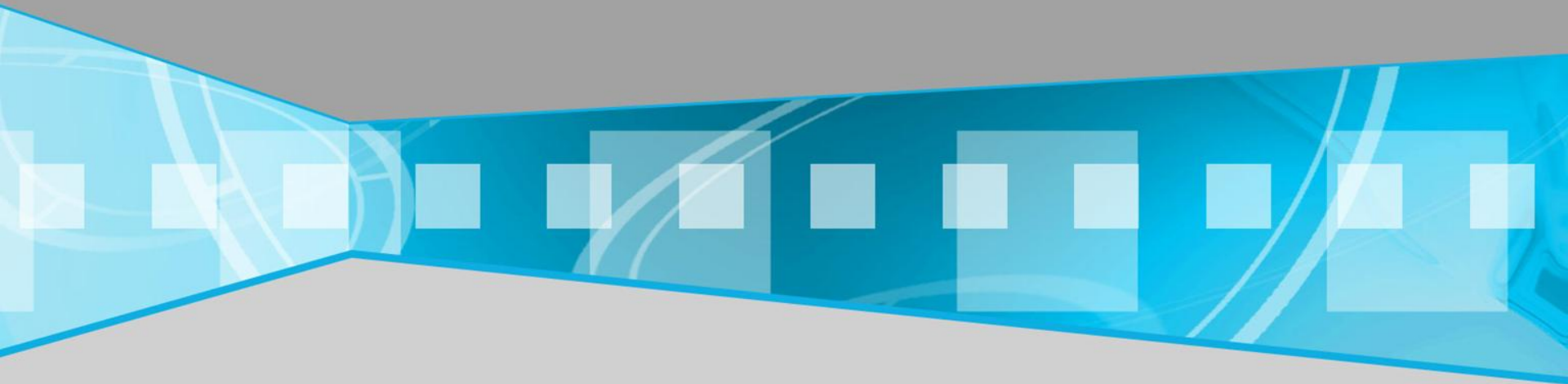
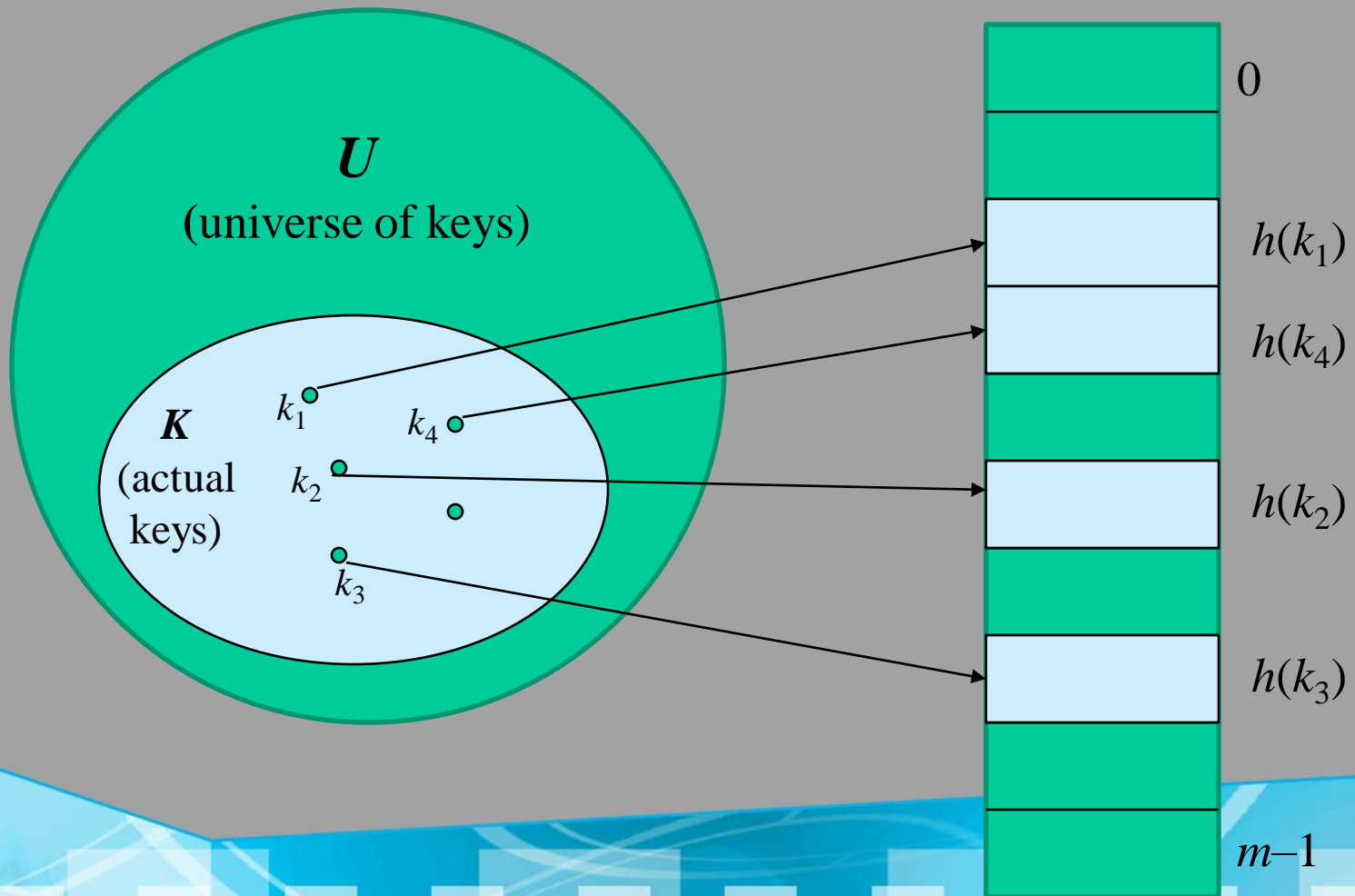


