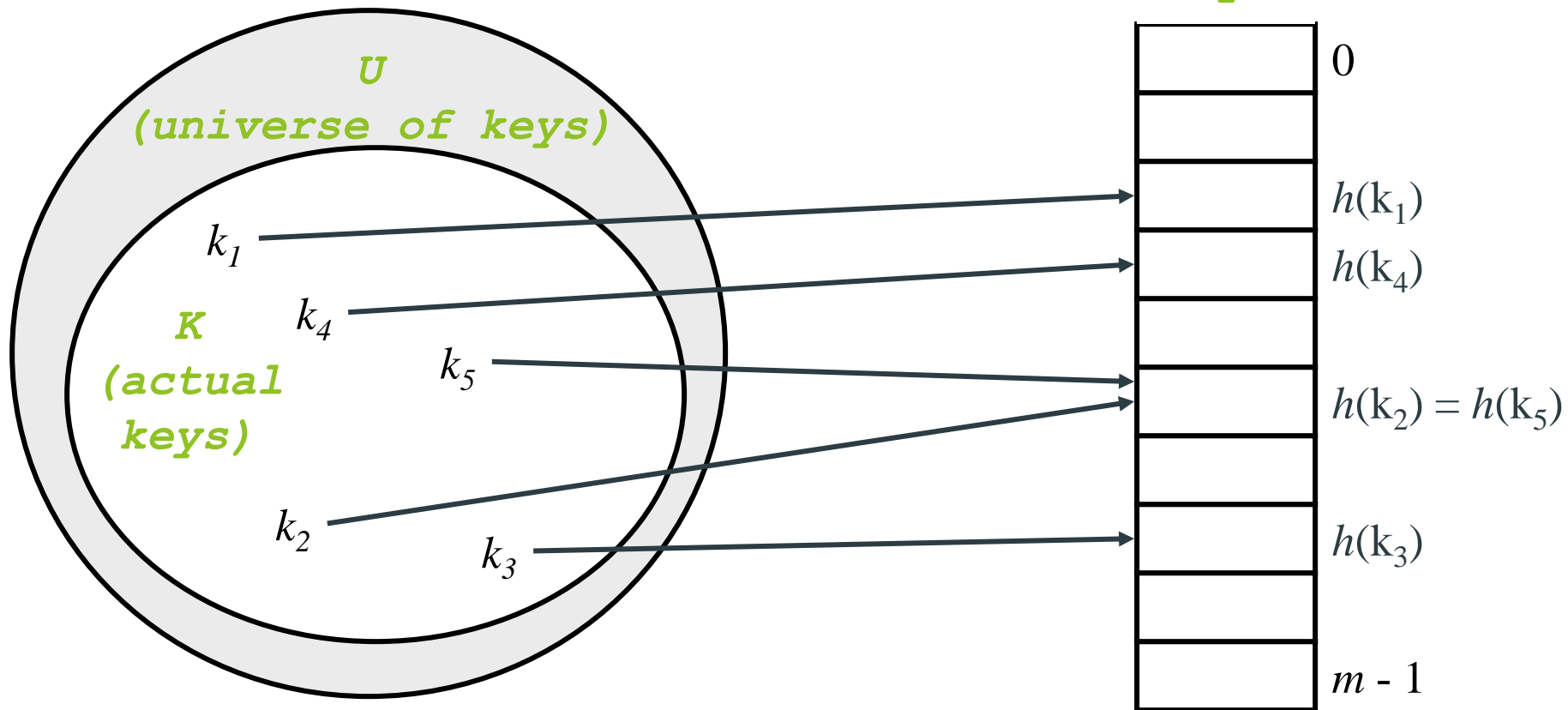
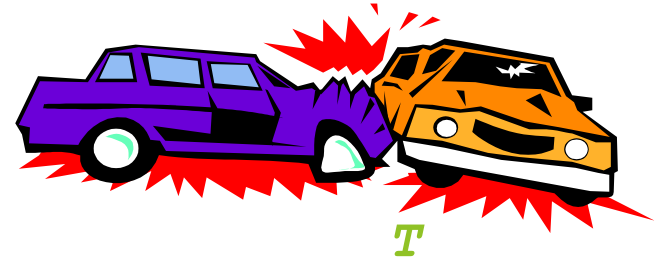
# The Problem With Direct Addressing

- Direct addressing works well when the range  $m$  of keys is relatively small
- But what if the keys are 32-bit integers?
  - ▶ Problem 1: direct-address table will have  $2^{32}$  entries, more than 4 billion
  - ▶ Problem 2: even if memory is not an issue, the time to initialize the elements to NULL may be ...
- Solution: map keys to smaller range  $0..m-1$
- This mapping is called a *hash function*



