

Data Structures & Algorithms

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Recap

- Elementary data structures
- ADT vs. Data structures
- Array based vs. linked node implementation
- Worst case time complexity to help us choose based on our needs

Today's Objectives

- What is a “MAP or Dictionary ADT”?
- What choices do we have to implement a MAP?
- What is a hash function and a hash table?
- What is collision and how to handle it?
- How to analyze time complexity of a Hash Map?

Map or Dictionary

Map or Dictionary

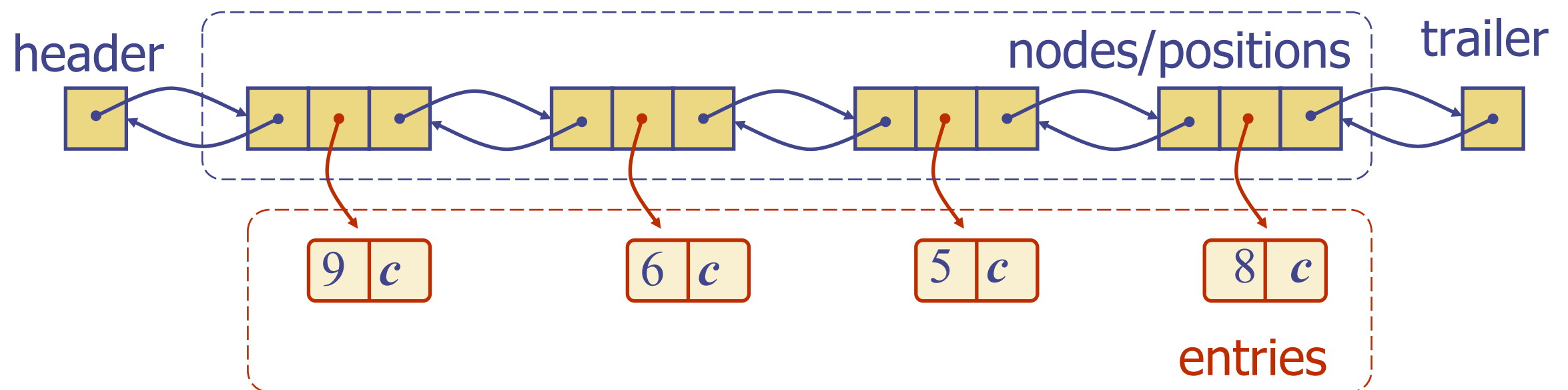
- Models a **searchable** **dynamic** set of **key-value** entries
- Main operations are: *searching*, *inserting*, and *deleting* items
- Applications:
 - Compiler symbol table
 - A news indexing service

The Map ADT

- **get(k):** if the map M has an entry with key k, return its associated value; else, return null
- **put(k, v):** insert entry (k, v) into the map M; if key k is not already in M, then return null; else, return old value associated with k
- **remove(k):** if the map M has an entry with key k, remove it from M and return its associated value; else, return null
- **size(), isEmpty()**
- **entrySet():** return an iterable collection of the entries in M
- **keySet():** return an iterable collection of the keys in M
- **values():** return an iterator of the values in M

A Simple List-Based Map

- We can implement a map using an unsorted list
- We store the items of the map in a list S (based on a doublylinked list), in arbitrary order



The get(k) Algorithm

Algorithm get(k):

while map.hasNext() **do**

p = map.next() { the next element in the map }

if p.element().getKey() = k **then**

return p.element().getValue()

return null {there is no entry with key equal to k }

The put(k,v) Algorithm

Algorithm put(k,v):

while map.hasNext() **do**

 p = map.next()

if p.element().getKey() = k **then**

 t = p.element().getValue()

 map.set(p,(k,v))

return t {return the old value}

map.addLast((k,v))

n = n + 1 {increment variable storing number of entries}

return null { there was no entry with key equal to k }

The remove(k) Algorithm

Algorithm remove(k):
while map.hasNext() **do**
 p = map.next()
 if p.element().getKey() = k **then**
 t = p.element().getValue()
 map.remove(p)
 n = n - 1 {decrement number of entries}
 return t {return the removed value}
return null {there is no entry with key equal to k}

Performance of a List-Based Map

- Performance:
 - **put** takes $O(1)$ time if we can insert the new item at the beginning or at the end of the sequence – assuming unique keys
 - **get** and **remove** take $O(n)$ time since in the worst case (the item is not found) we traverse the entire sequence to look for an item with the given key

Hash Map

Let's Start With this Question

- How much time does it take to lookup an item in an array, if you already know its index?

