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# Data Structures and Algorithms

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## Lecture 9: Searching, and Hashing

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# Outline of Today's Lecture

- Searching
  - Unsorted and Sorted Arrays
  - Self-Organizing Lists
  - Bit Vectors for Representing Sets
- Hashing
  - Hash Tables
  - Hash Functions
  - Open and Closed Hashing
  - Operations

# Problem definition

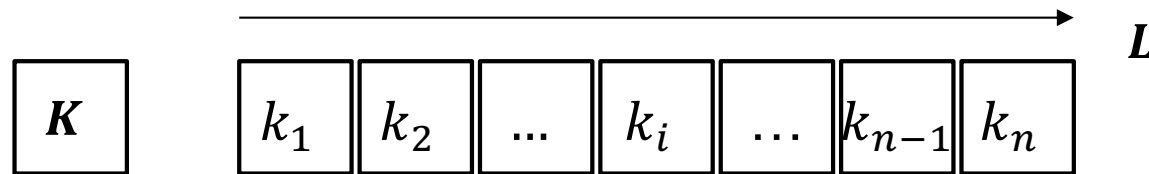
- Suppose we have a collection  $L$  of  $n$  records of the form  $(k_1, I_1), (k_2, I_2), \dots, (k_n, I_n)$ , where  $I_j$  is information associated with key  $k_j$  from record  $(k_j, I_j)$  for  $1 \leq j \leq n$ .
- Given a query  $K$ , the **Search Problem** is to locate a record  $(k_s, I_s)$  in  $L$  such that  $k_s = K$  (if one exists).
- Two types of **query** problems
  - An **exact-match query** is a search for the record whose key values matches a specified key value.
  - A **range query** is a search for all records whose key value falls within a specified range of key values.

# Search algorithms

- Three general approaches
  - Sequential and list methods
  - Direct access by key value -- hashing
  - Tree indexing methods (next lecture)

# Search in unsorted arrays (1/2)

- The **sequential search** algorithm
  - Basic idea: search from the **beginning** to the **end**
  - The simplest form of search



- Best case:  $\Theta(1)$
- Worst case:  $\Theta(n)$
- Average case:  $\Theta(n/2) = \Theta(n)$
- Sometimes called **linear search**.

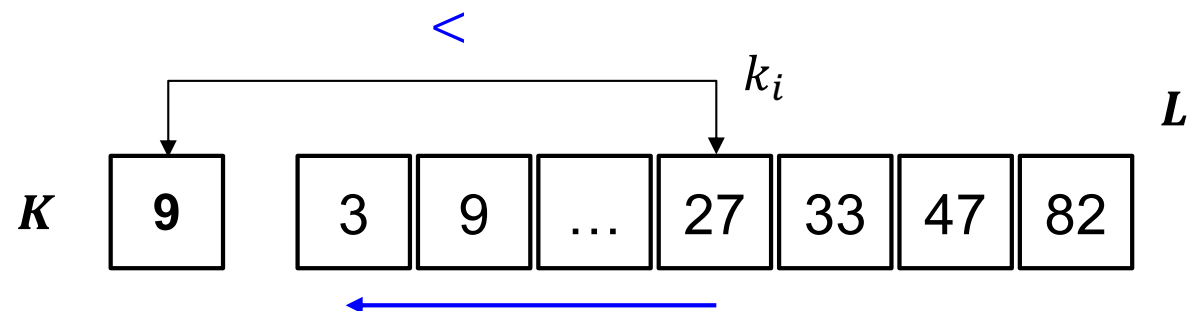
# Search in unsorted arrays (2/2)

- A simple implementatin for sequential search

```
/* Find the position in A that holds value K, if any
does */
int sequential(int A[], int size, int K) {
    for (int i=1; i<size; i++) // For each element
        if (A[i] == K) // if we found it
            return i; // return this position
    return size; // Otherwise, return the array length
}
```

# Search in sorted arrays

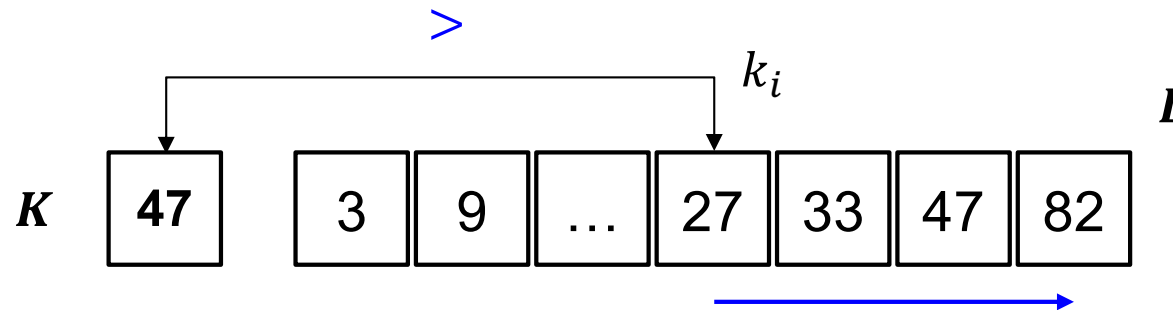
- Sequential search is somewhat slow.  $\Theta(n)$
- One way to reduce search time is to preprocess the records by **sorting** them.
- Given a sorted array, an obvious improvement over simple linear search is to test if the **current element** in  $L$  is **greater than**  $K$ .



# Search in sorted arrays - Jump search (1/3)

## ■ Jump search

- Suppose we look first **position  $i$**  and find that  $K$  is bigger, then we rule out position  $i$  as well as position 0 to  $i - 1$ .

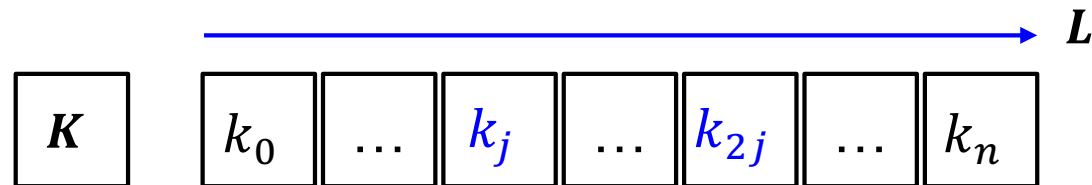


- What if we carry this to the extreme and look first at the **last position** in  $L$  and find that  $K$  is bigger?
  - Then we know **in one comparison** that  $K$  is not in  $L$ .



# Search in sorted arrays – Jump search (2/3)

- Basic idea of Jump search algorithm
  - For a jump size  $j$ , we check **every  $j$ -th** element in  $L$ .



- So long as  $K$  is greater than the checking values, we continue on.
  - Otherwise, we do a linear search on the piece of length  $j-1$  that we know brackets  $K$  if it is in the list.
- A typical **divide and conquer** algorithm.
- What is the right amount to jump?

## Search in sorted arrays - Jump search (3/3)

- Define  $m$  such that  $mj \leq n < (m + 1)j$ , then the total cost of this algorithm is at most  $m + j - 1$  3-way comparison. (**3-way**: less, equal, or greater)
- Therefore, the cost to run the algorithm on  $n$  items with a jump of size  $j$  is

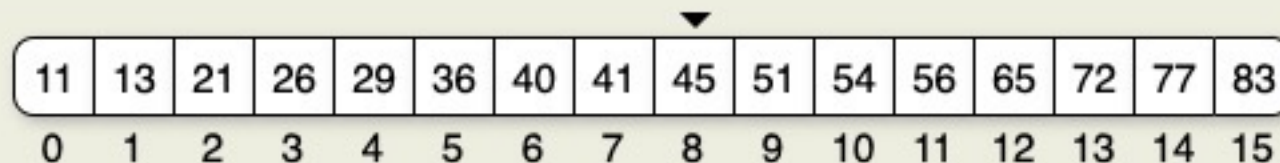
$$T(n, j) = m + j - 1 = \left\lceil \frac{n}{j} \right\rceil + j - 1$$

- Minimize the cost:
  - Take the derivative and solve for  $T'(n, j) = 0$  to find the minimum, which is  $j = \sqrt{n}$ .
- In this case, the worst case cost is roughly  $2\sqrt{n}$ .