HASH TABLE



HASHING



HASHING

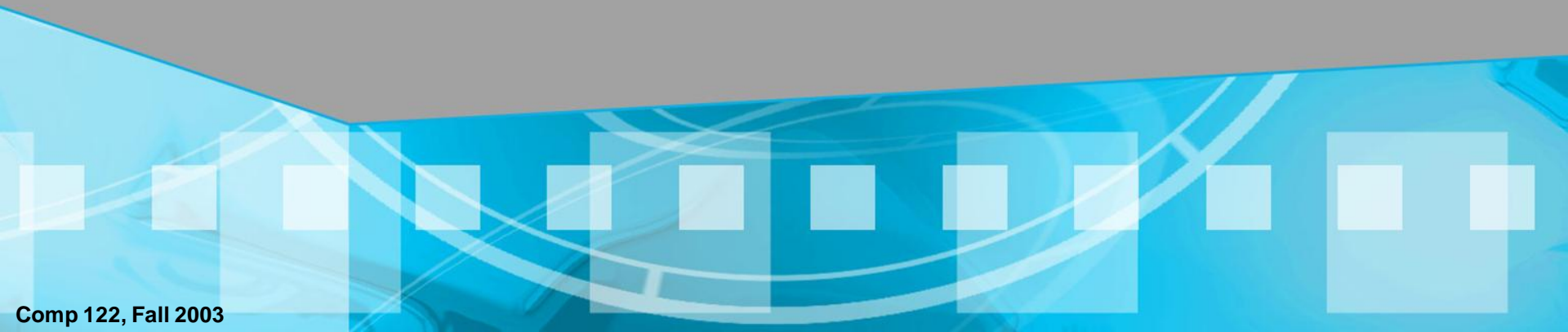
Hash function h : Mapping from U to the slots of a hash table $T[0..m-1]$.

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

With arrays, key k maps to slot $A[k]$.