Example

- Suppose you're writing a program to access employee records for a company with 1000 employees.
 - ❖ Each employee has a number from 1(founder) to 1000 (the most recent worker)
 - ❖ Employees are seldom laid off, and even when they are, their record stays in the database.
- Goal: fastest possible access to any individual record

Example (cont.)

- The easiest way to do this is by using an array (we already know the size)
- Each employee record occupies one cell of the array
- The index number of the cell is the employee number

```
empRecord rec = databaseArray[72];
```

```
databaseArray[totalEmployees++] = newRecord;
```

Example (cont.)

- Direct-Access-Table

DIRECT-ADDRESS-SEARCH(T, k)

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

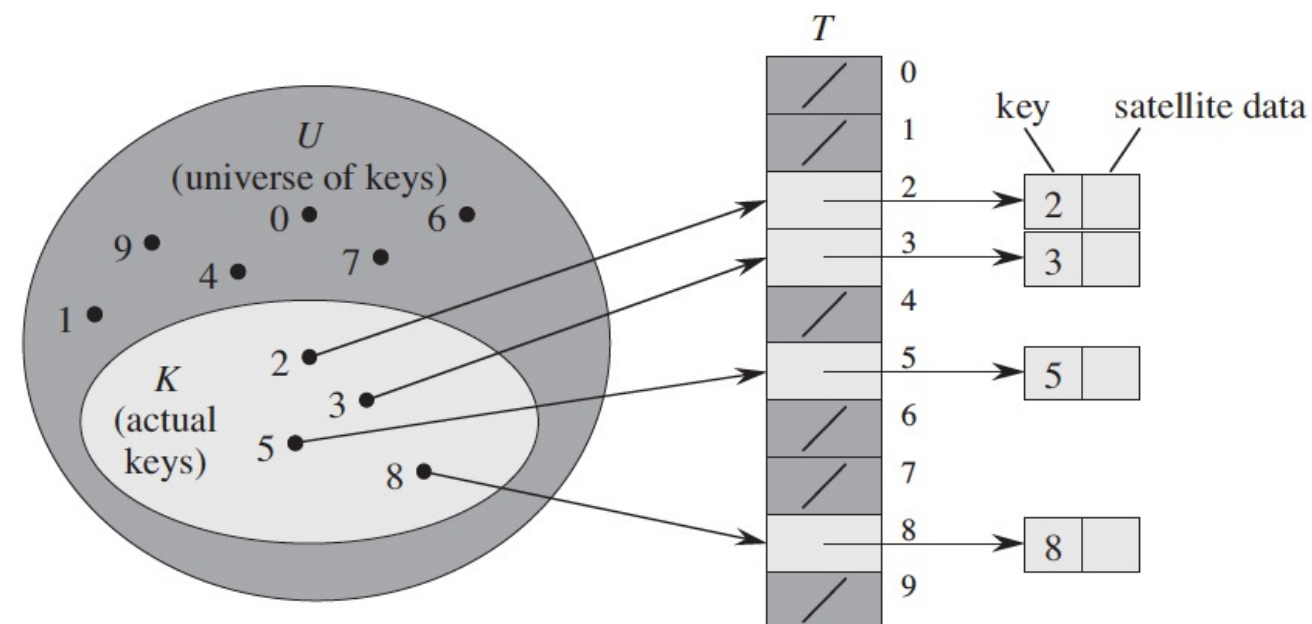
DIRECT-ADDRESS-INSERT(T, x)

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.



Example (cont.)

- The **speed** and **simplicity** of data access using this **direct-access-table** makes it very attractive.
- **However**, it works in our example only because keys are well organized
 - ❖ Sequentially from 1 to a known maximum
 - ❖ No deletions required
 - ❖ New items can be added sequentially at the end

Example (cont.)

- But mostly, the keys are **not so well behaved**
- A simple example would be when **keys are of type String**.
 - ❖ **Array indexing requires integer**
- **One more problem:** Even when using integers, the value could be outside of the range of the array

What Did We Learn From The Example?