# Hash Functions

■ Next problem: *collision*



# Resolving Collisions

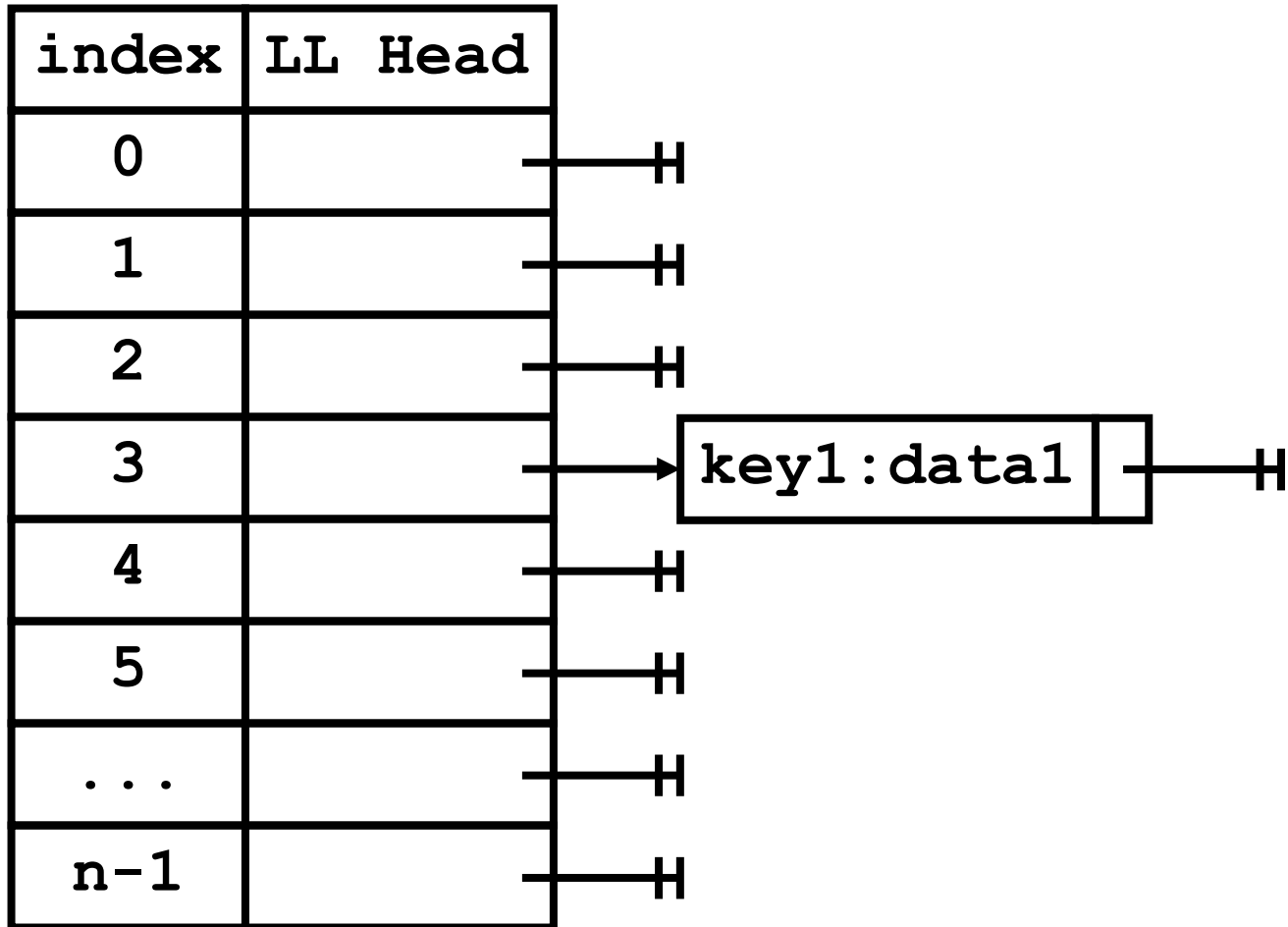
■ *How can we solve the problem of collisions?*