# Search in sorted arrays - Binary search (1/3)

- Basic idea: recursion

```
// Return the position of an element in sorted array A with value K.  
// If K is not in A, return A.length.  
public static int binarySearch(int[] A, int K) {  
    int low = 0;  
    int high = A.length - 1;  
    while(low <= high) { // Stop when low and high meet  
        int mid = (low + high) / 2; // Check middle of subarray  
        if( A[mid] < K) low = mid + 1; // In right half  
        else if(A[mid] > K) high = mid - 1; // In left half  
        else return mid; // Found it  
    }  
    return A.length; // Search value not in A  
}
```



11	13	21	26	29	36	40	41	45	51	54	56	65	72	77	83
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

## Search in sorted arrays - Binary search (2/3)

- An **optimal** algorithm for a sorted list.
- Time complexity:  $\Theta(\log n)$
- If the data are not sorted, using binary search requires to pay the cost of sorting the data, e.g.,  $\Theta(\log n)$  by a **balanced binary search tree (BST)**.
- Two special forms of binary search (see book):
  - Dictionary or interpolation search
  - Quadratic binary search

# Performance Comparison ( $n=400M$ )

- Running time of the sequential search is about 200 ms
- Running time of the binary search is only 0.002 ms
- Binary search is about 100,000 times faster for  $n=400M$

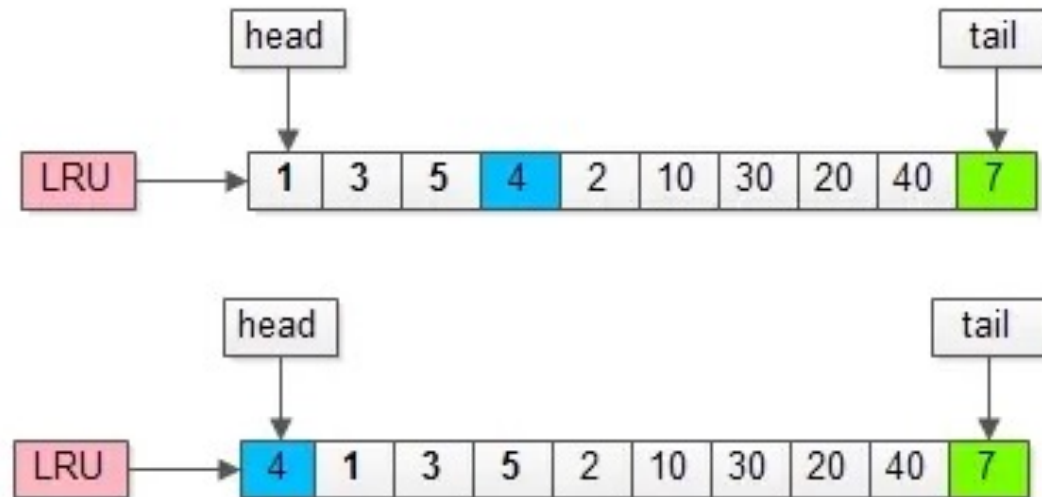
```
Sequential search, time:      188.933 (milli-seconds)
Binary search, time:         0.002 (milli-seconds)
```

# Self-organizing lists (1 / 6)

- Self-organizing lists are simply linked-lists of data
  - It reorganizes the data such that items that have been **accessed recently or more frequently**, are moved closer to the front of the list.
- Motivation for sorting by access frequency
  - Most searchable data sets contain some items that are accessed frequently, and many items that are accessed rarely.
  - Can speed up sequential search.

# Self-organizing lists (2/6)

- Usually, the frequencies are unknown.
- Self-organizing lists use a **heuristic** for deciding how to reorder the list.
  - Similar to the rules for managing **buffer pools**.  
E.g.,



# Self-organizing lists (3/6)

- Basic ideas
  - modify the order of records within the list based on the actual pattern of record access, by moving a found key nearer to the front of the list (insert and delete operations can stay the same).
- We consider three heuristics
  - Frequency Count
  - Move-To-Front
  - Transpose



# Self-organizing lists (4/6)

- Frequency Count
  - When a record is found, **move forward** the front of the list if its number of accesses becomes greater than a record preceding it.

	<u>ABCDEFGH</u>	<u>ABCDEFGH</u>	
F	<b>F</b> ABCDEGH	00000 <b>1</b> 00	<b>6</b>
D	<b>F</b> DABCEGH	000 <b>1</b> 0100	<b>5</b>
F	<b>F</b> DABCEGH	00010 <b>2</b> 00	<b>1</b>
G	<b>F</b> D <b>G</b> ABCEH	000102 <b>1</b> 0	<b>7</b>
E	<b>F</b> DG <b>E</b> ABCH	0001 <b>1</b> 210	<b>7</b>
G	<b>F</b> GDEABCH	000112 <b>2</b> 0	<b>3</b>
F	<b>F</b> GDEABCH	00011 <b>3</b> 20	<b>1</b>
A	<b>F</b> GDE <b>A</b> BCH	<b>1</b> 0011320	<b>5</b>
D	<b>F</b> G <b>D</b> EABCH	100 <b>2</b> 1320	<b>3</b>
F	<b>F</b> GDEABCH	10021 <b>4</b> 20	<b>1</b>
G	<b>F</b> GDEABCH	100214 <b>3</b> 0	<b>2</b>
E	<b>F</b> GD <b>E</b> ABCH	1002 <b>2</b> 430	<b>4</b>

# Self-organizing lists (5/6)

- Move-To-Front
  - When a record is found, **move** it to the front of the list.

	<u>ABCDEFGH</u>	
F	<b>F</b> ABCDEGH	6
D	<b>D</b> FABCEGH	5
F	<b>F</b> DABCEGH	2
G	<b>G</b> FDABCEH	7
E	<b>E</b> GFDABCH	7
G	<b>G</b> EFDABCH	2
F	<b>F</b> GEDABCH	3
A	<b>A</b> FGEDBCH	5
D	<b>D</b> AFGEBCH	5
F	<b>F</b> DAGEBCH	3
G	<b>G</b> FDAEBCH	4
E	<b>E</b> GFDABCH	5

# Self-organizing lists (6/6)

- Transpose
  - When a record is found, **swap** it with the record ahead of it.

	<u>ABCDEFGH</u>	
F	ABCD <b>F</b> EGH	6
D	AB <b>D</b> C FEGH	4
F	ABD <b>F</b> CEGH	5
G	ABDFC <b>G</b> EH	7
E	ABDFC <b>E</b> GH	7
G	ABDFC <b>G</b> EH	7
F	AB <b>F</b> DCGEH	4
A	<b>A</b> BFDCGEH	1
D	AB <b>D</b> FCGEH	4
F	AB <b>F</b> DCGEH	4
G	ABFD <b>G</b> CEH	6
E	ABFD <b>G</b> EC <b>H</b>	7

# Example: self-organizing lists

- Text compression and transmission
  - By the **move-to-front** rule
    1. If the word has been seen before, transmit the current position of the word in the list. Move the word to the front of the list.
    2. If the word is seen for the first time, transmit the word. Place the word at the front of the list.

The car on the left hit the car I left

The car on 3 left hit 3 5 I 5

# Bit vectors for representing sets

- Representing sets using a bit array with a bit position allocated for each potential member.
  - 1 denotes 'in the set'; 0 denote 'not in the set'.
- Example: a set of primes

0	0	1	1	0	1	0	1	0	0	0	1	0	1	0	0
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

- Benefits by the logical bit-wise operations
  - set union, intersection, and difference

# Hashing

# Hashing

Given  $n$  records with unique **keys**,

	insert	search	delete
■ Unsorted list	$\Theta(1)$	$\Theta(n)$	$\Theta(n)$
■ Sorted array	$\Theta(n)$	$\Theta(\log n)$	$\Theta(n)$
■ <i>Balanced</i> BST	$\Theta(\log n)$	$\Theta(\log n)$	$\Theta(\log n)$
■ <b>Magic array</b>	$\Theta(1)$	$\Theta(1)$	$\Theta(1)$

Sufficient "**magic**":

- Use **key** to map **array index** for a record in  $\Theta(1)$  time
- Search by direct access based on **key** value

# Data collection

- Data collection is a set of records (static or dynamic)
- Each record consists of two parts
  - A **key**: a unique identifier of the record.
  - **Data item**: it can be arbitrarily complex.
- The key is usually a number, but can be a string or any other data type.
  - Non-numbers are converted to numbers when applying hashing.