*With hash tables, key k maps or “**hashes**” to slot $T[h[k]]$.*

*$h[k]$ is the **hash value** of key k .*



HASHING EXAMPLE



KEY A



KEY B



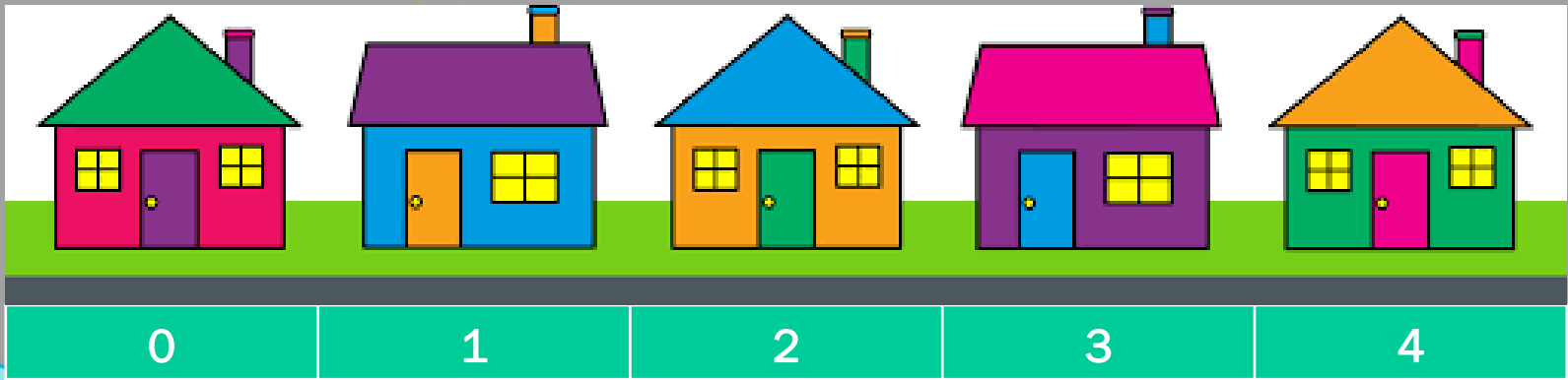
KEY C



KEY D



KEY E



HASHING EXAMPLE

0	1	2	3	4	5	6	7	8	9	10
	D		Z			C	Q			

Figure 10.3: A lookup table with length 11 for a map containing entries (1,D), (3,Z), (6,C), and (7,Q).

Hash Tables

Notation:

U – Universe of all possible keys.

K – Set of keys actually stored in the dictionary.

$|K| = n$.

When U is very large,

Arrays are not practical.

$|K| \ll |U|$.

Use a table of size proportional to $|K|$ – The hash tables.

However, we lose the direct-addressing ability.

Define functions that map keys to slots of the hash table.

HASING

Hash function h : Mapping from U to the slots of a hash table $T[0..m-1]$.