- Arrays are very fast when it comes to accessing an item based on its index
- But “key” → “index” mapping only works when
 - ❖ keys are integers, and
 - ❖ are within the bound, and
 - ❖ do not change

Hash Map

- An efficient implementation of a Map, implemented as a Hash Table

Hash Table

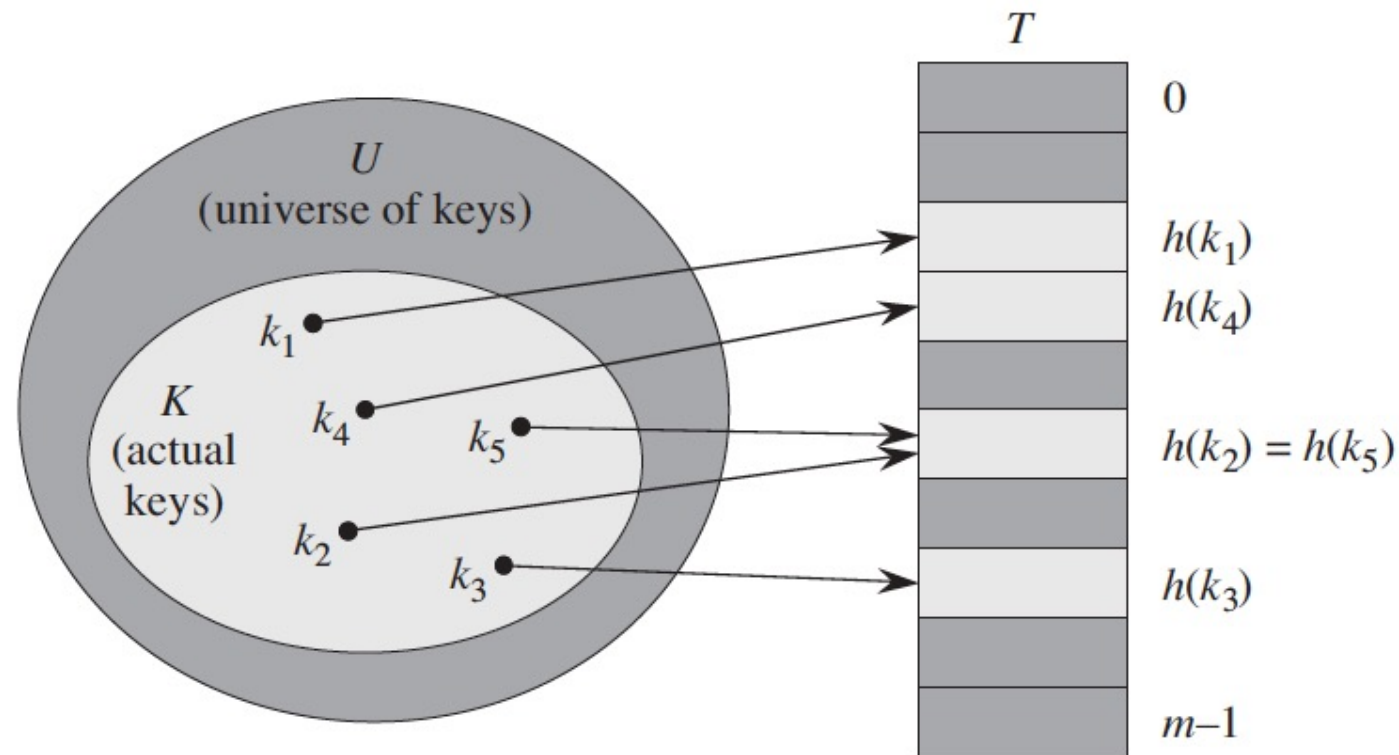
- A **hash table** for a given key type consists of
 1. Hash function h (a mathematical way of mapping an arbitrary **key** to an **index** in the array)
 2. Array (which is called **table**) of size N or m