# Basic ideas of hashing

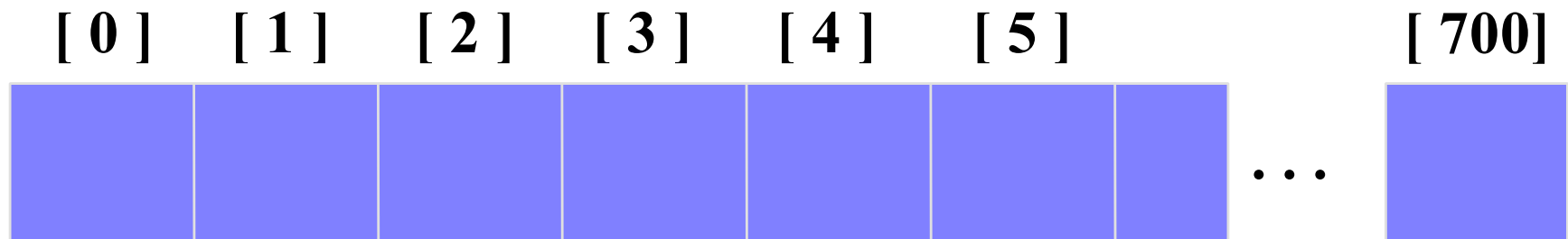
- Use *hash function* to map *keys* into positions in a *hash table*

## Ideally

- If data item or element  $e$  has key  $k$  and  $h$  is hash function, then  $e$  is stored in position  $h(k)$  of table
- To search for  $e$ , compute  $h(k)$  to locate position. If no element/item, dictionary does not contain  $e$ .

# A simple Hash Table

- The simplest kind of hash table is an array of records (elements).
- This example array has 701 cells.



An array of cells

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# Following the example

- We want to store a dictionary of Object Records, no more than 701 objects
- Keys are Object ID numbers, e.g., 506643548
- Hash function:  $h(k)$  maps  $k(=ID)$  into distinct table positions 0-700
- Operations: insert, delete, and search

# Complexity (ideal case)

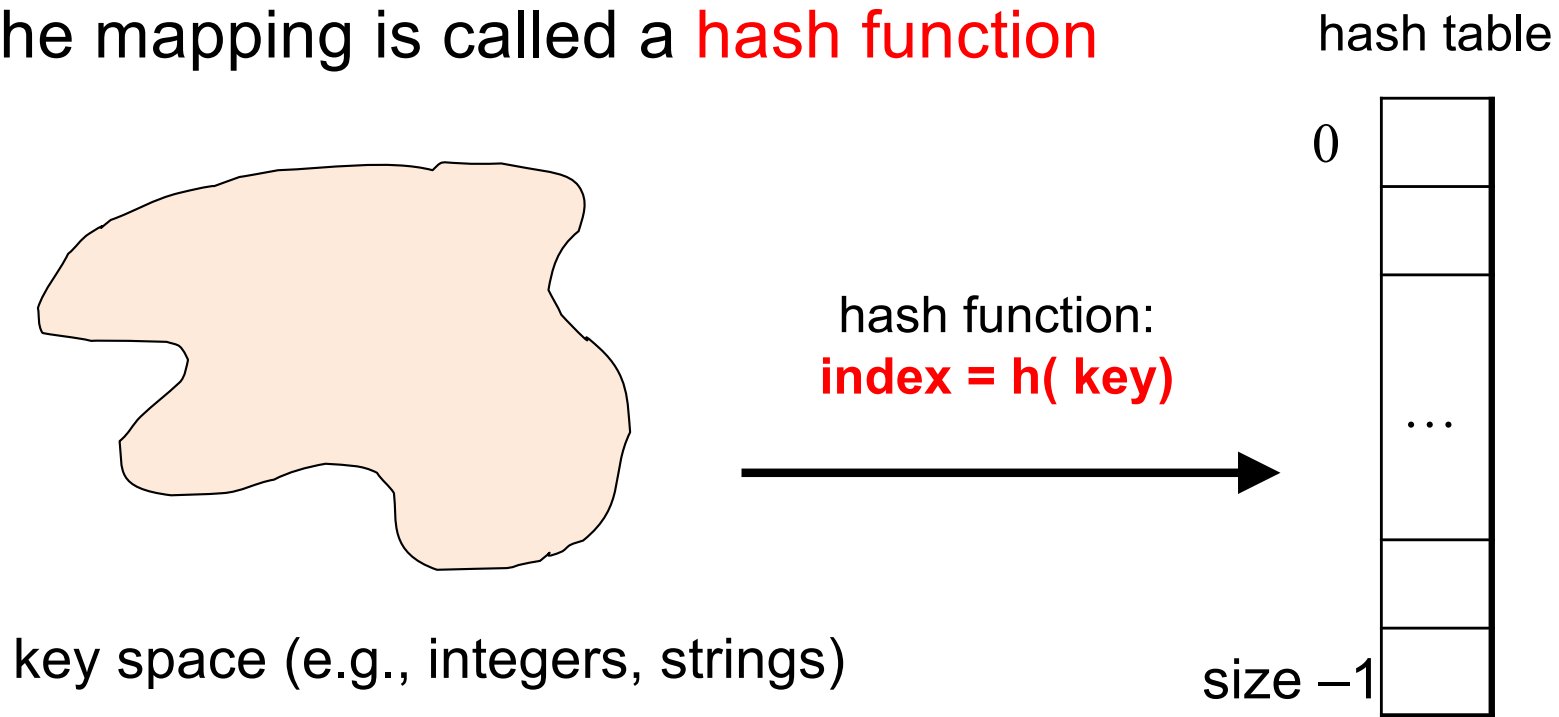
- Why hashing and hash table?
  - ▣ It is very efficient.
- $O(D)$  time to initialize hash table ( $D$  number of positions or cells in hash table)
- $O(1)$  time to perform *insert*, *remove*, *search*

# Limitations of the Simple Hash Table

1. The maximum number in array  $A$  must be  $\leq c \cdot n$ , where  $c$  is constant
  - Otherwise, the hash table may be too large
2. Keys must be integers
  - But may be strings, real numbers, etc. in a real application