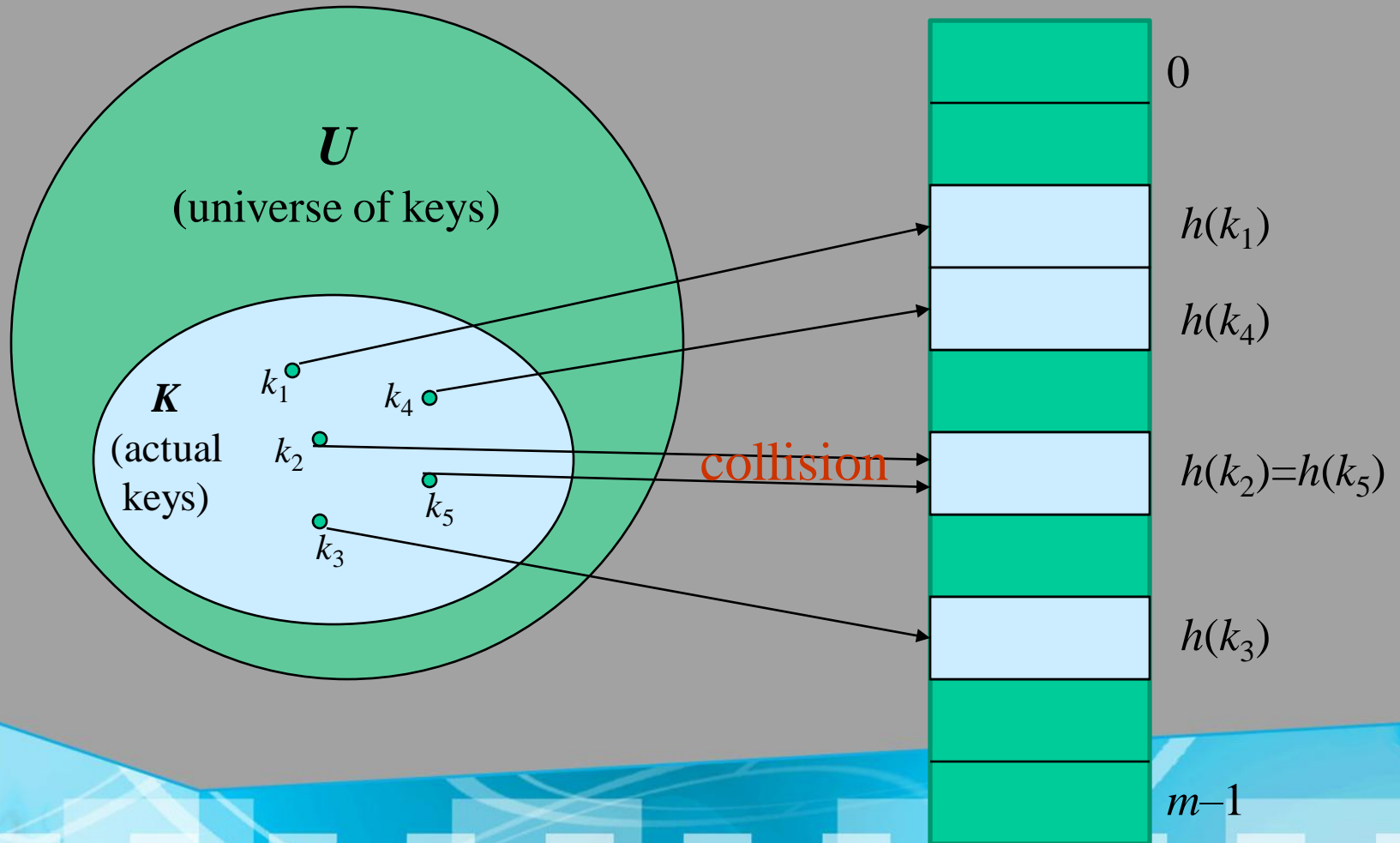
$$h : U \rightarrow \{0, 1, \dots, m-1\}$$

With arrays, key k maps to slot $A[k]$.