- ▶ Solution 1: *chaining*
- ▶ Solution 2: *open addressing*

# Chaining

index	LL Head	
0		— H
1		— H
2		— H
3		— H
4		— H
5		— H
...		— H
n-1		— H

# Chaining

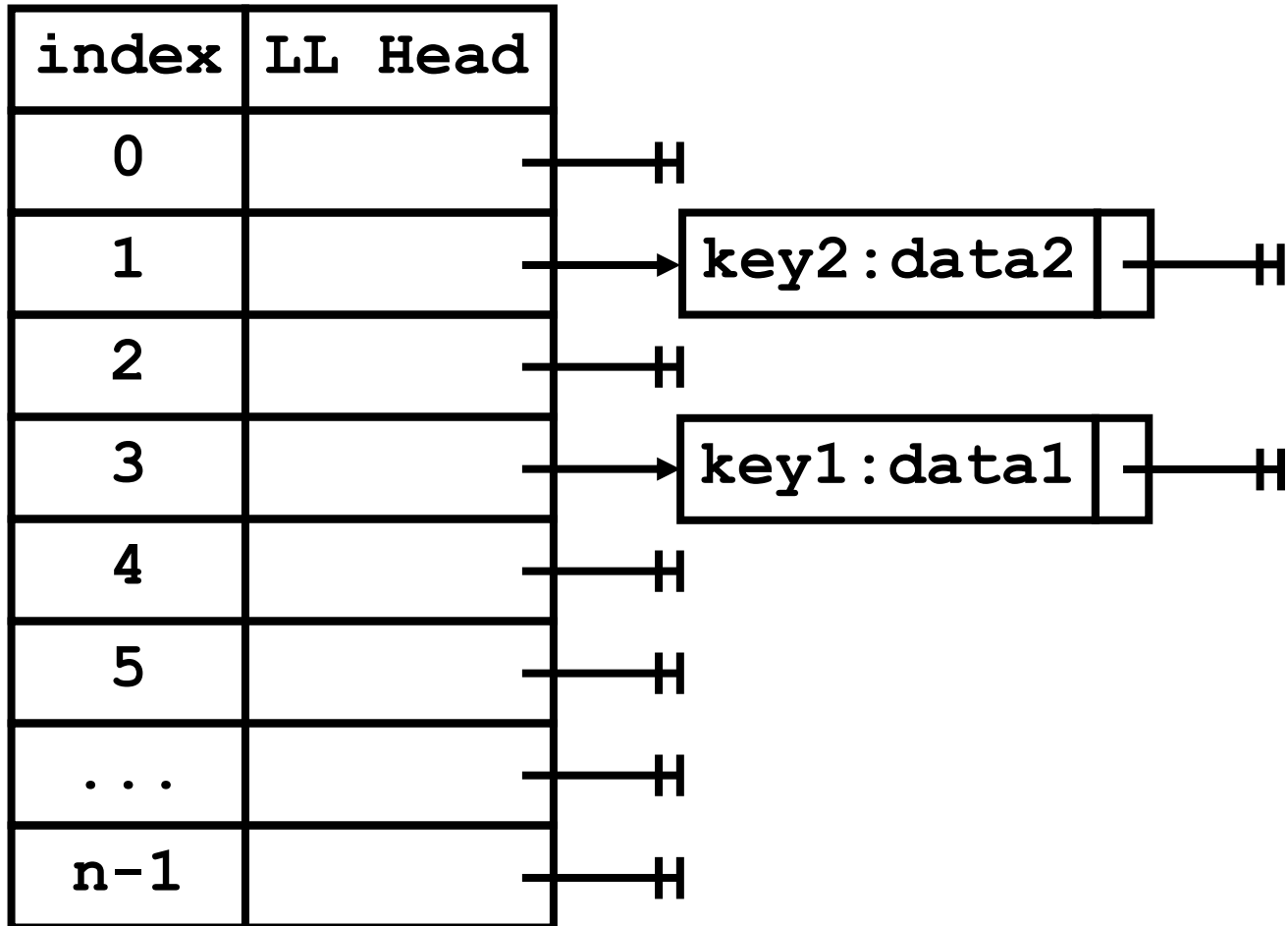