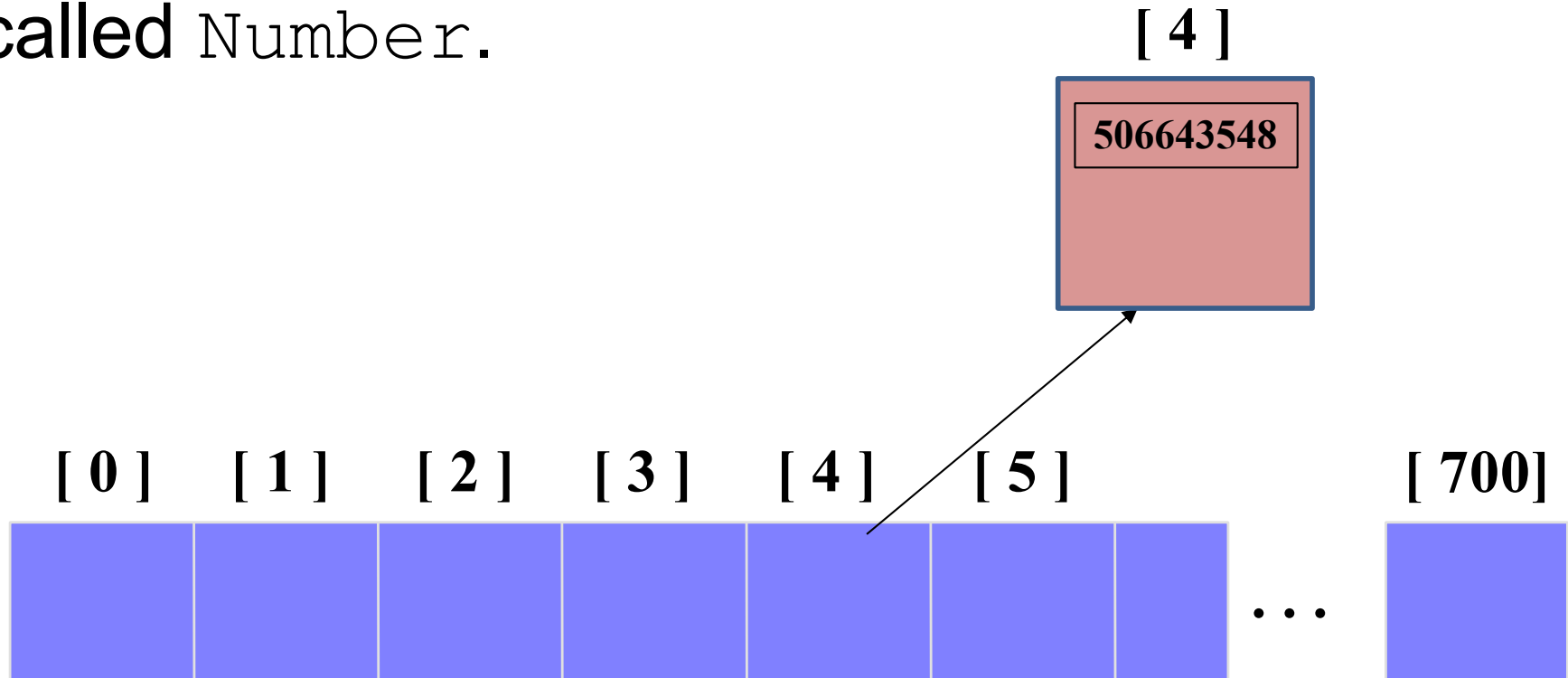
# Hash Table and Hash Function

- A **hash table** is an array of some fixed size, storing the records
- Each **key** is mapped into some *location* in the range 0 to size-1 in the table
- The mapping is called a **hash function**



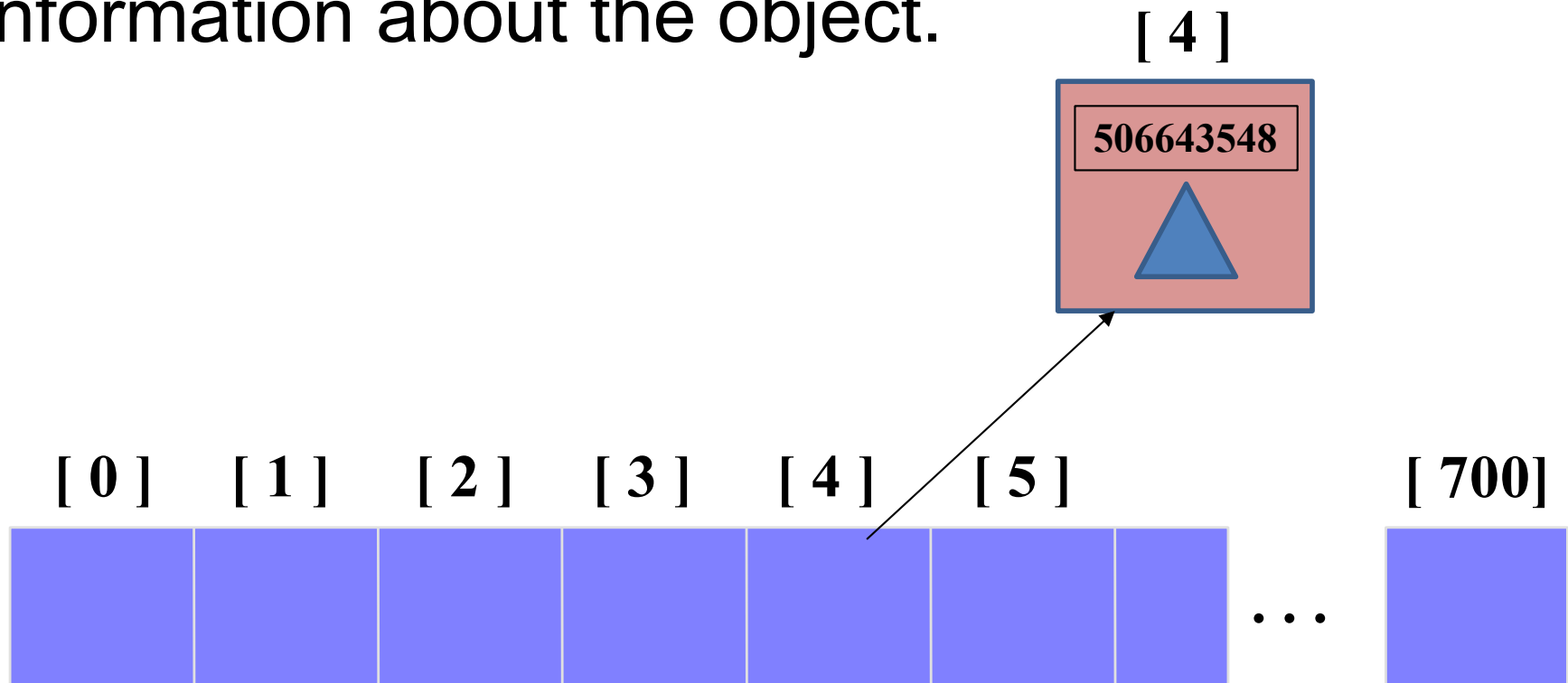
# Use the Hash Table

- Each record has a special field, i.e., its key.
- In this example, the key is a long integer field called `Number`.



# Use the Hash Table

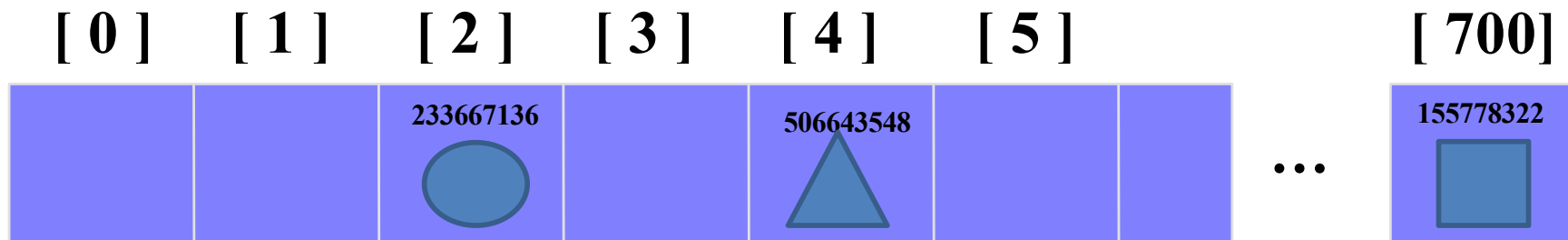
- The number is a object's identification number, and the rest of the record has information about the object.





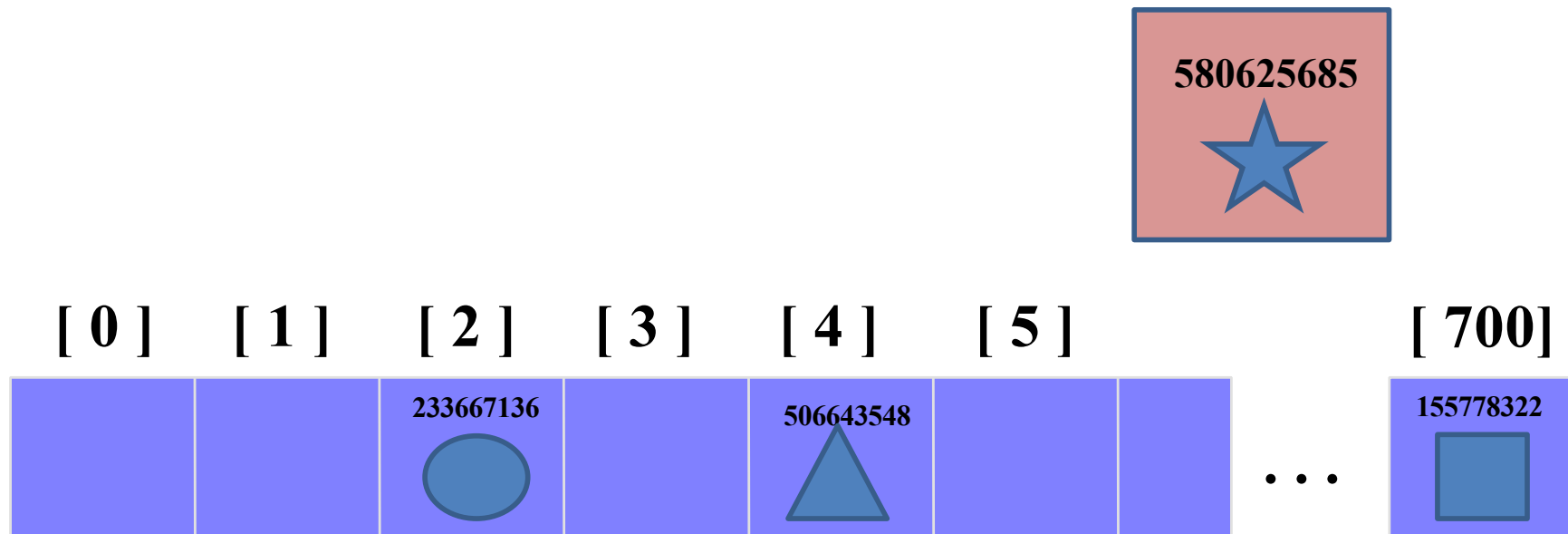
# Use the Hash Table

- When a hash table is in use, some spots contain valid records, and other spots are "empty".



