

# Algorithms

Lecture11

# Hash Tables

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

# Introduction

- Many applications require a dynamic set that supports only the **dictionary operations** INSERT, SEARCH, and DELETE.
- A hash table is effective for implementing a dictionary.
- The expected time to search for an element in a hash table is  $O(1)$ , under some reasonable assumptions.
- Worst-case search time is  $O(n)$ , however.
- A hash table is a generalization of an ordinary array.
- With an ordinary array, we store the element whose key is  $k$  in position  $k$  of the array.
- Given a key  $k$ , we find the element whose key is  $k$  by just looking in the  $k$ th position of the array. This is called **direct addressing**.

# Introduction

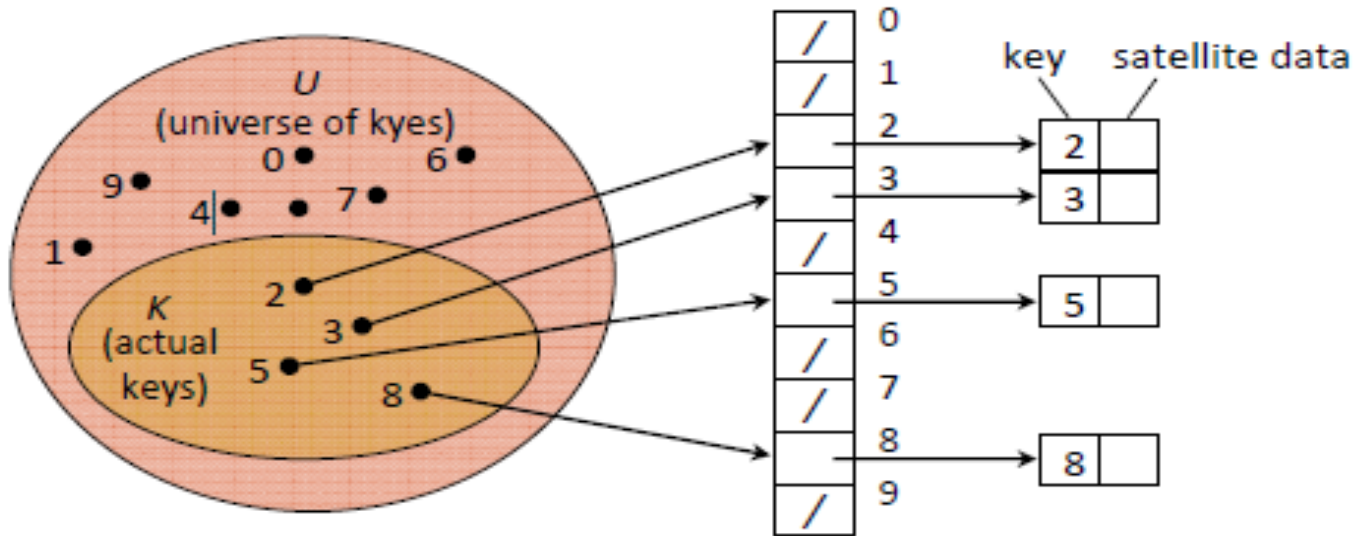
- We use **a hash table** when we do not want to (or can't) allocate an array with one position per possible key.
- Use a hash table when the number of keys actually stored is small relative to the number of possible keys.
- A hash table is an array, but it typically uses a size proportional to the number of keys to be stored (rather than the number of possible keys).
- Given a key  $k$ , don't just use  $k$  as the index into the array. Instead, compute a function of  $k$ , and use that value to index into the array. We call this function a **hash function**.

# Introduction

- Issues that we'll explore in hash tables:
- How to compute hash functions?
  - The multiplication methods.
  - The division methods.
- What to do when the hash function maps multiple keys to the same table entry? ( collision)
  - Chaining.
  - Open addressing.

# Direct-address tables

- Maintain a dynamic set.
- Each element has a key drawn from a universe  $U = \{0, 1, \dots, m - 1\}$  where  $m$  isn't too large.
- No two elements have the same key.
- Represent by **direct-address table**, or array,  $T[0..m-1]$ :
  - Each **slot**, or position, corresponds to a key in  $U$ .
  - If there is an element  $x$  with key  $k$ , then  $T[k]$  contains a pointer to  $x$ .
  - Otherwise,  $T[k]$  is empty, represented by NIL.