*With hash tables, key k maps or “**hashes**” to slot $T[h[k]]$.*

*$h[k]$ is the **hash value** of key k .*

COLLISION



COLLISION



KEY A



KEY B



KEY C



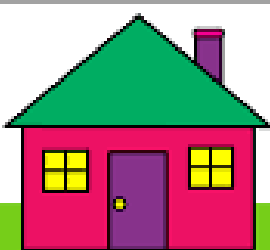
KEY D



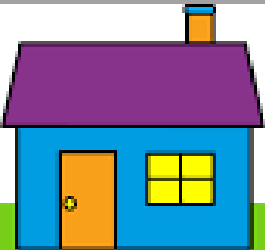
KEY E



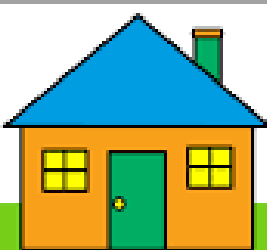
COLLISION



0



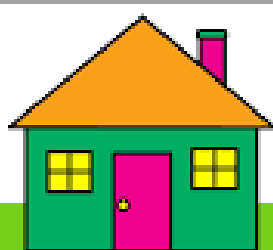
1



2



3



4

COLLISION

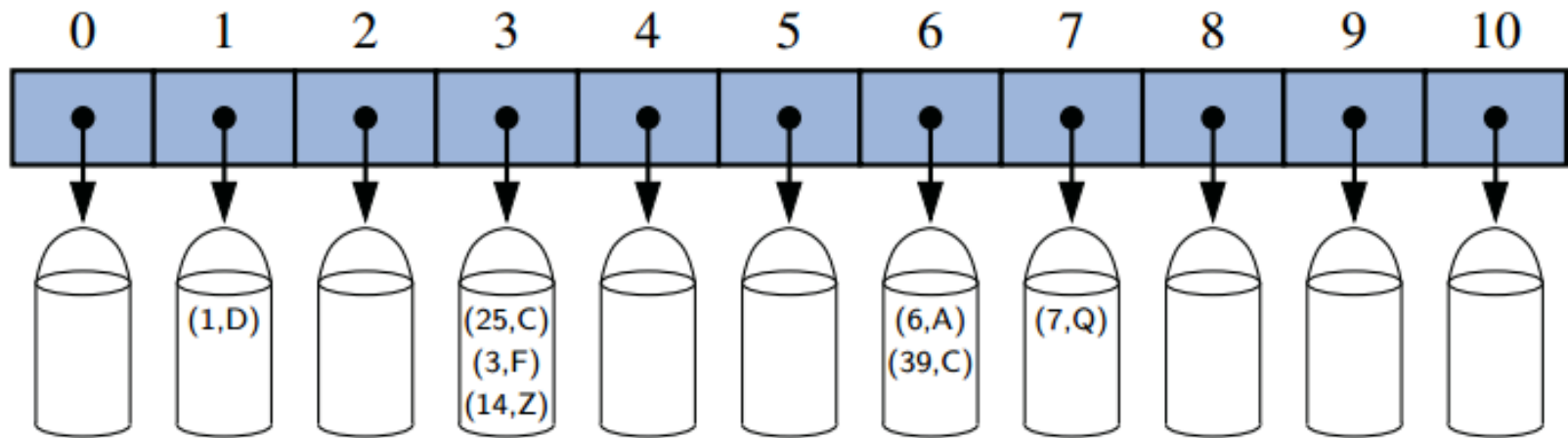


Figure 10.4: A bucket array of capacity 11 with entries (1,D), (25,C), (3,F), (14,Z), (6,A), (39,C), and (7,Q), using a simple hash function.

ISSUES WITH HASHING

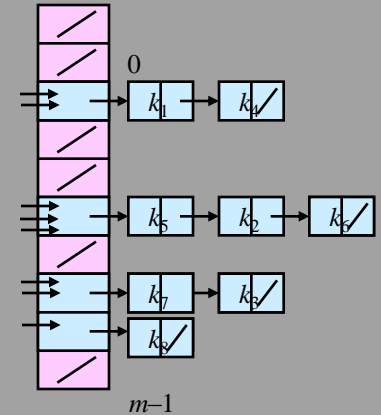
- *Multiple keys can hash to the same slot – collisions are possible.*
Design hash functions such that collisions are minimized.
But avoiding collisions is impossible.
Design collision-resolution techniques.
- *Search will cost $\Theta(n)$ time in the worst case.*
However, all operations can be made to have an expected complexity of $\Theta(1)$.

Methods of Resolution

Chaining:

Store all elements that hash to the same slot in a linked list.

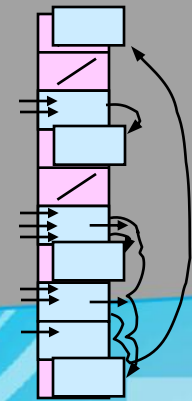
Store a pointer to the head of the linked list in the hash table slot.