key1: data1



3

# Chaining

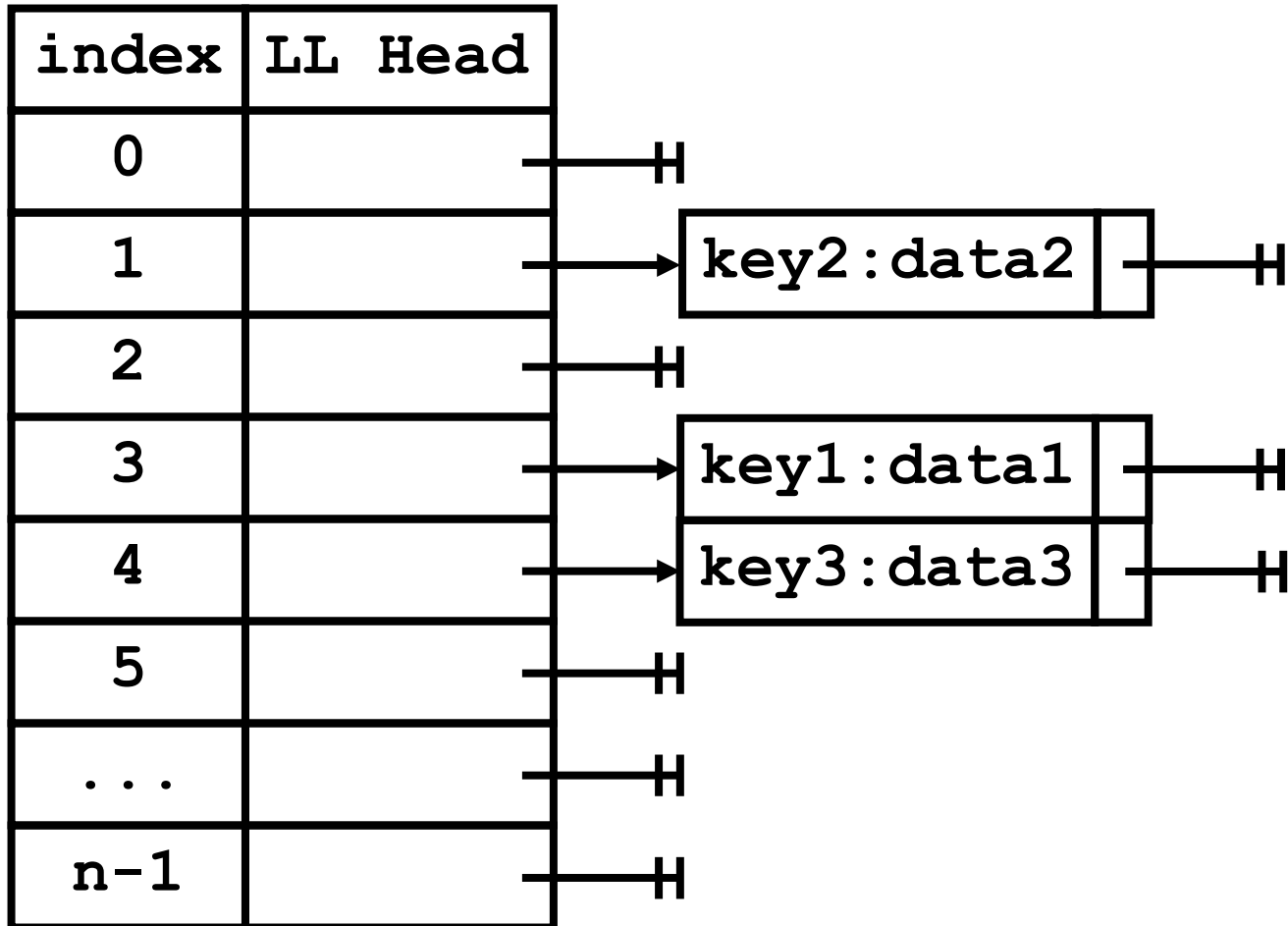


key2: data2



1

# Chaining

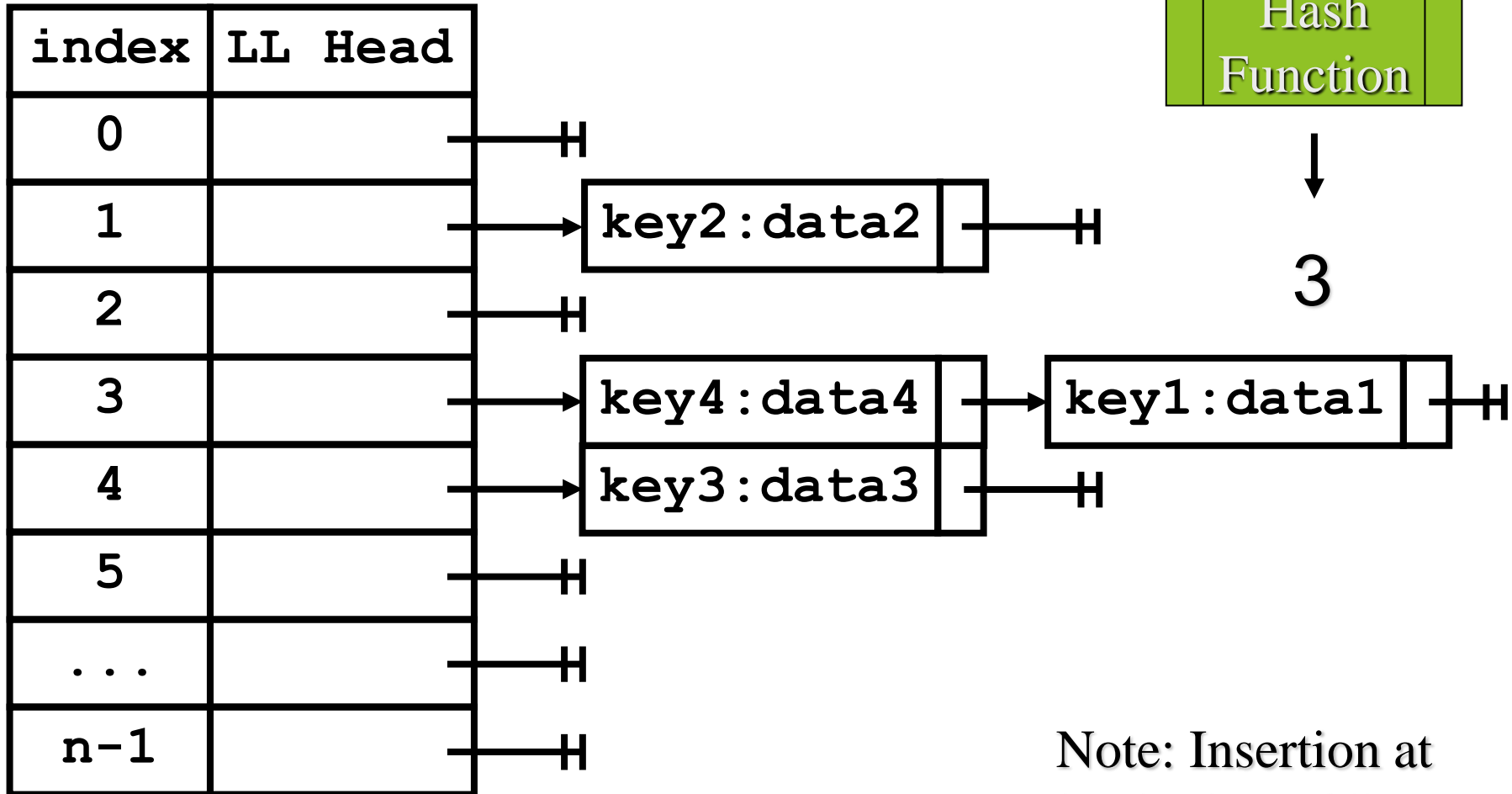


key3: data3



4

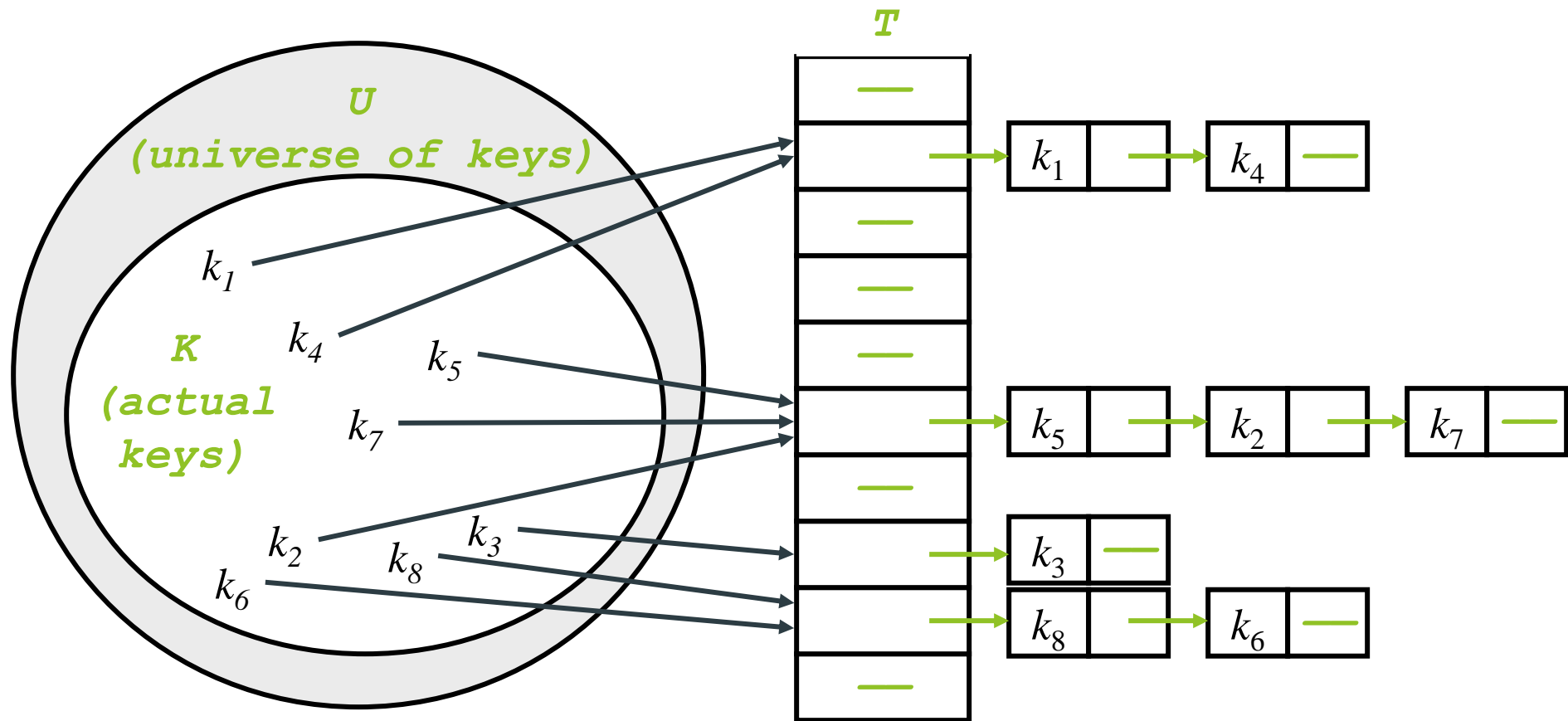
