



**CSCI2100B**  
**CSCI2100S** DATA STRUCTURES

.....

Spring 2011

Hash Tables

# Introduction

---

- Many applications require a dynamic set that supports only the **dictionary** operations insert, search and delete
  - The set is often called a **symbol table** (ST)
- A hash table is an **effective** data structure for implementing dictionaries.
- Though searching an element in a hash table can take as long as  **$\Theta(N)$**  in the **worst** case, the **expected** time to search for an element is only  **$O(1)$** .

# Simple Example of Hashing

---

We want to build a symbol table for the list of words

onion	potato	salt
tomato	mushroom	chicken

To convert the characters into array indices, encode the characters as follows:

$$a = 1, b = 2, \dots, z = 26$$

Then sum up the values in each word

$$h(\text{onion}) = 15 + 14 + 9 + 15 + 14 = 67$$

$$h(\text{chicken}) = 53$$

$$h(\text{potato}) = 16 + 15 + 20 + 1 + 20 + 15 = 87$$

$$h(\text{tomato}) = 84$$

$$h(\text{salt}) = 19 + 1 + 12 + 20 = 52$$

$$h(\text{mushroom}) = 122$$

# Simple Example of Hashing

---

Divide all sums by 6 and use the remainder as the array index:

$$h(\text{onion}) \bmod 6 = 67 \bmod 6 = 1$$

$$h(\text{potato}) \bmod 6 = 87 \bmod 6 = 3$$

$$h(\text{salt}) \bmod 6 = 52 \bmod 6 = 4$$

$$h(\text{tomato}) \bmod 6 = 84 \bmod 6 = 0$$

$$h(\text{mushroom}) \bmod 6 = 122 \bmod 6 = 2$$

$$h(\text{chicken}) \bmod 6 = 53 \bmod 6 = 5$$

tomato

onion

mushroom

potato

salt

chicken

To check whether a word belongs to the symbol table:

***Identify the slot by first summing the character values and then modulo 6.***

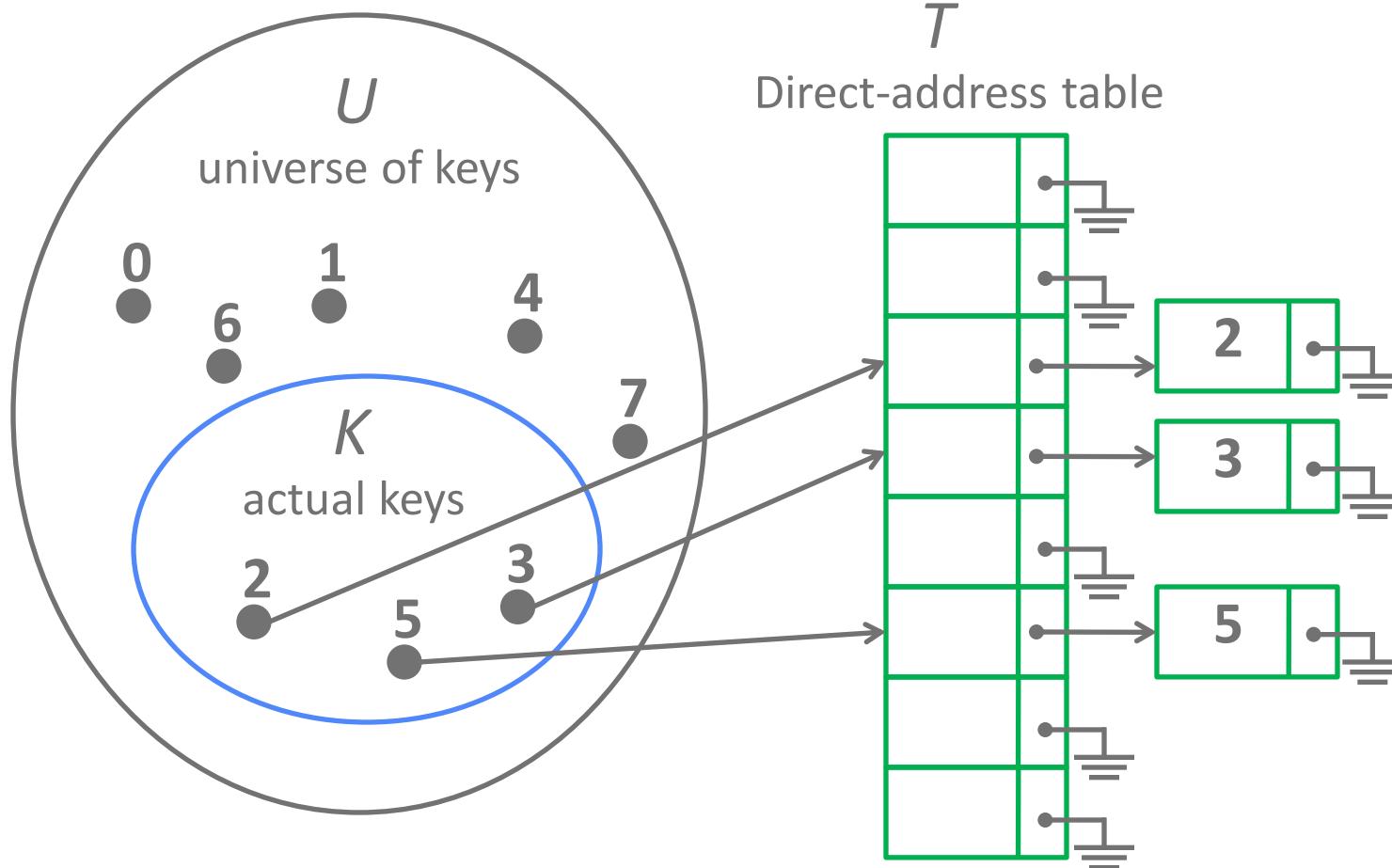
If it is not empty, the word is present.