# Inserting a New Record

- In order to insert a new record, the **key** must somehow be **mapped to** an array **index** using a **hash function**.
- The index is called the **hash value** of the key.



# Three design considerations of hash

1. Design a **general hash function**  $h(K)$  that maps a record with **key**  $K$  to a **location** in hash table
2. Given any two records with keys  $K_1$  and  $K_2$ , the **probability** that they are **mapped** to the **same location** in the hash table should be as **small** as possible, i.e.,
  - $\Pr[h(K_1) = h(K_2)]$  is very small
  - Otherwise, many records are mapped to the same location, which is called a **collision**
3. Solve the collision problem

# Hash functions

- Popular hash functions: hashing by division  
 $h(k) = k \bmod D$ , where  $D$  is number of cells in hash table

- Example: hash table with 701 cells

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

$$h(80) = 80 \bmod 701 = 80$$

$$h(1000) = 1000 \bmod 701 = 299$$

# Hash Function design – a simple mod function

- Consider  $n=5$  keys
  - $A[5]=11, 35, 54, 99, 42$
- Allocate an array  $Table[10]$  with size  $M=10$
- Hash function  $h(key) = key \% 10$
- Place 11 at location  $11\%10=1$  in hash table

	11	42		54	35				99	
index	0	1	2	3	4	5	6	7	8	9

- But there may be many collisions
- Consider other 5 keys
  - $B[5]=11, 21, 31, 41, 51$
  - Each key is mapped to location 1

# Hash Function design – a better hash function

- Consider  $n=5$  keys
  - $B[5]=11, 21, 31, 41, 51$
- Allocate an array **Table[10]** with size  $M=10$
- Hash function by **mid-square**, given a key  $K$ ,
  - Location is the middle  $r$  digits of value  $K^2$
  - $11^2=121, 21^2=441, 31^2=961, 41^2=1681, 51^2=2601$
  - Consider the middle digit, i.e.,  $r=1$

	51		11		21		31		41	
index	0	1	2	3	4	5	6	7	8	9

- The location is correlated with **all digits** in the key, not just the lowest digit.

# Hash function for a string-A simple way

- Given a string of characters, e.g. "AZ"
- First consider the ASCII value of each character
  - E.g., 65 for "A", 90 for "Z"
- Then, sum up the ASCII values of the characters
  - E.g.,  $65 + 90 = 155$
- Finally, mod M, where M is the size of the hash table
  - E.g.,  $155 \% 10 = 5$ ;
- String "AZ" is mapped to location 5 in the hash table

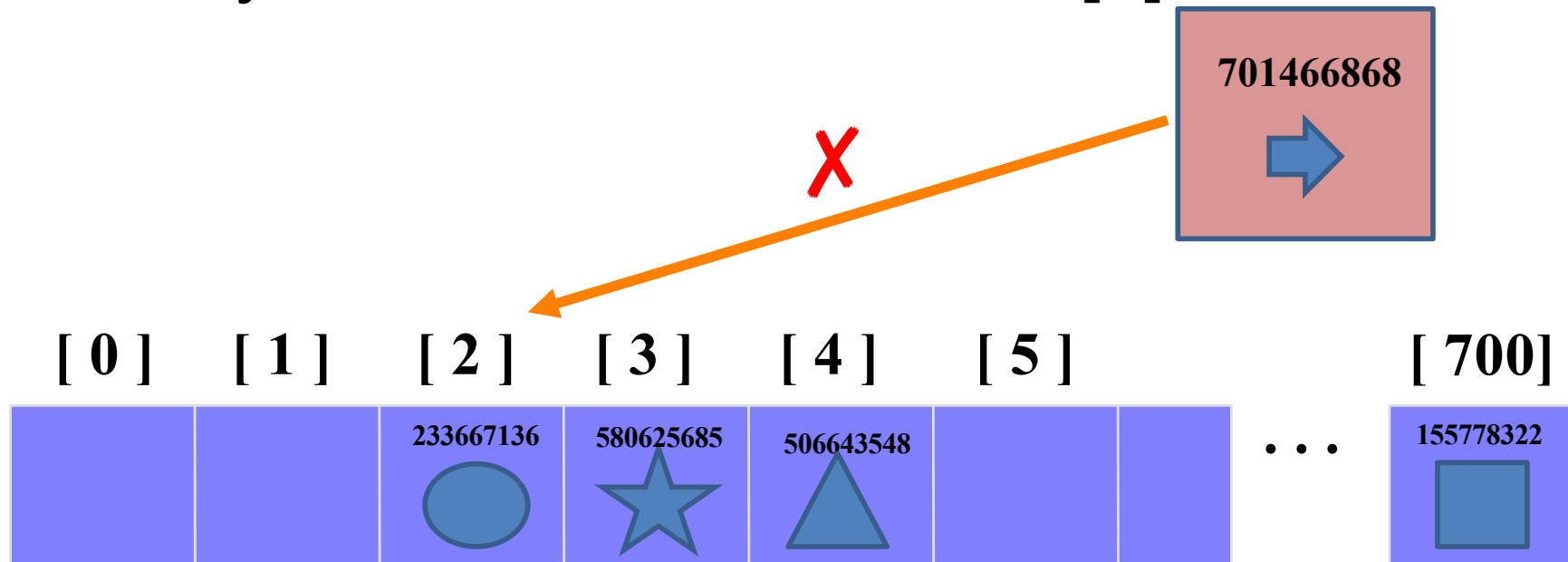
# Collisions

- Problem: **collision**
  - two **keys** may be mapped to the **same location**
  - Can we ensure that any two distinct keys get different locations?
    - **No**, if the size of the **key space** is **larger** than the size of the **hash table**



# Collisions - example

- Suppose we insert a new record, with a hash value of 2.
- This is called a **collision**, because there is already another valid record at [2].



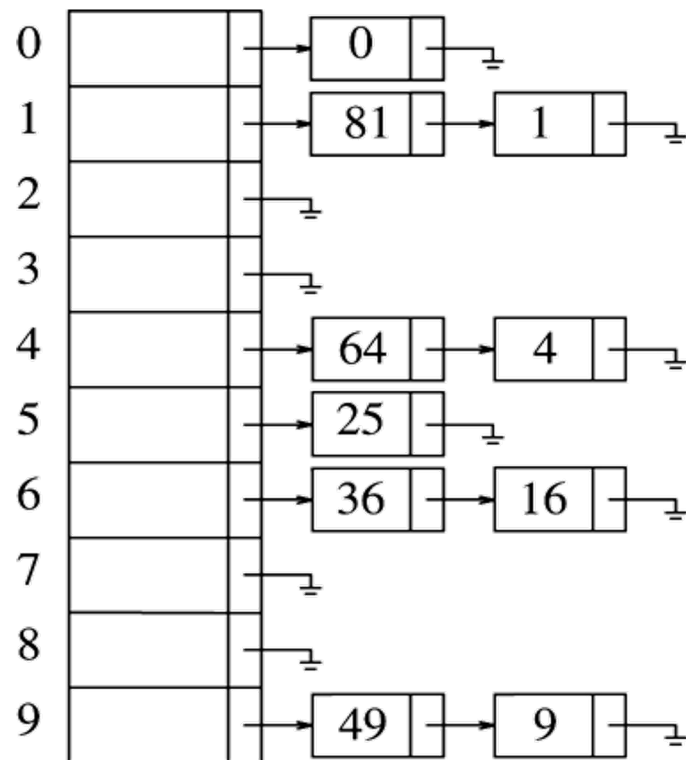
# Collision Resolution Techniques

- Two strategies:
  - ▣ (1) **Open hashing**, a.k.a. separate chaining
  - ▣ (2) **Closed hashing**, a.k.a. open addressing
- Difference has to do with whether collisions are stored *outside the table* (**open hashing**) or whether collisions result in storing one of the records at *another slot in the table* (**closed hashing**).