Hash Table

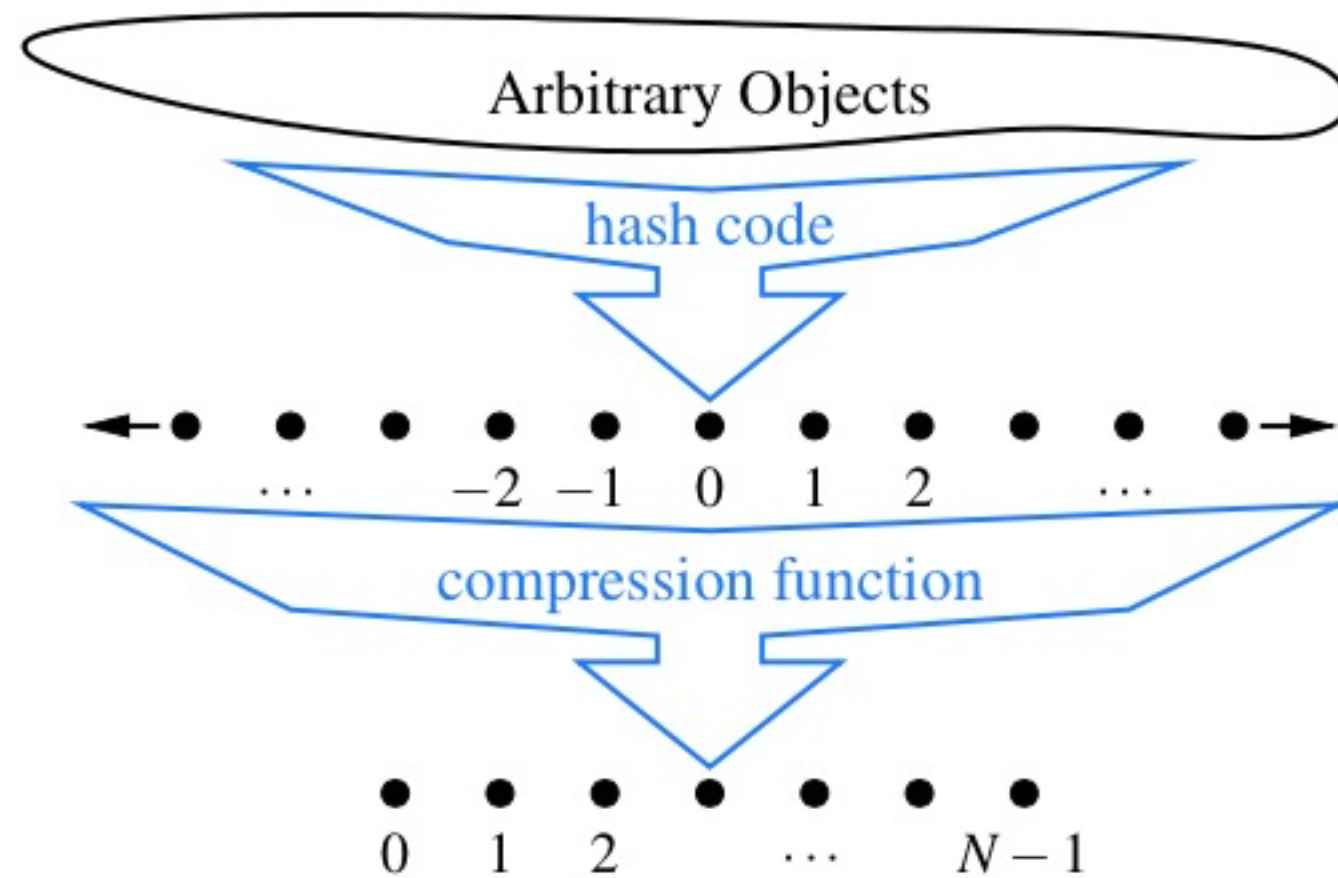
- A **hash table** for a given key type consists of
 - ❖ Hash function h (a mathematical way of mapping an arbitrary key to an index in the array)
 - ❖ Array (which is called table) of size N or m
- When implementing a map with a hash table, the goal is to store item (k, o) at index $i = h(k)$, where k is the key, o is the data object

Hash Function

- A **hash function** h maps keys of a given type to integers in a fixed interval $[0, m - 1]$



Parts of a Hash Function



General Hash Functions

- A hash function is usually specified as the **composition** of two functions:
- The hash code is applied first, and the compression function is applied next on the result, i.e.,

Hash code:

$h_1: \text{keys} \rightarrow \text{integers}$

Compression function:

$h_2: \text{integers} \rightarrow [0, N - 1]$

$$h(x) = h_2(h_1(x))$$

Ideal Hash Function

- ❖ Every resulting hash value has exactly one input that will produce it
- ❖ Same key hashes to the same index (**repeatable**)
- ❖ Hash value is widely different if even a single bit is different in the key (**avalanche**)
- ❖ Should work in general (for different types)

Some Common Hash Codes

key \rightarrow Integer

Hash Codes

1. Memory address as the Hash Code:

- We reinterpret the memory address of the key object as its integer hash code (default hash code of all Java objects)
- Good in general, except that it is not repeatable

Hash Codes (cont.)