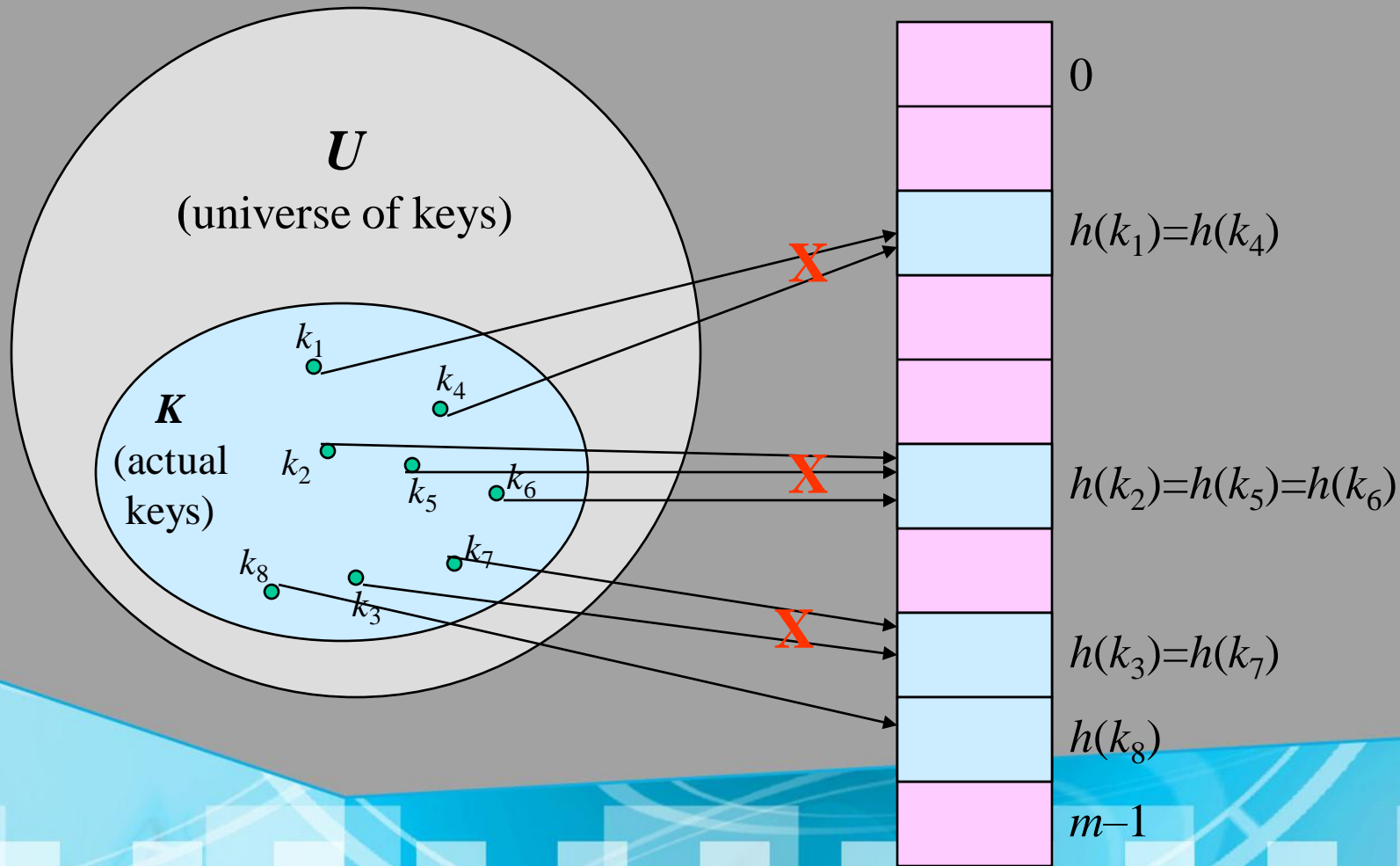
Open Addressing:

All elements stored in hash table itself.