# Open hashing / Separate Chaining

- Instead of a hash table, use a table of **linked list**
- keep **a linked list** of records with **keys** mapped to the **same location**

$$h(K) = K \bmod 10$$



# Separate Chaining (cont.)

- To search a record with key  $K$ 
  - Calculate  $h(K)$ , takes  $\Theta(1)$  time
  - Search the linked list at  $table[h[K]]$ , which takes  $\Theta(d)$  time,  $d$  is the list size
- Average list size  $\alpha = \frac{n}{m}$ ,  $n$ : # of records,  $m$ : hash table size
- Searching time is  $\Theta(1+\alpha)$  on average

# Improve performance of separate chaining

- Searching time is  $\Theta(1+\alpha)$  on average
- $\alpha = \frac{n}{m}$  usually is called the **load factor**
- When the  $\alpha$  exceeds a **threshold**, e.g. **1.5**, double the table size
- **Rehash** each record in the **old** table into the **new table**
- Then, the value of  $\alpha$  decreases
- Searching time is  $\Theta(1+\alpha) = \Theta(1)$  on average

# Separate Chaining (cont.)

- **Advantage:** implementation is easy for inserting, searching, and deleting
- **Disadvantage:** memory allocation for a new node will slow down the program

# Closed hashing / Probing hash tables

- Basic Idea:
  - To insert a key **K**, compute  **$h(K)$** . If location  **$h(K)$**  is **empty**, insert it there
  - If a **collision** occurs, probe **alternative locations**  **$h_1(K)$ ,  $h_2(K)$ , ...**, until an **empty** location is found
- **$h_i(K) = (h(K) + f(i)) \% TableSize$** ,
  - **$f(.)$ : *collision resolution strategy***
- **All data** are stored inside the table, hash **table size** must be **larger** than the **number of records**
  - ***i.e.,  $m \geq n$***
  - Otherwise, no alternative locations can be found

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# Probing hash tables

- Three approaches
  - Linear Probing
  - Quadratic Probing
  - Double Hashing



# Solution 1: Linear Probing

- $f$  is a **linear** function of  $i$ : i.e.,  $f(i)=i$ 
  - Locations are probed **sequentially**
  - $h_i(K) = (h(K) + i) \% \text{TableSize}$
- **Insertion:**
  - Let **K** be a new key to be inserted, compute  **$h(K)$  first**
  - For  $i = 0$  to  $\text{TableSize}-1$ 
    - compute  **$L = (h(K) + i) \% \text{TableSize}$**
    - **$\text{Table}[L]$**  is empty, then we put **K** there and **stop**.

# Example of linear probing

- $h_i(K) = (h(K) + i) \% m$ 
  - ▣ E.g, inserting keys 89, 18, 49, 58, 69 with  $h(K)=K \% 10$
- A **clustering** problem: **small clusters** grow to **big** clusters

	Empty Table	After 89	After 18	After 49	After 58	After 69
0				49	49	49
1					58	58
2						69
3						
4						
5						
6						
7						
8			18	18	18	18
9		89	89	89	89	89

To insert 49,  
probe T[9], T[0]

To insert 58,  
probe T[8], T[9],  
T[0], T[1]

To insert 69,  
probe T[9], T[0],  
T[1], T[2]

# Solution 2: Quadratic Probing

- $f(i) = i^2$
- $h_i(K) = (h(K) + i^2) \% \text{TableSize}$ , e.g.,  $h(K) = K \% 10$ 
  - E.g., inserting keys 89, 18, 49, 58, 69

	Empty Table	After 89	After 18	After 49	After 58	After 69
0				49	49	49
1						
2					58	58
3						69
4						
5						
6						
7						
8			18	18	18	18
9		89	89	89	89	89

To insert 49,  
probe  $T[9]$ ,  $T[0]$

To insert 58,  
probe  $T[8]$ ,  $T[9]$ ,  
 $T[(8+2^2) \bmod 10]$

To insert 69,  
probe  $T[9]$ ,  
 $T[(9+1) \bmod 10]$ ,  
 $T[(9+2^2) \bmod 10]$

# Quadratic Probing

- Two **keys** with different **initial** hash **locations** will have different probe sequences
  - $h(k_1)=30$ ,  $h(k_2)=29$ , with **difference only one**
  - probe sequence for  $k_1$ : **30, 31, 34, 39, ...**
  - probe sequence for  $k_2$ : **29, 30, 33, 38, ...**
- If the table size  **$m$**  is **prime**, then a new key can always be inserted if the table is **at least half empty**

# Solution 3: Double Hashing

- Use two hash functions:  $h()$  and  $h2()$
- $f(i) = i * h2(K)$
- $h_i(K) = (h(K) + f(i)) \% m$ 
  - E.g.  $h2(K) = R - (K \bmod R)$ , with  $R$  is a prime smaller than  $m$
- The probe sequence  $f(1), f(2), \dots$  is independent of its initial location  $h(K)$

# Double Hashing

- $h_i(K) = (h(K) + f(i)) \% m$ ;  $h(K) = K \% m$
- $f(i) = i * h2(K)$ ;  $h2(K) = R - (K \bmod R)$ ,
- **Example:**  $m=10$ ,  $R = 7$  and insert keys 89, 18, 49, 58, 69

	Empty Table	After 89	After 18	After 49	After 58	After 69
0						69
1						
2						
3					58	58
4						
5						
6				49	49	49
7						
8			18	18	18	18
9		89	89	89	89	89

To insert 49,  
 $h2(49)=7$ , 2<sup>nd</sup>  
 probe is  $T[(9+7)$   
 $\bmod 10]$

To insert 58,  
 $h2(58)=5$ , 2<sup>nd</sup>  
 probe is  $T[(8+5)$   
 $\bmod 10]$

To insert 69,  
 $h2(69)=1$ , 2<sup>nd</sup>  
 probe is  $T[(9+1)$   
 $\bmod 10]$

# Choice of hash function $h_2()$

- $h_2(K)$  cannot be 0, as  $i \cdot 0 = 0$
- For any key  $K$ ,  $h_2(K)$  must be **relatively prime** to the table size  $m$ . Otherwise, we may probe only **a fraction** of the table entries.
  - e.g., if  $h(K)=0$  and  $h_2(K) = m/2$ , ( $m$  is **even**), then we will only examine entries  $Table[0]$ ,  $Table[m/2]$ , and nothing else!
- One solution is to make  $m$  **prime**, and choose  $R$  to be a prime **smaller than  $m$** , and set
$$h_2(K) = R - (K \bmod R)$$
- Quadratic probing, however, does not require the use of a second hash function
  - likely to be **simpler** and **faster** in practice

# The performance of probing hash tables

- Load factor  $\alpha = \frac{n}{m} \leq 1$  as  $n \leq m$
- Collision probability is  $\alpha$  for each probe
- Insert successfully at 1<sup>st</sup> probe with probability  $1-\alpha$
- Insert successfully at 2<sup>nd</sup> probe with prob.  $\alpha(1-\alpha)$
- Insert successfully at 3<sup>rd</sup> probe with prob.  $\alpha^2(1-\alpha)$
- ...
- Insert successfully at  $k^{\text{th}}$  probe with prob.  $\alpha^{k-1}(1-\alpha)$
- Average probe times are  $\frac{1}{1-\alpha}$
- Insert and search average time is  $\Theta\left(\frac{1}{1-\alpha}\right) = \Theta(1)$  if  $\alpha$  is small, e.g.,  $\alpha=0.5$