2. Integer cast (Use the bit representation of the object as a hash code):

- We reinterpret the bits of the key as an integer
- Suitable for keys of length less than or equal to the number of bits of the integer type (e.g., byte, short, int and float in Java)

Hash Codes (cont.)

3. **Component sum:**

- We partition the bits of the key into components of fixed length (e.g., 16 or 32 bits) and we sum the components
- Fails to treat permutations differently (“abc”, “cba”, “cab”)

Hash Codes (cont.)

4. Polynomial accumulation:

- We partition the bits of the key into a sequence of components of fixed length (e.g., 8, 16 or 32 bits)

$$a_0, a_1, \dots, a_{n-1}$$

- We evaluate the polynomial

$$p(z) = a_0 + a_1 z + a_2 z^2 + \dots \\ \dots + a_{n-1} z^{n-1}$$

at a fixed value z

Hash Codes (Summary)

- Memory Address
- Integer Cast
- Component Sum
- Polynomial Accumulation

Two Common Compression Functions

Hash code → Index

Compression Functions

1. Division:

- $h_2(y) = y \bmod N$
- *y is the integer has code, N is the size of the array*
- N is usually chosen to be a prime
- Helps “spread out” the distribution of hashed values
- Try inset keys with hash codes {200, 205, 210, 215, ..., 600}
into a table size of 100 vs. 101

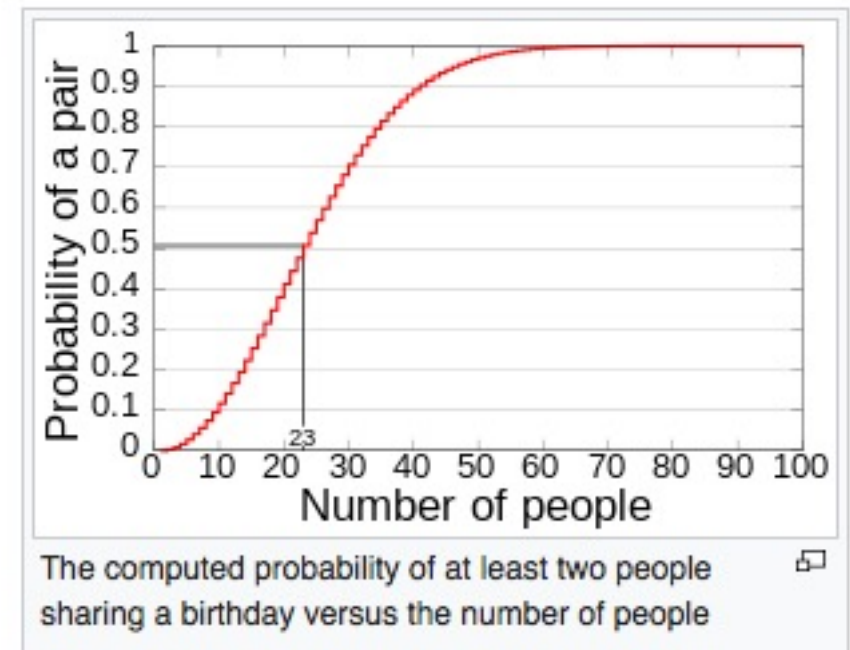
Compression Functions

2. Multiply, Add and Divide (MAD)

- $h_2(y) = [(ay + b) \bmod p] \bmod N$
- p is a prime number larger than N
- a and b are integers from the interval $[0, p - 1]$, with $a > 0$

Things to Remember

1. If n items are placed in m buckets, and n is greater than m , one or more buckets contain two or more items (**Pigeonhole Principle**)
 - ❖ This is called **collision** (two keys hash to the same index)
2. Birthday paradox