# Direct-Address Tables

---

- Direct addressing is a simple technique that works well when the universe  $U$  (the complete set) of keys is reasonably **small**.
- Suppose that an application needs a **dynamic** set in which each element has a **key** drawn from  $U = \{0, 1, \dots, M - 1\}$  and  $M$  is not too large.
- If we assume no two elements have the same key, then we can use an array, or **direct-address table**, in which each position (slot) corresponds to a key in  $U$ .

# Direct-Address Tables: Example



# Dictionary Operations

---

- If we use an array with size  $M$ , then
  - *search*: return  $T[k]$
  - *insert*:  $T[\text{key}(x)] = x$
  - *delete*:  $T[\text{key}(x)] = \text{NULL}$
- Each of the above operations is fast:  $O(1)$
- **Storage** of the elements:
  1. with the direct-address table itself
  2. externally with a pointer from a slot in the table

# Hash Tables

---

- The difficulty with direct addressing is obvious: if the universe  $U$  is large, storage a table of size  $|U|$  is impractical.
  - e.g. ASCII string with 10 chars:  $26^{10}$
- The set  $K$  of keys actually stored may be so small relative to  $U$ .
  - e.g. actually words with 10 chars
- When the set  $K$  is small, a hash table requires much less storage than a direct address table. Memory required:  $\Theta(|K|)$