# Performance Comparison ( $n=400M$ )

- Sequential search : 200 ms
- Binary search: 0.002 ms
- Hash search: < 0.001 ms

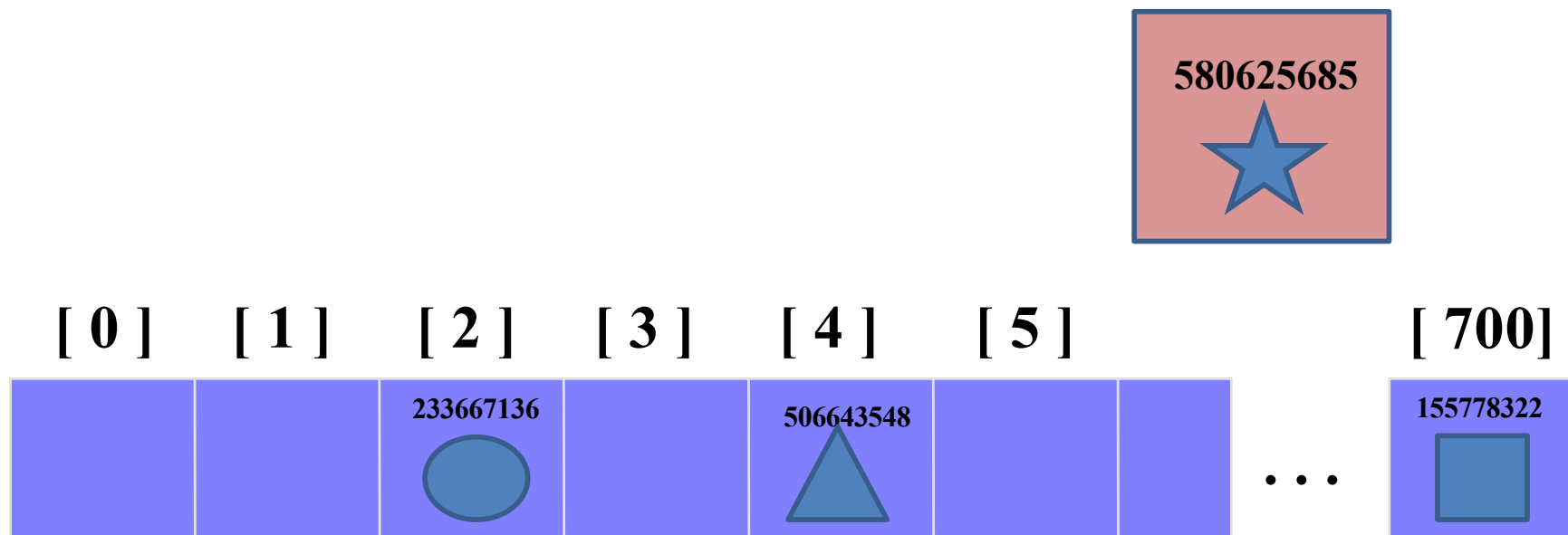
# Insert

- Apply hash function to **get a location**
- Try to insert key at the location
- Deal with **collision**

# Inserting a New Record

- Let us find the **hash value** for 580625685

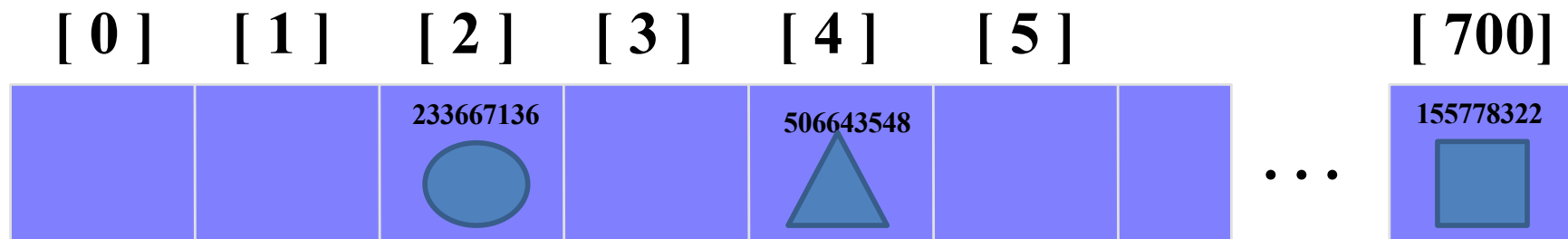
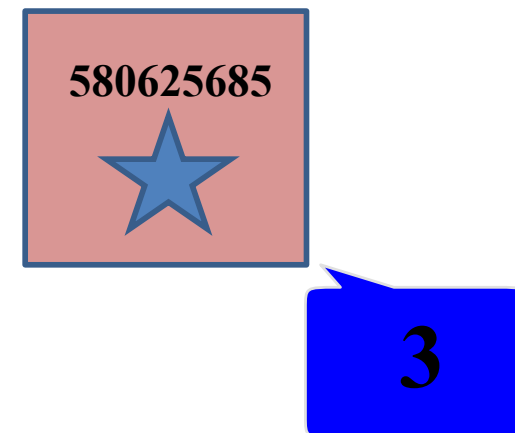
What is  $(580625685 \bmod 701)$  ?



# Inserting a New Record

- Let us find the **hash value** for 580625685

$$580625685 \bmod 701 = 3$$



# Inserting a New Record

The hash value is used to find the location of the new record.



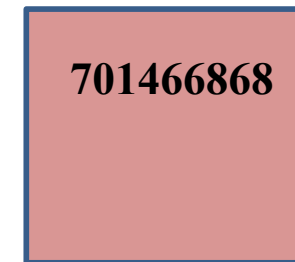
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# Search

- Apply the **hash function** to get a location
- Look at that location.
- Deal with collision.

# Searching for a Key

- The data that's attached to a key can be found fairly quickly.



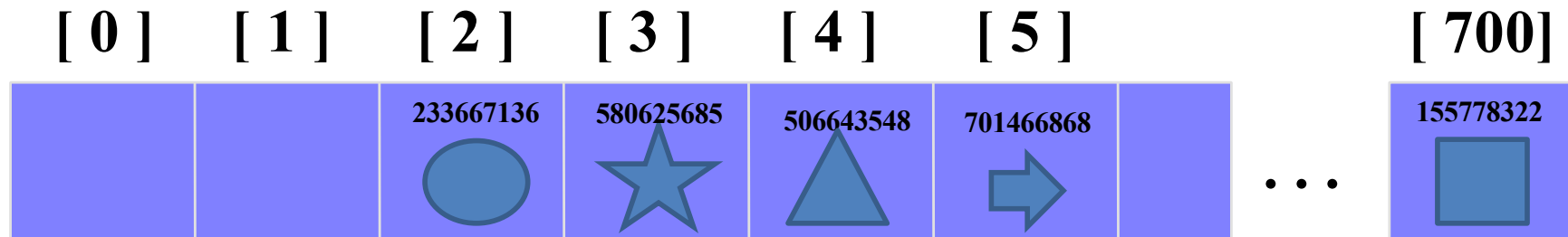
# Searching for a Key

- Calculate the hash value.
- Check that location of the array for the key.

701466868

The hash value of  
701466868 is 2

**Not me.**





# Searching for a Key

- Keep moving forward until you find the key, or you reach an empty spot.

701466868

The hash value of  
701466868 is 2

Not me.



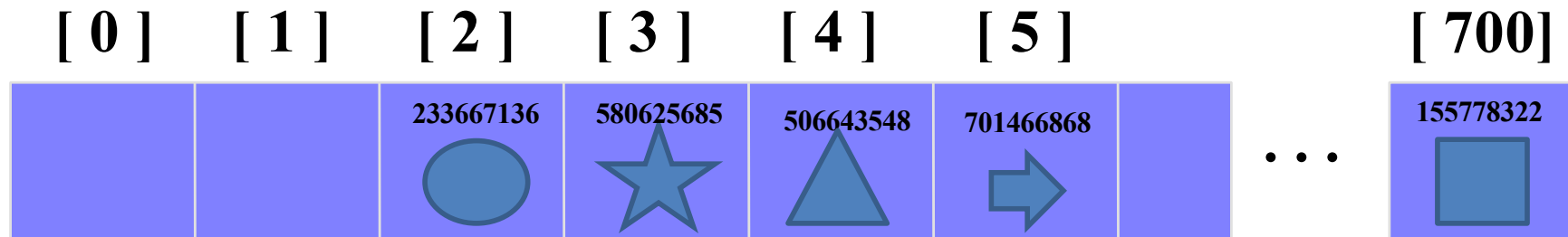
# Searching for a Key

- Keep moving forward until you find the key, or you reach an empty spot.