Collisions

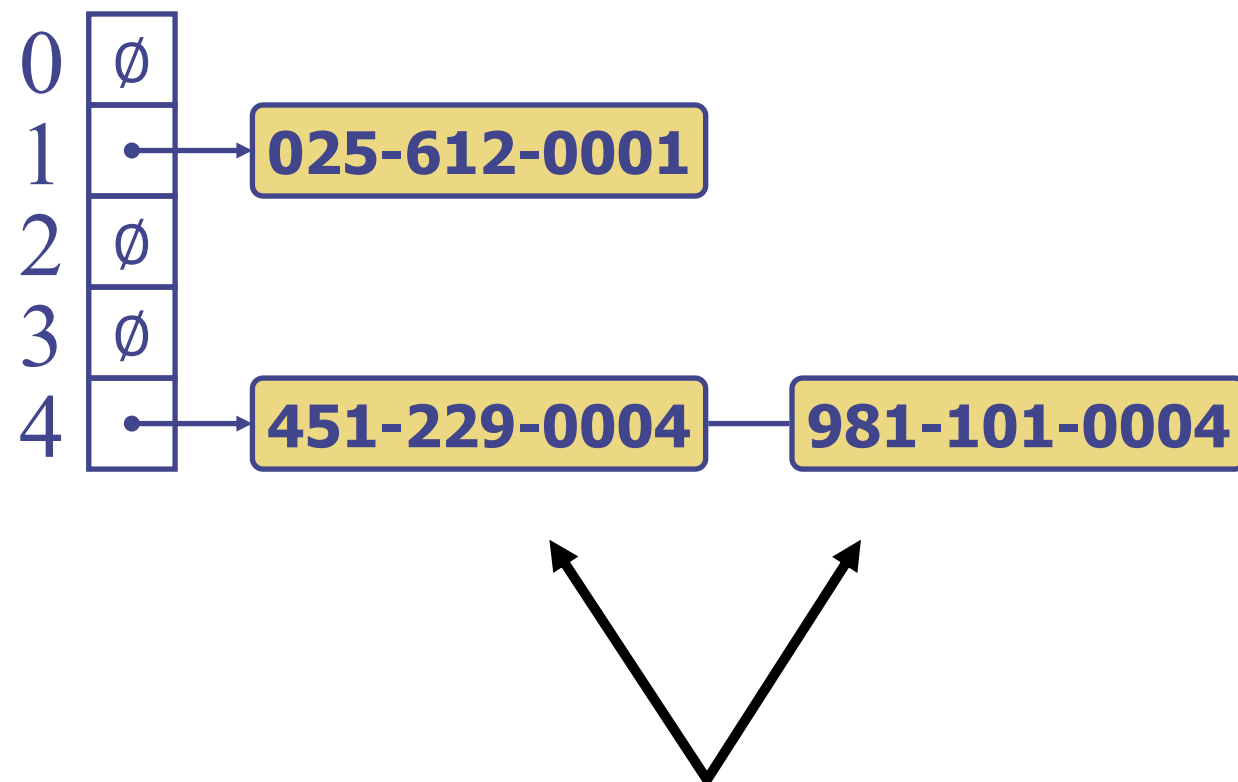
- So collisions are inevitable
- Our goal should therefore be to minimize collisions
- We will achieve it through:
 - ❖ Generating better hash codes
 - ❖ Performing better compression
 - ❖ Handling collisions

Collision Handling

- Let n be the number of items inserted into a hash table of size m
- Two main ways to handle collisions
 - ❖ Separate Chaining: m much smaller than n
 - ❖ Open Addressing: m much larger than n

Collision Handling

1. **Separate Chaining:** let each cell in the table point to a linked list of entries that map there