- 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*( $T, x$ )

$T[\text{key}[x]] \leftarrow \text{NIL}$

## Problem:

- If the universe  $U$  is large, storing a table of size  $|U|$  may be impractical or impossible.
- The set  $K$  of keys actually stored is small, compared to  $U$ , so that most of the space allocated for  $T$  is wasted.

## Solution: Hash tables

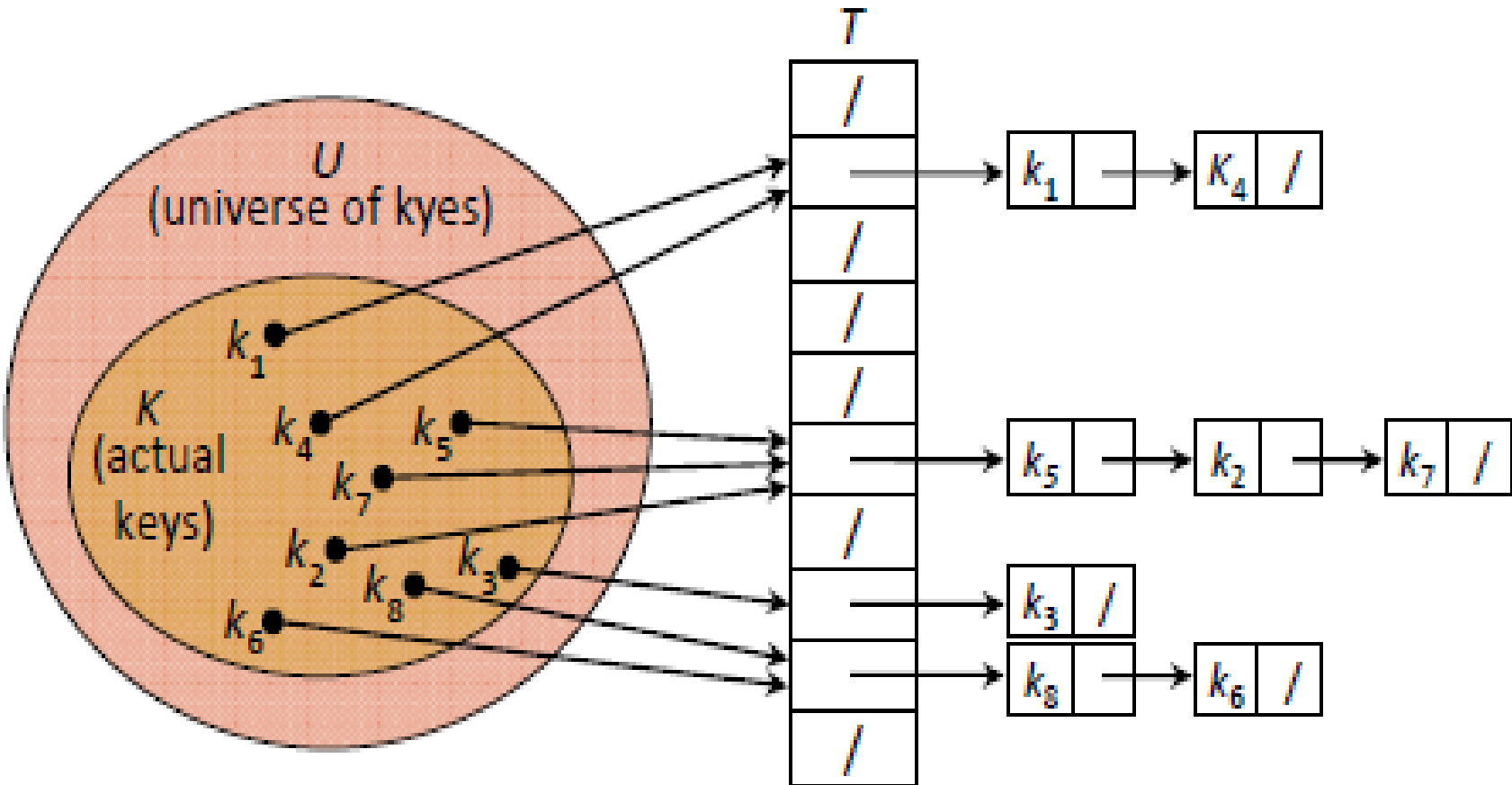
- When  $K$  is much smaller than  $U$ , a hash table requires much less space than a direct-address table.
- Storage requirements can be reduced to  $^{\circ}(|K|)$ .
- Searching for an element requires  $O(1)$  time, but in the **average case**, not the **worst case**.