# Hash Function

---

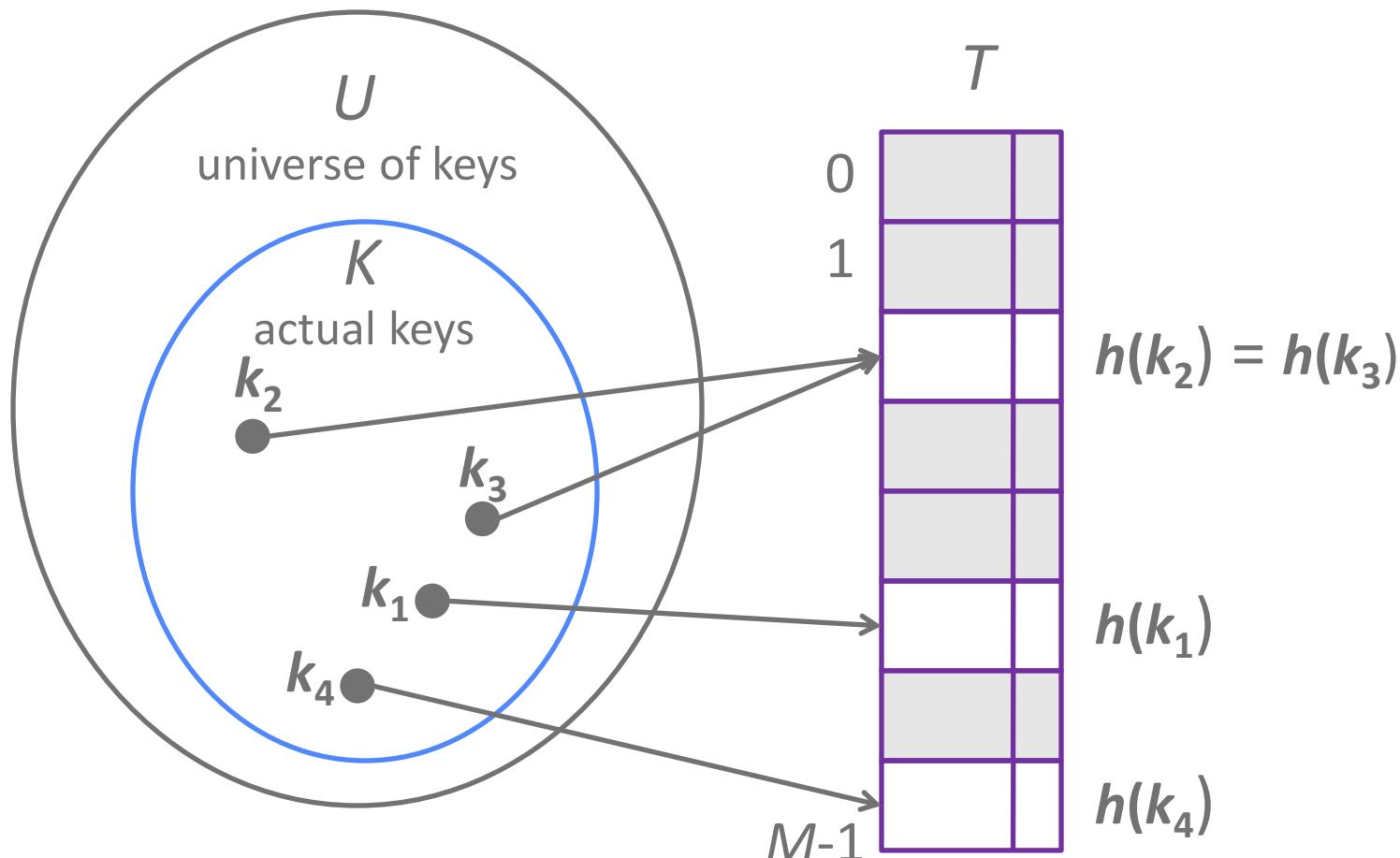
- With hashing, an element with key  $k$  is stored in slot  $h(k)$ .
- We use a hash function to compute the slot from a key  $k$ .
  - Let  $M$  be the size of the hash table, then
$$h: U \rightarrow \{0, 1, \dots, M - 1\}$$
- We say that an element with key  $k$  **hashes** to slot  $h(k)$  while  $h(k)$  is the **hash value** of key  $k$ .
- Hash function **reduces** the range of indices needed to be handled:  $M$  instead of  $|U|$ .

# Hash Function (2)

---

- The performance of the hashing depends on **how well** the hash function **distribute** the keys to the  $m$  slots.
- **Simple uniform hashing:** any given element is **equally likely** to be hash into any of the  $M$  slots.
- **Hitch**   
**Collisions:** two keys may hash to the **same** slot
  - We need to find methods to handle collisions – **collision solution** (handling).