# Chaining



Note: Insertion at  
beginning of list

# Chaining

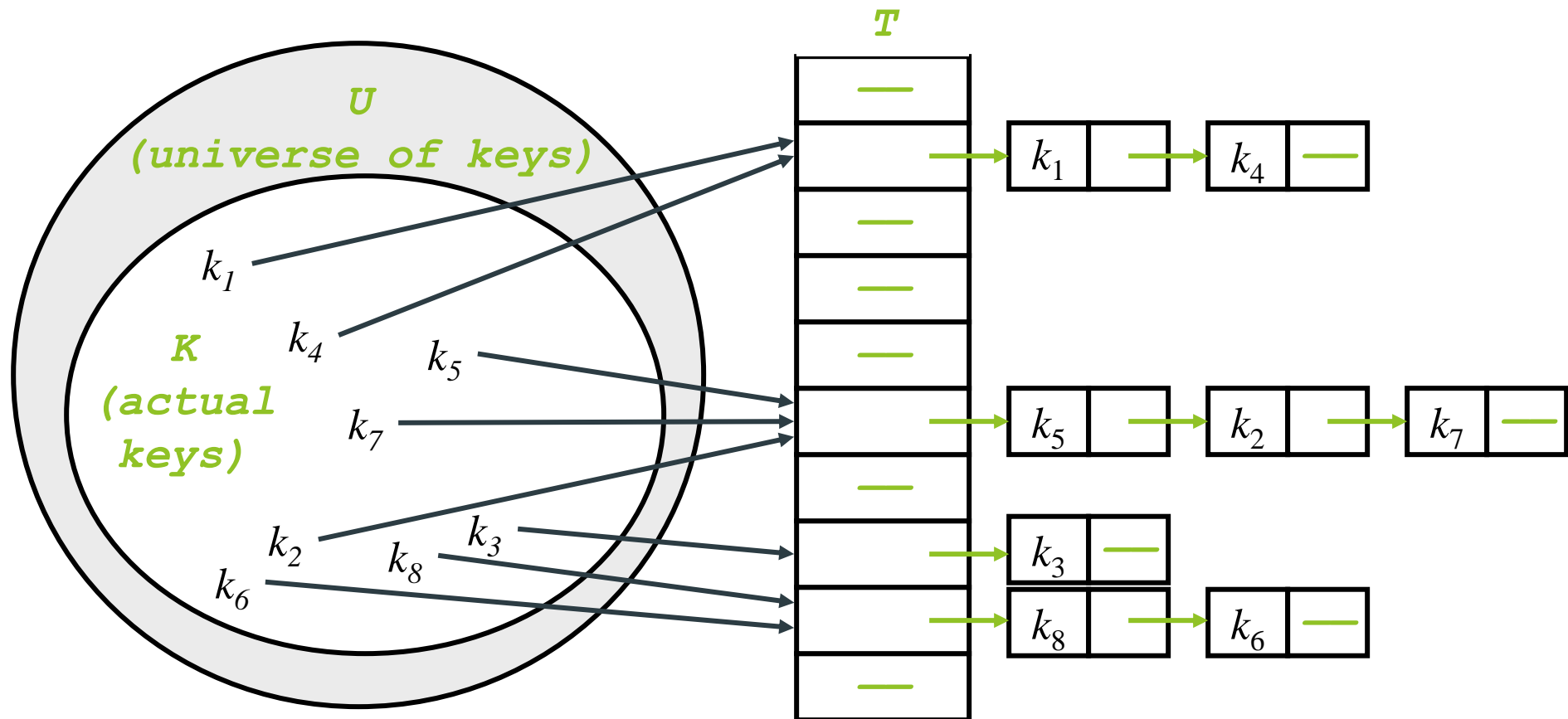
- Chaining puts elements that hash to the same slot in a linked list:





# Chaining

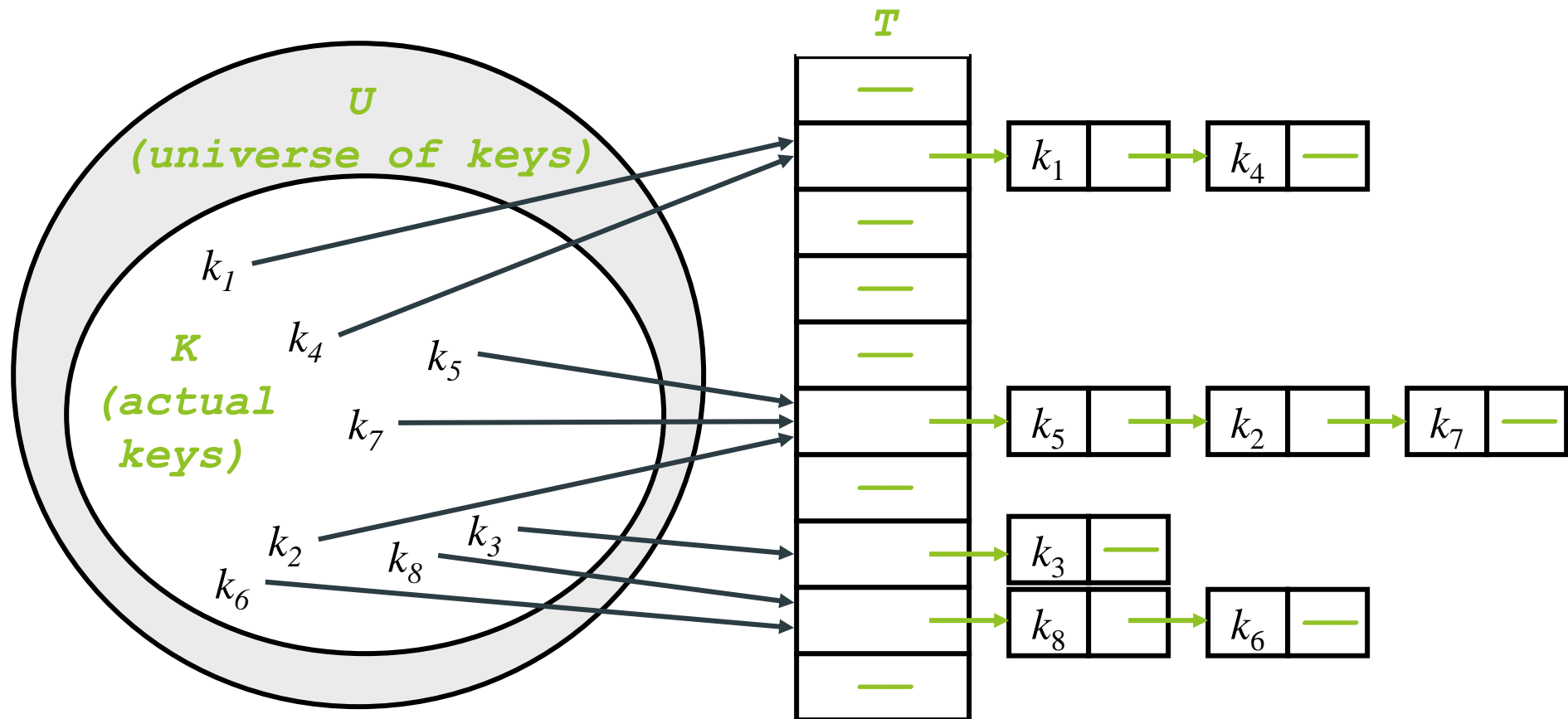
■ *How do we insert an element?*



# Chaining

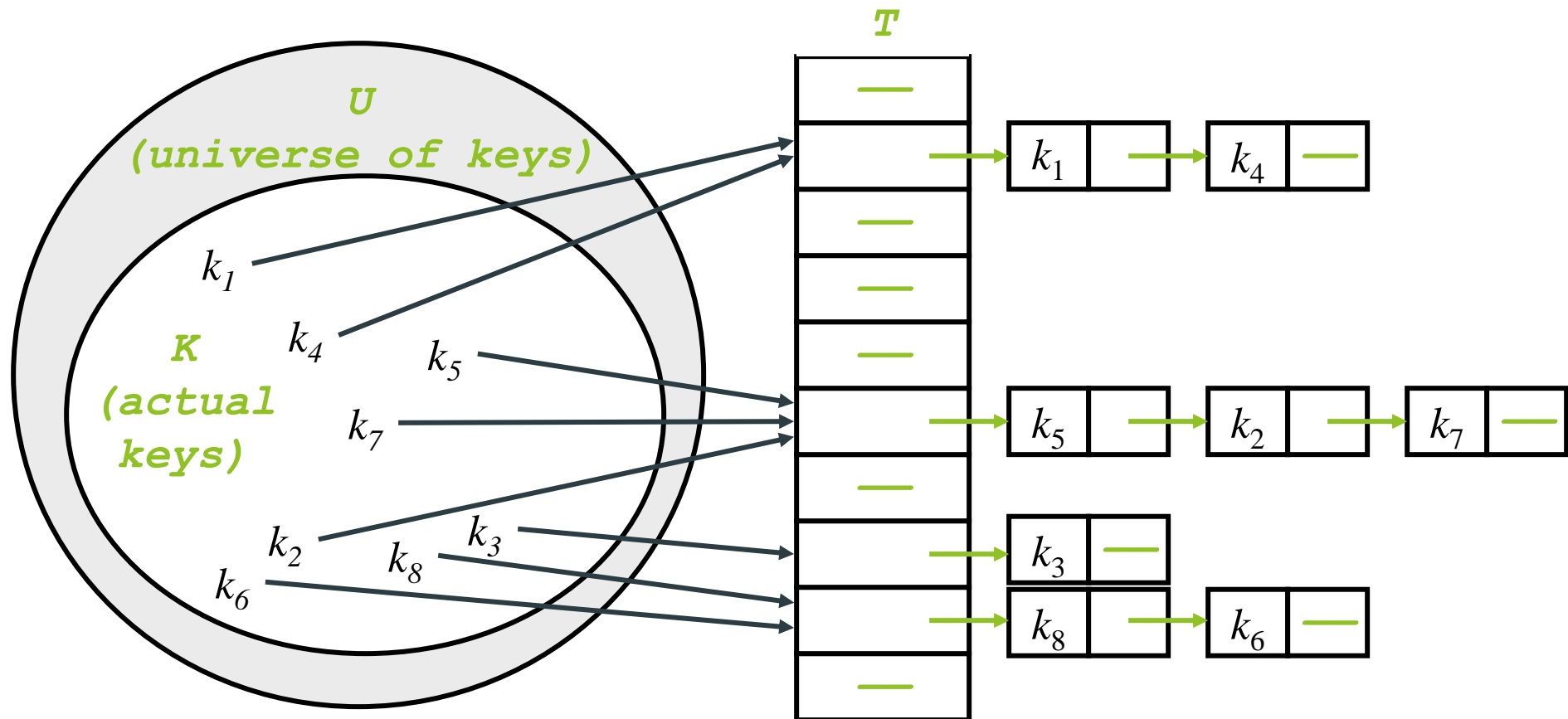
## ■ How do we delete an element?

► Do we need a doubly-linked list for efficient delete?



# Chaining

- How do we search for an element with a given key?



# Chaining

- ▶ CHAINED-HASH-INSERT( $T, x$ )  
insert  $x$  at the head of list  $T[h(\text{key}[x])]$
- ▶ CHAINED-HASH-SEARCH( $T, k$ )  
search for an element with key  $k$  in list  $T[h[k]]$
- ▶ CHAINED-HASH-DELETE( $T, x$ )  
delete  $x$  from the list  $T[h(\text{key}[x])]$

# Practice Problems

- Draw a hash table after we insert the keys 5; 28; 19; 15; 20; 33; 12; 17; 10 with collisions resolved by chaining. Let the table have 9 slots,  $T[0, \dots, 8]$ , and the hash function be  $h(k) = k \bmod 9$ .

i	0	1	2	3	4	5	6	7	8
$T[i]$	[]	[10, 19, 28]	[20]	[12]	[]	[5]	[33, 15]	[]	[17]

# Analysis of Chaining

- Assume *simple uniform hashing*: each key in table is equally likely to be hashed to any slot
- Given  $n$  keys and  $m$  slots in the table, the *load factor*  $\alpha = n / m =$  average # keys per slot
- *What will be the average cost of an unsuccessful search for a key?*
  - ▶  $O(1 + \alpha)$
- *What will be the average cost of a successful search?*
  - ▶  $O(1 + \alpha/2) = O(1 + \alpha)$

# Analysis of Chaining

- So, the cost of searching
  - ▶  $O(1 + \alpha)$
- *If the number of keys  $n$  is proportional to the number of slots in the table, what is  $\alpha$ ?*
  - ▶  $\alpha = O(1)$
  - ▶ In other words, we can make the expected cost of searching constant if we make  $\alpha$  constant

What does this analysis mean? If the number of hash-table slots is at least proportional to the number of elements in the table, we have  $n = O(m)$  and, consequently,  $\alpha = n/m = O(m)/m = O(1)$ .

# Choosing A Hash Function

- Clearly choosing the hash function well is crucial
- *What are desirable features of the hash function?*
  - ▶ Should distribute keys uniformly into slots
  - ▶ Should not depend on patterns in the data

## **Interpreting keys as natural numbers**

Most hash functions assume that the universe of keys is the set  $\mathbb{N} = \{0, 1, 2, \dots\}$  of natural numbers. Thus, if the keys are not natural numbers, we find a way to interpret them as natural numbers. For example, we can interpret a character string as an integer expressed in suitable radix notation. Thus, we might interpret the identifier `pt` as the pair of decimal integers (112, 116), since `p` = 112 and `t` = 116 in the ASCII character set; then, expressed as a radix-128 integer, `pt` becomes  $(112 \cdot 128) + 116 = 14452$ . In the context of a given application, we can usually devise some such method for interpreting each key as a (possibly large) natural number. In what follows, we assume that the keys are natural numbers.