701466868

The hash value of  
701466868 is 2

**Not me.**



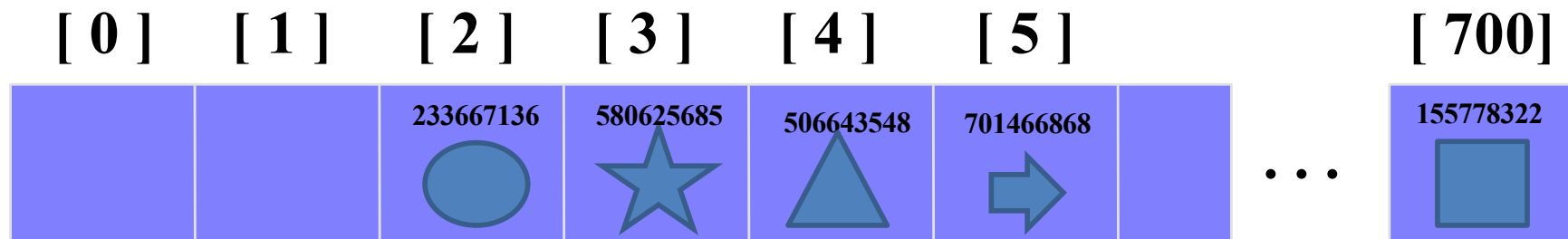
# Searching for a Key

- Keep moving forward until you find the key, or you reach an empty spot.

701466868

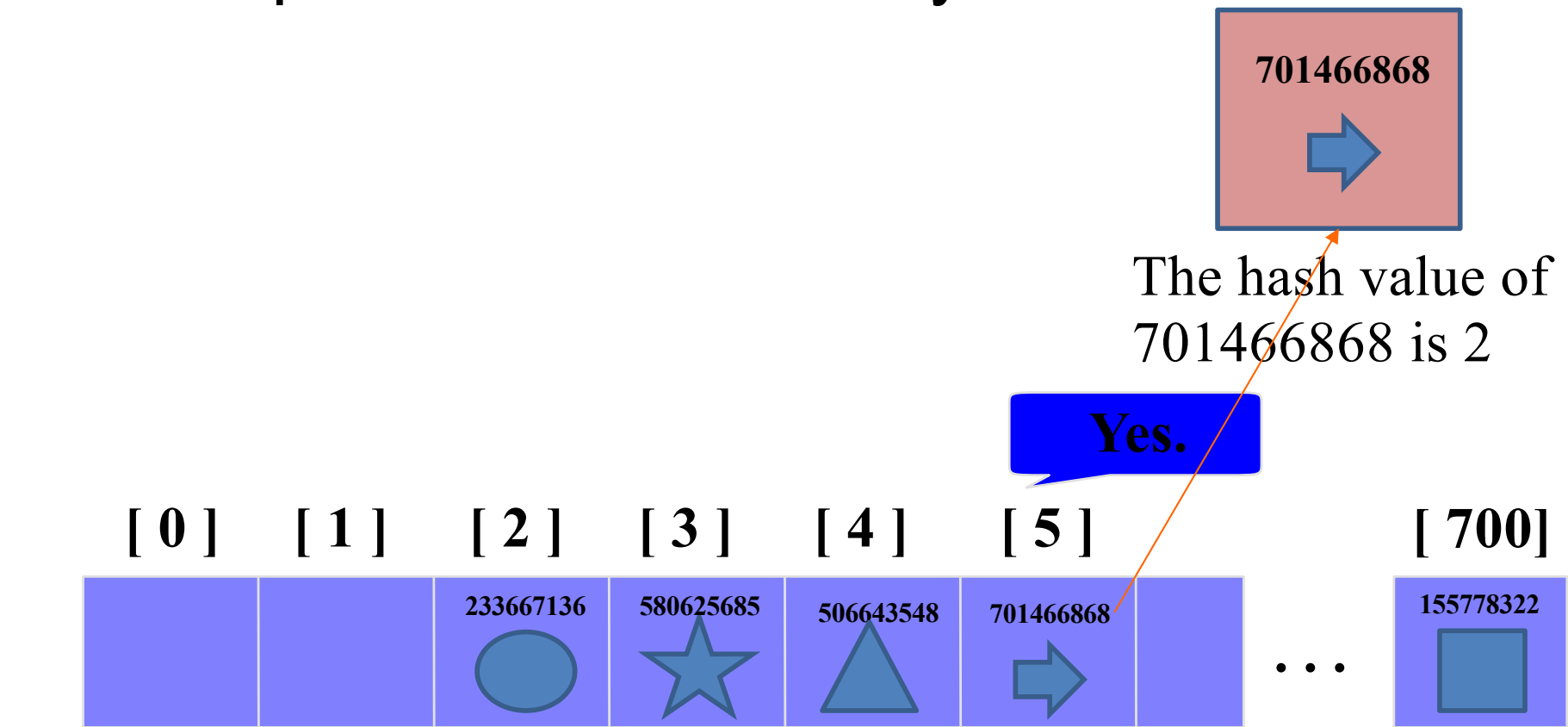
The hash value of  
701466868 is 2

Yes.



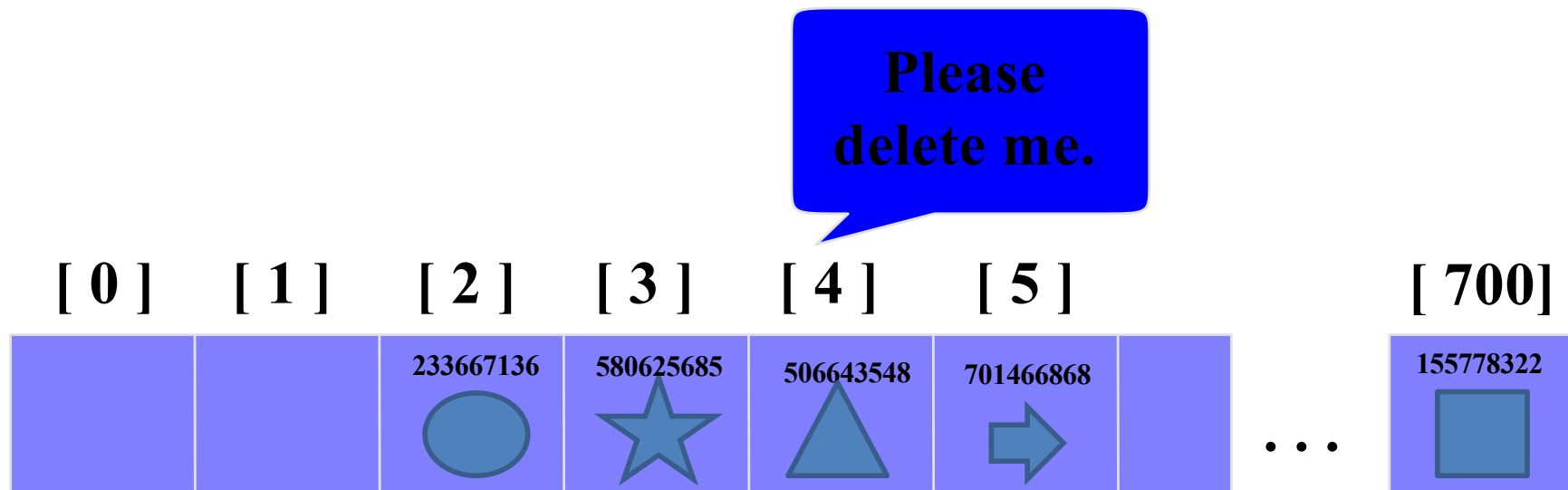
# Searching for a Key

- When the item is found, the information can be copied to the necessary location.



# Deleting a Record

- Records may also be deleted from a hash table.



# Deleting a Record

- Records may also be deleted from a hash table.
- But the location must not be left as an ordinary "empty spot" since that could interfere with searches.



# Deleting a Record

- Records may also be deleted from a hash table.
- But the location must not be left as an ordinary "empty spot" since that could interfere with searches.
- The location must be marked in some special way so that a search can tell that the spot used to have something in it.



# Conclusions

- **Sequential search:**  $\Theta(n)$  on average
  - improve to  $\Theta(\sqrt{n})$  by Jump search with ordered arrays
- **Binary search:**  $\Theta(\log n)$
- Self-organizing lists, and Bit vectors
- **Hashing:**  $\Theta(1)$  on average
  - Hash table size usually is prime.
  - Hash functions
    - mod function, mid-square, sum for strings
  - Collision solutions
    1. Separating chaining
    2. Probing hash tables
      - linear probing, quadratic probing, double hashing