# Hash Tables: Example



# Choosing a Hash Function

---

- The hash function  $h$  should be:
  - be **easy** and **quick** to be computed.
  - achieve an **even** distribution of the keys.
  - **deterministic** that a given input  $k$  always produce same output  $h(k)$ .
- It is a good idea to ensure that the table size is **prime**. **WHY?**

## Hash Function for Integers

If the input keys are integers, then simply returning  $K \bmod M$  is generally a reasonable strategy.

When the keys are random integers, this function is simple to compute and can distribute keys evenly.

*This applies to data that interpreted as integers  
(e.g. floating point numbers considered as bytes)*

# A Simple Hash Function for Strings

---

- Strings are common keys as well.
- A Simple hash function for strings:  
add up the ASCII values of the characters and mod the sum.

```
unsigned int hash(char *key, unsigned int m){  
    unsigned int hash_val = 0;  
    while (*key != '\0')  
        hash_val += *key++;  
    return (hash_val % m);  
}
```

 **Problem:** this hash function **does not hash evenly**.

Consider strings shorter than 8 in a hash table with  $M = 10007$ ,  
the function can only return values up to  **$127 \times 8 = 1016$** .

# A Better Hash Function

---

- Examines the three letters of the keys
- The magic number **27** comes from the total number of English alphabets and the space (you may have to adjust it).
- If all 3 letters are **random**, then we have  $26^3 = 17,576$  combinations that can fit quite well on a table with  $M = 10,007$ .
- Unfortunately, English words are **not** random. A large on-line dictionary reveals only 2,851 combinations with the first 3 letters.

```
unsigned int hash(char *key, unsigned int m){  
    return (key[0] + 27 * key[1] + 729 * key[2]) % m;  
}
```

# A Good Hash Function

- This hash function includes all characters in the key.
- It computes  $\sum_{i=0}^{N-1} k[N - i - 1] \times 27^i \Rightarrow \sum_{i=0}^{N-1} k[N - i - 1] \times 32^i$  where  $N$  is the length of the key.
- It computes the polynomial function using **Horner's rule**.
- **More efficiency**: the multiplication of 32 can be achieved by shifting 5 bits to the left.

```
unsigned int hash(char *key, unsigned int m){  
    unsigned int hash_value = 0;  
    while (*key != '\0')  
        hash_val = (hash_val << 5) + *key++;  
    return (hash_val % m);  
}
```

Horner's rule:

$$a_0 + a_1x + a_2x^2 + \cdots + a_nx^n = a_0 + x(a_1 + x(a_2 + \cdots + x(a_{n-1} + a_nx) \cdots))$$