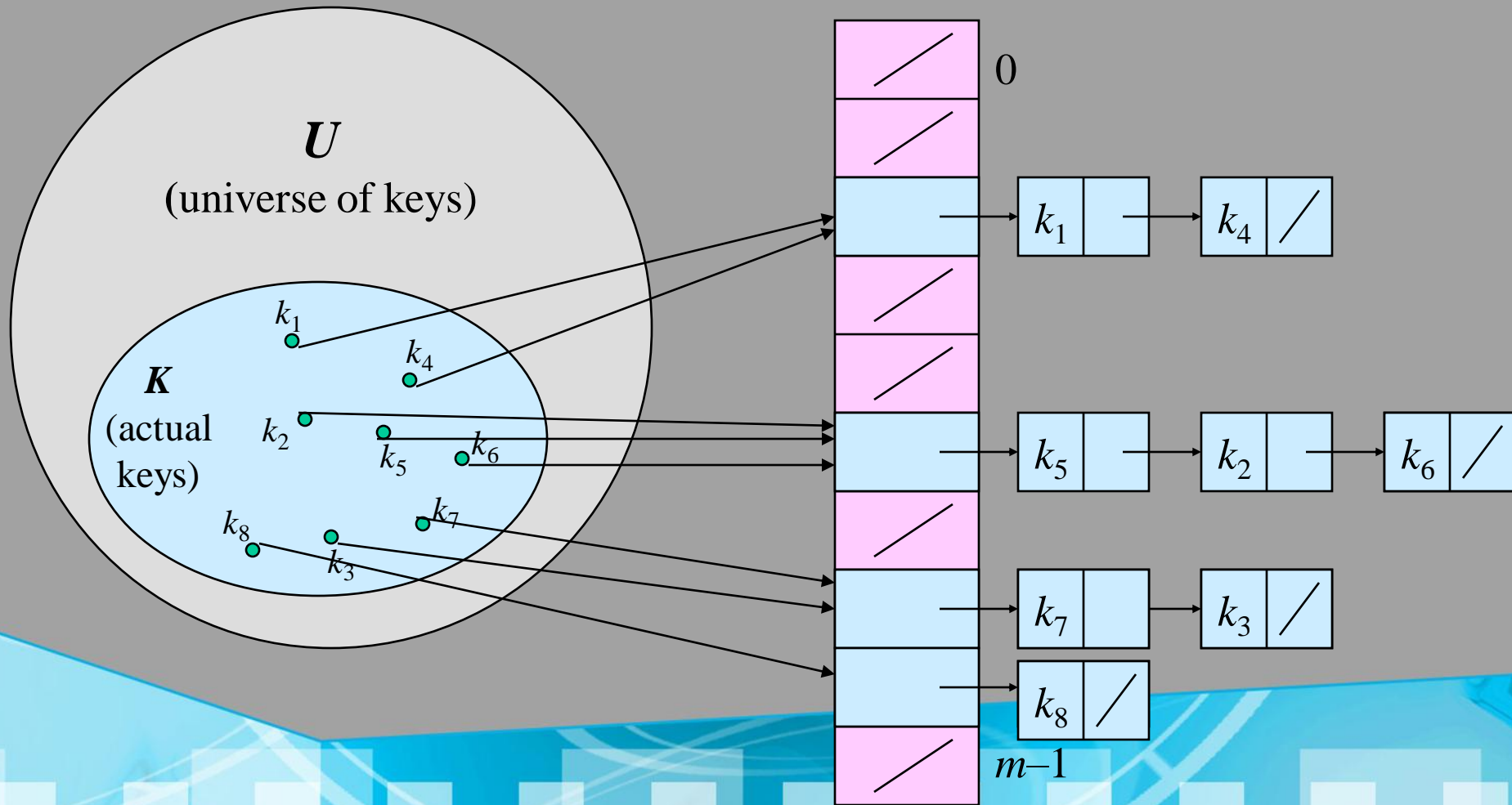
When collisions occur, use a systematic (consistent) procedure to store elements in free slots of the table.



Collision Resolution by Chaining



Collision Resolution by Chaining



Hashing with Chaining

Dictionary Operations:

Chained-Hash-Insert (T, x)

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

Worst-case complexity – $O(1)$.