A Collision: indexed to the same position in the table

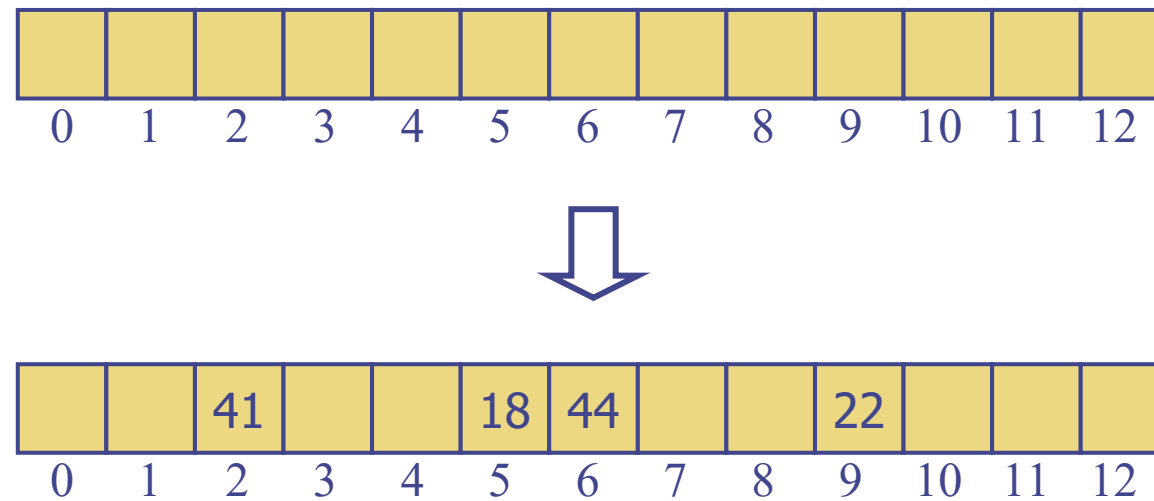
Collision Handling

2. **Open Addressing:** the colliding item is placed in a different cell of the table
 - A. **Linear Probing:** handles collision by placing the item in the next (circularly) available cell
 - ❖ Each cell inspected is called a **probe**
 - ❖ Colliding items lump together, causing future collisions to cause a longer sequence of probes

Example

□ Example:

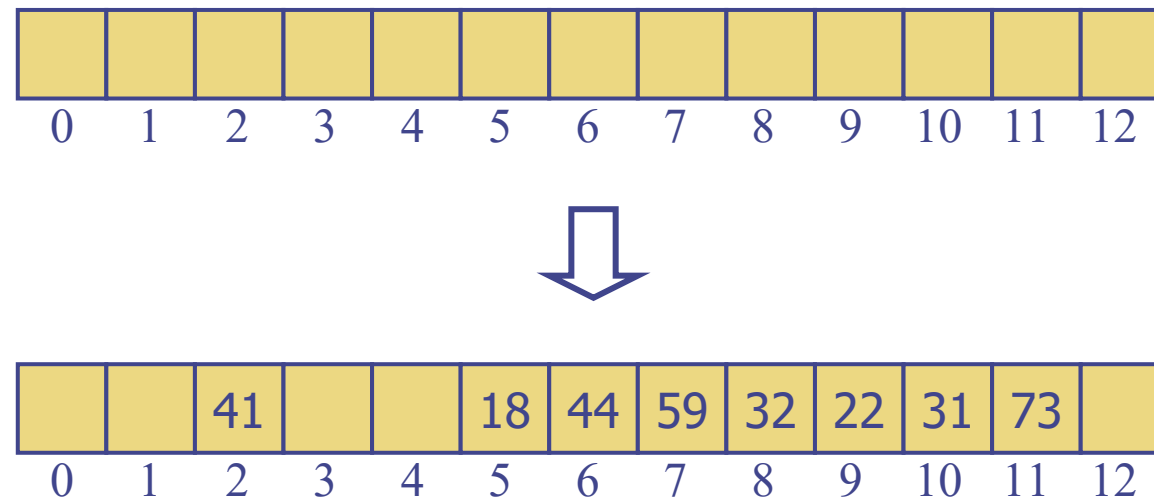
- *Linear probing*
- $h(x) = x \bmod 13$
- Insert keys 18, 41, 22, 44, 59, 32, 31, 73, in this order



Example

□ Example:

- *Linear probing*
- $h(x) = x \bmod 13$
- Insert keys 18, 41, 22, 44, 59, 32, 31, 73, in this order



Example

- Example:
 - *Linear probing*
 - $h(x) = x \bmod 13$
 - How will you search for 44?

		41			18	44	59	32	22	31	73	
0	1	2	3	4	5	6	7	8	9	10	11	12

Search with Linear Probing

- Consider a hash table A that uses linear probing
- **get(k)**
 - We start at cell $h(k)$
 - We probe consecutive locations until one of the following occurs
 - ❖ An item with key k is found, or
 - ❖ *An empty cell is found*, or
 - ❖ N cells have been unsuccessfully probed

Algorithm **get(k)**

```
 $i \leftarrow h(k)$   
 $p \leftarrow 0$   
repeat  
     $c \leftarrow A[i]$   
    if  $c = \emptyset$   
        return null  
    else if  $c.getKey() = k$   
        return  $c.getValue()$   
    else  
         $i \leftarrow (i + 1) \bmod N$   
         $p \leftarrow p + 1$   
until  $p = N$   
return null
```

Example

□ Example:

- *Linear probing*
- $h(x) = x \bmod 13$
- Let's say 18 has been deleted
- How will you search for 44?