# Hash Functions:

## The Division Method

### ■ 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: power of 2 and non-prime numbers

# The Division Method

- A good choice for  $m$ : a prime number, not too close to an exact power of 2
- e.g., allocate a hash table, with collisions resolved through chaining
  - ▶  $n = 2000$ -character strings (8 bits/character)
  - ▶ Choose  $m$  roughly  $n/3$ :  $m = 701$  (prime near  $2000/3$ , not near a power of 2)
  - ▶  $h(k) = k \bmod 701$

# Hash Functions: The Multiplication Method

## Idea:

- Multiply key  $k$  by a constant  $A$ ,  $0 < A < 1$
- Extract the fractional part of  $kA$
- Multiply the fractional part by  $m$
- Take the floor of the result

$$h(k) = \lfloor m \underbrace{(kA \bmod 1)}_{\text{fractional part of } kA} \rfloor$$

$$\text{fractional part of } kA = kA - \lfloor kA \rfloor$$

- **Disadvantage:** Slower than division method
- **Advantage:** Value of  $m$  is not critical: typically  $2^p$

# The Multiplication Method

- For a constant  $A$ ,  $0 < A < 1$ :

- $h(k) = \lfloor m (kA - \lfloor kA \rfloor) \rfloor$



*Fractional part of  $kA$*

- Choose  $m = 2^P$

- Choose  $A$  not too close to 0 or 1

- Knuth: Good choice for  $A = (\sqrt{5} - 1)/2$

# Hash Functions: Universal Hashing

- As before, when attempting to foil a malicious adversary: randomize the algorithm
- *Universal hashing:*
  - ▶ Guarantees good performance on average, no matter what keys adversary chooses
  - ▶ Suppose we want the hash function to uniformly distribute hash values over the hash table of size  $m$
  - ▶ Given  $h(x)$ , we want  $\text{Prob}\{h(x)=h(y)\}=1/m$
  - ▶ The # of functions  $|f|$  in  $H$  s.t.  $h(x)=h(y)$  for any  $x, y$  in  $U$
  - ▶  $|f| / |H| = 1/m$  equivalent to  $|f| = |H| / m$

Let  $\mathcal{H}$  be a finite collection of hash functions that map a given universe  $U$  of keys into the range  $\{0, 1, \dots, m-1\}$ . Such a collection is said to be **universal** if for each pair of distinct keys  $k, l \in U$ , the number of hash functions  $h \in \mathcal{H}$  for which  $h(k) = h(l)$  is at most  $|\mathcal{H}|/m$ . In other words, with a hash function randomly chosen from  $\mathcal{H}$ , the chance of a collision between distinct keys  $k$  and  $l$  is no more than the chance  $1/m$  of a collision if  $h(k)$  and  $h(l)$  were randomly and independently chosen from the set  $\{0, 1, \dots, m-1\}$ .

# A Universal Hash Function

- Choose table size  $m$  to be prime
- Decompose key  $x$  into  $r+1$  bytes, so that  $x = \{x_0, x_1, \dots, x_r\}$ 
  - ▶ Only requirement is that max value of byte  $< m$
  - ▶ Let  $a = \{a_0, a_1, \dots, a_r\}$  denote a sequence of  $r+1$  elements chosen randomly from  $\{0, 1, \dots, m - 1\}$
  - ▶ Define corresponding hash function  $h_a$

$$\star h_a(x) = \sum_{i=0}^r a_i x_i \bmod m$$

## e.g. Universal Hash Functions

We now define the hash function  $h_{ab}$  for any  $a \in \mathbb{Z}_p^*$  and any  $b \in \mathbb{Z}_p$  using a linear transformation followed by reductions modulo  $p$  and then modulo  $m$ :

$$h_{ab}(k) = ((ak + b) \bmod p) \bmod m . \quad (11.3)$$

For example, with  $p = 17$  and  $m = 6$ , we have  $h_{3,4}(8) = 5$ . The family of all such hash functions is

$$\mathcal{H}_{pm} = \{h_{ab} : a \in \mathbb{Z}_p^* \text{ and } b \in \mathbb{Z}_p\} . \quad (11.4)$$

Each hash function  $h_{ab}$  maps  $\mathbb{Z}_p$  to  $\mathbb{Z}_m$ . This class of hash functions has the nice property that the size  $m$  of the output range is arbitrary—not necessarily prime—a feature which we shall use in Section 11.5. Since we have  $p - 1$  choices for  $a$  and  $p$  choices for  $b$ , the collection  $\mathcal{H}_{pm}$  contains  $p(p - 1)$  hash functions.

## e.g. Universal Hash Functions

$$p = 17, m = 6$$

$$h_{a,b}(k) = ((ak + b) \bmod p) \bmod m$$

$$h_{3,4}(8) = ((3 \cdot 8 + 4) \bmod 17) \bmod 6$$

$$= (28 \bmod 17) \bmod 6$$

$$= 11 \bmod 6$$

$$= 5$$



# Open Addressing

- If we have enough contiguous memory to store all the keys ( $m > N$ )
  - ⇒ store the keys in the table itself
- No need to use the linked lists anymore
- Collision resolution
  - ▶ Put the elements that collide in the available empty places in the table

# Open Addressing

## ■ Basic idea:

- ▶ To insert: if slot is full, try another slot, ..., until an open slot is found (*probing*)
- ▶ To search, follow same sequence of probes as would be used when inserting the element
  - ★ If reach element with correct key, return it
  - ★ If reach a NULL pointer, element is not in table

## ■ Good for fixed sets (adding but no deletion)

- ▶ Example: spell checking

## ■ Table needn't be much bigger than $n$

# Open Addressing

index	data
0	
1	
2	
3	
4	
5	
...	
n-1	

# Open Addressing

index	data
0	
1	
2	
3	key1: data1
4	
5	
...	
n-1	

key1: data1



3

# Open Addressing

index	data
0	
1	key2: data2
2	
3	key1: data1
4	
5	
...	
n-1	

key2: data2



1

# Open Addressing

index	data
0	
1	key2 : data2
2	
3	key1 : data1
4	key3 : data3
5	
...	
n-1	

key3: data3



4

# Open Addressing

index	data
0	
1	key2 : data2
2	
3	key1 : data1
4	key3 : data3
5	
...	
n-1	

key4: data4



3



# Open Addressing

index	data
0	
1	key2 : data2
2	
3	key1 : data1
4	key3 : data3
5	key4 : data4
...	
n-1	