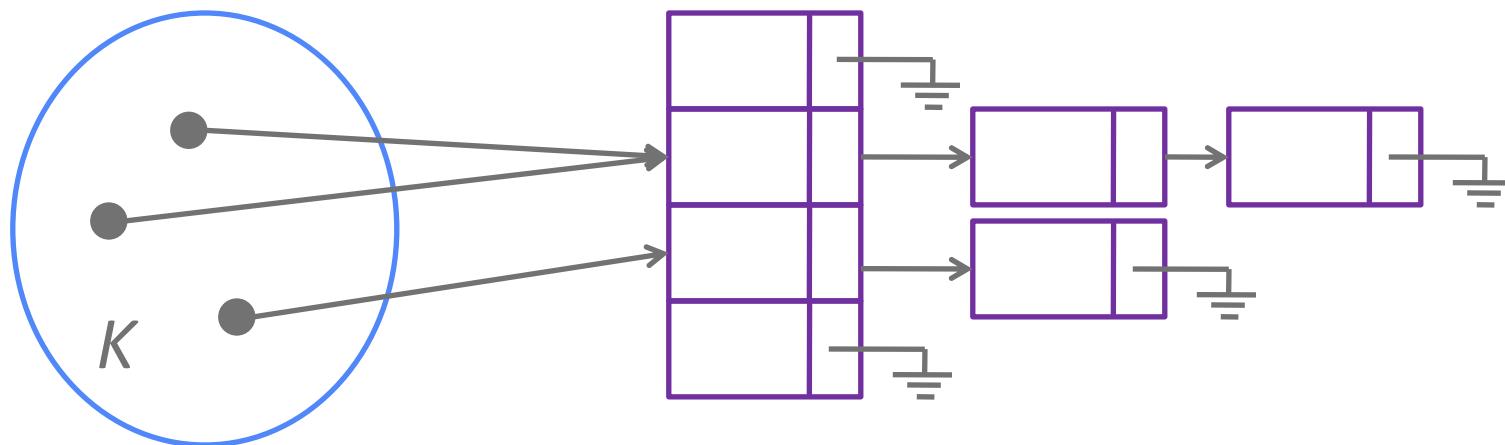
# More About Hash Functions

---

- It is common not to use all characters – the length and properties of the keys would influence the choice.
- The hash function might include a couple characters from each field.
- **Truncation:** ignore part of the key and use the remaining part directly as the index.
  - e.g. 62,538,194 hash to 394 by taking 1st, 2nd and 5th digits.
- **Folding:** partition the key into several parts and combine parts in a **convenient** way.
  - e.g. 62,538,194 maps to  $625 + 381 + 94 = 1100$  and is truncated to 100.

# Collision Resolution by Chaining

- In **chaining**, we put all elements that hash to the same slot in a **linked list** (or normal list).
- Slot  $j$  contains a **header** node of the list that stores **ALL** elements that are hashed to  $j$ .
  - contains a **NULL** link if it is empty.



# Hash Table: Implementation

---

- Suppose we reuse our list ADT in the implementation of the hash table, the table is actually an array of lists.

```
typedef struct hashtable_s {
    int m;                      /* hash table size */
    list_t *slots;               /* array of lists */
} hashtable_t;
```

```
hashtable_t *hash_init_tbl(int m){
    int i;
    hashtable_t *ht = (hashtable_t *)malloc(sizeof(hashtable_t));
    ht->m = next_prime(m);
    ht->slots = (list_t *)malloc(ht->m * sizeof(list_t));
    for (i = 0; i < ht->m; i++)
        ht->slots[i] = list_create();

    return ht;
}
```

# Dictionary Operations: Insert

- $\text{insert}(T, x)$ : insert a new element  $x$  to the hash table  $T$  where the **key** of  $x$  is  $\text{key}(x)$
- We need to hash  $\text{key}(x)$  to get the list stored at the slot and insert  $x$  to the list.

```
void hash_insert(hashtbl_t *ht, int key){  
    list_t list = ht->slots[hash(key, ht->m)];  
    if (list_find(list, key))  
        return; /* key found: do nothing */  
    list_insert(list, list, key);  
}
```

Complexity =  $\Theta(1 + \alpha)$   
since list insertion is  $O(1)$ .

$\alpha$  is the average length of the list

# Dictionary Op.: Delete & Find

---

- The two operations are similar: first hash to find the slot, get the list and then continue with the basic list search/delete operations.
- Complexity =  $\Theta(1 + \alpha)$  *average length of the list*

```
void hash_delete(hashtable_t *ht, int key){  
    list_t list = ht->slots[hash(key, ht->m)];  
    list_delete(list, key);  
}
```

```
pos_t hash_find(hashtable_t *ht, int key){  
    list_t list = ht->slots[hash(key, ht->m)];  
    return list_find(list, key);  
}
```

# Analysis of Chaining

## Load Factor $\alpha$

Given a hash table  $T$  with  $M$  slots that stores  $N$  elements, we define the load factor  $\alpha = N/M$ .

- The load factor depends on how well the hash function distributes the keys among  $m$  slots:
  - **Worse-case:**  
Horrible. The hash table is the same as a linked list with all  $N$  elements hash to the same slot.  $\alpha = N$ .
  - **Average-case:**  
Assume simple uniform hashing, any given element is **equally likely** to hash into any of the  $M$  slots.  $\alpha = N/M$ .
- If we choose  $M$  s.t.  $N = O(M)$  then  $\alpha = O(1)$ . Both **find** and **delete** are  **$O(1)$**  on **average**.