Chained-Hash-Delete (T, x)

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

Worst-case complexity – proportional to length of list with singly-linked lists. $O(1)$ with doubly-linked lists.

Chained-Hash-Search (T, k)

Search an element with key k in list $T[h(k)]$.

Worst-case complexity – proportional to length of list.

Analysis on Chained-Hash-Search

Load factor $\alpha = n/m$ = average keys per slot.

m – number of slots.

n – number of elements stored in the hash table.

Worst-case complexity: $\Theta(n)$ + time to compute $h(k)$.

Average depends on how h distributes keys among m slots.

Assume

Simple uniform hashing.

Any key is equally likely to hash into any of the m slots, independent of where any other key hashes to.

$O(1)$ time to compute $h(k)$.

Time to search for an element with key k is $\Theta(|T[h(k)]|)$.

Expected length of a linked list = load factor = $\alpha = n/m$.

Good Hash Functions

Satisfy the assumption of simple uniform hashing.

Not possible to satisfy the assumption in practice.

Often *use heuristics*, based on the domain of the keys, to create a hash function that performs well.

Regularity in key distribution should not affect uniformity. *Hash value should be independent of any patterns that might exist in the data.*

E.g. Each key is drawn independently from U according to a probability distribution P :

$$\sum_{k:h(k)=j} P(k) = 1/m \quad \text{for } j = 0, 1, \dots, m-1.$$

An example is the division method.

Keys as Natural Numbers

Hash functions assume that the keys are natural numbers.

When they are not, have to interpret them as natural numbers.

Example: *Interpret a character string as an integer expressed in some radix notation. Suppose the string is CLRS:*

ASCII values: C=67, L=76, R=82, S=83.

There are 128 basic ASCII values.

So, CLRS = $67 \cdot 128^3 + 76 \cdot 128^2 + 82 \cdot 128^1 + 83 \cdot 128^0$ = 141,764,947.

Division Method

Map a key k into one of the m slots by taking the remainder of k divided by m . That is,

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

Example: $m = 31$ and $k = 78 \Rightarrow h(k) = 16$.

Advantage: *Fast, since requires just one division operation.*

Disadvantage: *Have to avoid certain values of m .*

Don't pick certain values, such as $m=2^p$

Or hash won't depend on all bits of k .

Good choice for m :

Primes, not too close to power of 2 (or 10) are good.

Multiplication Method

If $0 < A < 1$, $h(k) = \lfloor m (kA \bmod 1) \rfloor = \lfloor m (kA - \lfloor kA \rfloor) \rfloor$
where $kA \bmod 1$ means the fractional part of kA , i.e.,
 $kA - \lfloor kA \rfloor$.

Disadvantage: Slower than the division method.

Advantage: Value of m is not critical.

Typically chosen as a power of 2, i.e., $m = 2^p$, which makes implementation easy.

Example: $m = 1000$, $k = 123$, $A \approx 0.6180339887...$

$$\begin{aligned} h(k) &= \lfloor 1000(123 \cdot 0.6180339887 \bmod 1) \rfloor \\ &= \lfloor 1000 \cdot 0.018169... \rfloor = 18. \end{aligned}$$

The MAD Method(*Multiply-Add-and-Divide*)

- $[(ai+b) \bmod p] \bmod N$
 - N is the size of the bucket array
 - p is a prime number larger than N , and a and b are integers chosen at random from the interval $[0, p-1]$, with $a > 0$.
- This compression function is chosen in order to eliminate repeated patterns in the set of hash codes and get us closer to having a “good” hash function, that is, one such that the probability any two different keys collide is $1/N$. This good behavior would be the same as we would have if these keys were “thrown” into A uniformly at random.

MAP _ DEFINITION

A set is a collection that lets you quickly find an existing element.

A map stores key/value pairs. You can find a value if you provide the key.

For example, you may store a table of employee records, where the keys are the employee IDs and the values are Employee objects.