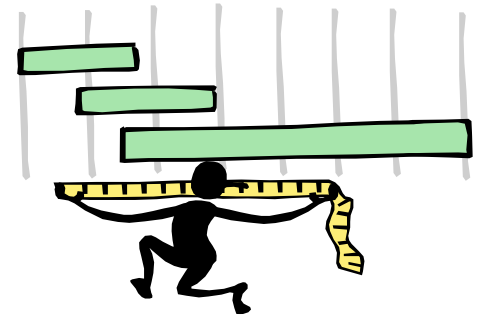
key4: data4



3



# 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 NIL
```

**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”
```

# Linear Probing

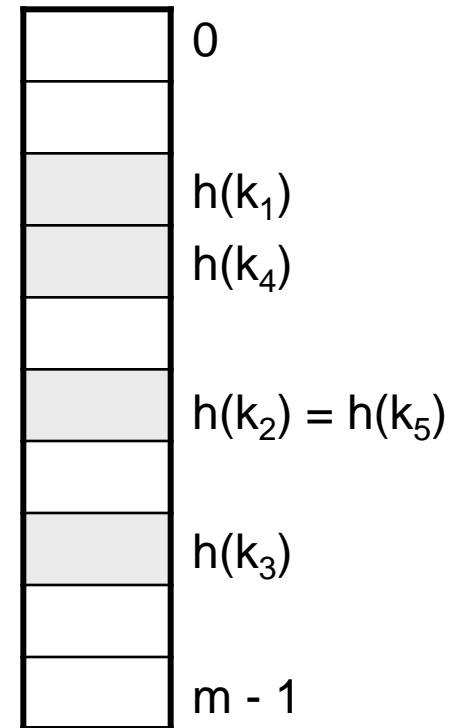
- When there is a collision, check the next available position in the table (probing)

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

- First slot probed:  $h_1(k)$ , second:  $h_1(k) + 1$  and so on

# Linear Probing

- **Searching** for a key
- Three situations:
  - ▶ Position in table is occupied with an element of equal key
  - ▶ Position in table is empty
  - ▶ Position in table occupied with a different element
- Probe the next higher index until the element is found or an empty position is found
- The process wraps around to the beginning of the table



# Linear Probing

Assume  $\text{hash}(x) = \text{hash}(y) = \text{hash}(z) = i$ . And assume  $x$  was inserted first, then  $y$  and then  $z$ .

In open addressing:  $\text{table}[i] = x$ ,  $\text{table}[i+1] = y$ ,  $\text{table}[i+2] = z$ .

Now, assume you want to delete  $x$ , and set it back to `NULL`.

When later you will search for  $z$ , you will find that  $\text{hash}(z) = i$  and  $\text{table}[i] = \text{NULL}$ , and you will return a wrong answer:  $z$  is not in the table.

To overcome this, you need to set  $\text{table}[i]$  with a special marker indicating to the search function to keep looking at index  $i+1$ , because there might be element there which its hash is also  $i$ .

## ■ Deleting a key

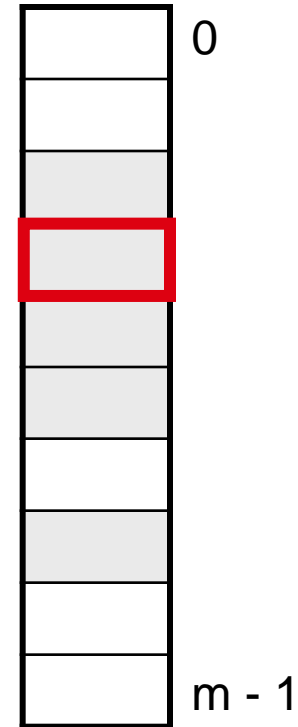
- ▶ Cannot mark the slot as empty
- ▶ Impossible to retrieve keys inserted after that slot was occupied

## ■ 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



# Practice Problems

- Consider a hash table of length 11 using open addressing with the primary hash function  $h_1(k) = k$ . Illustrate the result of inserting 31, 4, 15, 28, 59 using linear probing.

- ▶ Hash function  $h(k,i) = (h_1(k) + i) \bmod m = (k+i) \bmod m$

- ★  $h(31,0) = 9 \rightarrow T[9] = 31$

- ★  $h(4,0) = 4 \rightarrow T[4] = 4$

- ★  $h(15,0) = 4$  collision

- $h(15,1) = 5 \rightarrow T[5] = 15$

- ★  $h(28,0) = 6 \rightarrow T[6] = 28$

- ★  $h(59,0) = 4$  collision

- $h(59,1) = 5$  collision

- $h(59,2) = 6$  collision

- $h(59,3) = 7 \rightarrow T[7] = 59$

# Double Hashing

- Use a first hash function to determine the first slot
- Use a second hash function to determine the increment for the probe sequence

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

- Initial probe:  $h_1(k)$ , second probe is offset by  $h_2(k) \bmod m$ , so on
- **Advantage:** avoids clustering
- **Disadvantage:** harder to delete an element

# Double Hashing

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

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

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

■ Insert key 14:

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

$$h_2(14) = 1 + (14 \bmod 11) = 4$$

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

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

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

# Double Hashing

Choosing the second hash function

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

- $h_2$  should not evaluate to 0  $\Rightarrow$  infinite loop
- $h_2$  should be relatively prime to the table size  $m$ 
  - ▶  $m = 2 h_2$ : only few slots will be visited
- Solution 1
  - ▶ Choose  $m$  prime
  - ▶ Design  $h_2$  such that it produces an integer less than  $m$
- Solution 2
  - ▶ Choose  $m = 2^p$
  - ▶ Design  $h_2$  such that it always produces an odd number



# Practice Problems

- Consider a hash table of length 11 using open addressing. Illustrate the result of inserting 31, 4, 15, 28, 59 using double hashing with functions  $h_1(k) = k$  and  $h_2(k) = 1 + (k \bmod (m-1))$ .

► Hash function  $h(k,i) = (k + i(1 + (k \bmod (m-1)))) \bmod m$

★  $h(31,0) = 9 \rightarrow T[9] = 31$

★  $h(4,0) = 4 \rightarrow T[4] = 4$

★  $h(15,0) = 4$  collision

○  $h(15,1) = 10 \rightarrow T[10] = 15$

★  $h(28,0) = 6 \rightarrow T[6] = 28$

★  $h(59,0) = 4$  collision

○  $h(59,1) = 3 \rightarrow T[3] = 59$

# Thanks to contributors

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