# Hash Tables

- **Idea:** Instead of storing an element with key  $k$  in slot  $k$ , use a function  $h$  and store the element in slot  $h(k)$ .
- We call  $h$  a **hash function**.
- $h : U \rightarrow \{0, 1, \dots, m - 1\}$ , so that  $h(k)$  is a legal slot number in  $T$ .
- We say that  $k$  **hashes** to slot  $h(k)$ .
- We also say that  $h(k)$  is the **hash value** of key  $k$ .

# Hash Tables



**Collisions:** When two or more keys hash to the same slot.

- Can happen when there are more possible keys than slots ( $|U| > m$ ).

Methods to resolve the collision problem.

- **Chaining**
- **Open addressing**
- Chaining is usually better than open addressing.

# Collision resolution by chaining

- 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 stored elements that hash to  $j$ .
- If there are no such elements, slot  $j$  contains NIL.

# Dictionary Operations

How to implement dictionary operations with chaining:

*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.

*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  $h(k)$ .

# Dictionary Operations....

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

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

- Given pointer  $x$  to the element to delete, so no search is needed to find this element.
- Worst-case running time is  $O(1)$  time if the lists are **doubly linked**.
- If the lists are **singly linked**, then deletion takes as long as searching, because we must find  $x$ 's predecessor in its list.

# Analysis of hashing with chaining

Given a key, how long does it take to find an element with that key?

Analysis is in terms of the **load factor**  $\alpha = n / m$ :

- $n = \#$  of elements in the table.
- $m = \#$  of slots in the table =  $\#$  of (possibly empty) linked lists.
- Load factor is average number of elements per linked list.
- Can have  $\alpha < 1$ ,  $\alpha = 1$ , or  $\alpha > 1$ .
- **Worst case** is when all  $n$  keys hash to the same slot
- get a single list of length  $n$
- worst-case time to search is  $\Theta(n)$ , plus time to compute hash function.
- **Average case** depends on how well the hash function distributes the keys among the slots.`

# Average-case performance

- Assume **simple uniform hashing**: any given element is equally
- likely to hash into any of the  $m$  slots.
- For  $j = 0, 1, \dots, m-1$ , denote the length of the list  $T[j]$  by  $n_j$ , so
- that  $n = n_0 + n_1 + \dots + n_{m-1}$ .
- Average value of  $n_j$  is  $E[n_j] = \frac{n}{m} = n/m$ .
- Assume that the hash value  $h(k)$  can be computed in  $O(1)$  time.
- Time for the element with key  $k$  depends on the length  $n_{h(k)}$
- of the list  $T[h(k)]$ .
- We consider two cases:
  - contains no element with key  $k \rightarrow$  unsuccessful.
  - contain an element with key  $k \rightarrow$  successful.



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- Assume that the hash value  $h(k)$  can be computed in  $O(1)$  time.
  
- Time for the element with key  $k$  depends on the length  $n_{h(k)}$  of the list  $T[h(k)]$ .
  
- We consider two cases:
  - contains no element with key  $k \rightarrow$  unsuccessful.
  - contain an element with key  $k \rightarrow$  successful.

# Theorem 11.1

- An **unsuccessful search** takes expected time<sup>o</sup>  $(1 + \alpha)$ .

## Proof:

- Under the assumption of simple uniform hashing, any key not already in the table is equally likely to hash to any of the  $m$  slots.
- To search unsuccessfully for any key  $k$ , need to search to the end of the list  $T[h(k)]$ .
- This list has expected length  $E[nh(k)] = \alpha$ .
- Therefore, the expected number of elements examined in an unsuccessful search is .
- Adding in the time to compute the hash function.
- The total time required is<sup>o</sup>  $(1 + \alpha)$ .

# Theorem 11.2

- An **successful search** takes expected time  $\Theta(1 + \alpha)$ .

Proof:

- Assume the element being searched for is equally likely to be any of the  $n$  elements in the table  $T$ .
- During a successful search for  $x$ , the # of elements examined = # of elements in the list before  $x + 1$ .
- The expected length of that list is  $(n - i)/m$ .
- The expected # of elements examined in a successful search is

$$\frac{1}{n} \sum_{i=1}^n \left( 1 + \frac{n-i}{m} \right) = 1 + \frac{1}{nm} \sum_{i=1}^n (n-i) = 1 + \frac{1}{nm} \left( \frac{n(n-1)}{2} \right) = 1 + \frac{\alpha}{2} - \frac{\alpha}{2n}.$$

- The total time is  $\Theta(2 + \alpha/2 - \alpha/2n) = \Theta(1 + \alpha)$ .