		41				44	59	32	22	31	73	
0	1	2	3	4	5	6	7	8	9	10	11	12

Updates with Linear Probing

- To handle insertions and deletions, we introduce a special object, called *DEFUNCT*, which replaces deleted elements
- *remove*(*k*)
 - ❖ We search for an entry with key *k*
 - ❖ If such an entry (*k*, *o*) is found, we replace it with the special item *DEFUNCT* and we return element *o*
 - ❖ Else, we return *null*

Example

- Example:

- *Linear probing*
- $h(x) = x \bmod 13$
- Let's say 18 has been deleted
- How will you search for 44?

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Collision Handling

B. Open Addressing: the colliding item is placed in a different cell of the table

Double Hashing: uses a secondary hash function $d(k)$ and handles collision by placing an items in the first available of cell of the series

$$(h(k) + jd(k)) \bmod N$$

$$\text{for } j = 1, \dots, N - 1$$

Double Hashing

- The secondary hash function cannot have zero values
- The table size ***N*** must be prime to allow probing of all the cells.

Double Hashing

- Common choice of compression function for the secondary hash function:

$$d(k) = q - (k \bmod q)$$

where

$$q < N$$

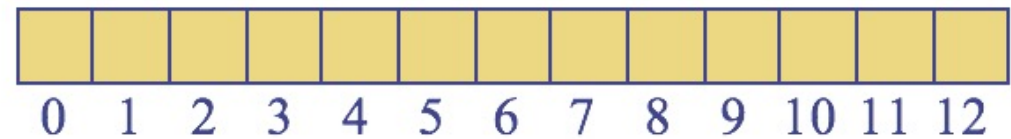
q is a prime

The possible values for $d(k)$ are

$$1, 2, \dots, q$$

Example

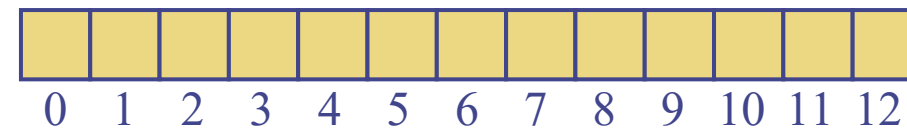
- Consider a hash table storing integer keys that handles collision with double hashing
 - $N = 13$
 - $h(k) = k \bmod 13$
 - $d(k) = 7 - k \bmod 7$
- Insert keys 18, 41, 22, 44, 59, 32, 31, 73, in this order



Example

- Consider a hash table storing integer keys that handles collision with double hashing
 - $N = 13$
 - $h(k) = k \bmod 13$
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- Insert keys 18, 41, 22, 44, 59, 32, 31, 73, in this order

k	$h(k)$	$d(k)$	Probes	
18	5	3	5	
41	2	1	2	
22	9	6	9	
44	5	5	5	10
59	7	4	7	
32	6	3	6	
31	5	4	5	9 0
73	8	4	8	



Example

- Consider a hash table storing integer keys that handles collision with double hashing
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73	8	4	8	

0	1	2	3	4	5	6	7	8	9	10	11	12



31		41			18	32	59	73	22	44		
0	1	2	3	4	5	6	7	8	9	10	11	12

Analysis of get(k) in Separate Chaining

- **Worst case:** all elements get hashed to the same index or bucket, thus search will take $O(n)$
- But if hash function is chosen well, the worst case is highly unlikely
- Thus we analyze the **expected** time complexity using the *load factor*

Analysis of get(k) in Separate Chaining – In Tutorial

- **Load factor (α)**: the ratio of n and m , represents the expected length of a chain
- The expected length of a chain in this case is $O(\alpha)$
- Thus, the expected time complexity in terms of the *load factor* - $O(1 + \alpha)$
- It is usually made sure that α doesn't exceed some constant, thus $O(1)$

Analysis of get(k) in Open Addressing - In Tutorial

- Load factor (α): the ratio of n and m
- The worst case time complexity in terms of the *load factor* - $O\left(\frac{1}{1-\alpha}\right)$
- It is usually made sure that α doesn't exceed some constant, thus $O(1)$

Did we achieve today's objectives?

- What is a “MAP ADT”?
- What choices do we have to implement a MAP?
- What is a hash function and a hash table?
- What is collision and how it handle it?
- How to analyze time complexity of a Hash Map?