# Average No. of Probes in Chaining

---

When chaining is used, the average number of probes required to search in a hash table of size  $M$  that contains  $N = \alpha M$  keys is about

$$\frac{1}{2} \left( 1 + \frac{1}{1 - \alpha} \right)$$

for hits

$$\frac{1}{2} \left( 1 + \frac{1}{(1 - \alpha)^2} \right)$$

for misses

load factor ( $\alpha$ )	1/2	2/3	3/4	9/10
search hit	1.5	2.0	3.0	5.5
search miss	2.5	5.0	8.5	55.5

Remarks:

*Chaining performs badly when the hash table is filling up.*

# Open Addressing

---

- In **open addressing**, all elements are stored in the hash table itself.
  - Each slot contains either **an element** or **NULL**.
- The hash table can ***fill up*** so that no further insertions can be made →  $\alpha$  can never exceed 1.
- **Advantage:** pointers are avoided.
- We computer the **sequence** of slots to be examined.
- ***insert*:** successively probe the hash table until we find an empty slot to save the key.
- Probing in fixed order  $0, 1, \dots, M - 1$  is **unwise** since it requires  **$\Theta(N)$**  search time.

# Open Addressing (Cont')

---

- The **probe sequence** depends upon the key being inserted.
- We extend the hash function to include the probe number as a **second input**.

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

- With open addressing, for every key  $k$ , the probe sequence is:

$$\langle h(k, 0), h(k, 1), \dots, h(k, m - 1) \rangle$$

as an permutation of  $\{0, 1, \dots, M - 1\}$

# Open Addressing: *insert*

---

- We assume the element only contains the key itself
- Each slot contains either a key or a value for NULL (**HASH\_NULL**).

```
unsigned int hash_insert(hashtbl_t *ht, int key){  
    int i;  
    unsigned int j;  
  
    for (i = 0; i != ht->m; i++){  
        j = oahash(key, i, ht->m);  
        if (ht->slots[j] == HASH_NULL){  
            ht->slots[j] = key;  
            return j;  
        }  
    }  
    perror("oahash.c: hash table overflow\n");  
    return HASH_NULL;  
}
```

# Open Addressing: *find*

---

- The search **terminates** when it finds an empty slot, since the **key** would have been inserted there.
- **Assumption:**  
the key is **not** deleted from the table once before

```
unsigned int hash_find(hashtbl_t *ht, int key){  
    int i = 0;  
    unsigned int j;  
  
    do {  
        j = oahash(key, i, ht->m);  
        if (ht->slots[j] == key)  
            return j;  
    } while (ht->slots[j] != HASH_NULL && i++ != ht->m);  
  
    return HASH_NULL;  
}
```

# Open Addressing: *delete*

---

- Delete from an open-addressing hash table is **difficult**.
- When we delete a key from slot  $i$ , we cannot simply mark that slot as empty by storing a NULL.
  - then we fail to retrieve any key whose probe sequence contains slot  $i$ .
- **Solution:** mark the slot deleted instead. (you need to modify insert accordingly)
  - **Drawback:** the search time is no longer dependent on the load factor  $\alpha$
- When keys have to be deleted, **chaining** is more preferred.

# Linear Probing

---

- Given an ordinary hash function  $h'$ , the method of linear probing uses the hash function

$$h(k, i) = (h'(k) + i) \bmod M$$

- If the slot  $T[h'(k)]$  is occupied, we continue to probe the next one  $T[h'(k) + 1]$  and so on, until we have an empty slot.
- Easy to implement, but suffers from **primary clustering**: long runs of occupied slots build up, increasing average search time.

```
static unsigned int oahash(int key, unsigned int i, unsigned int m){  
    return (hash(key, m) + i) % m;  
}
```



# Linear Probing: Example

- $h'(k) = k \bmod 10, M = 10$
- Insert 89, 18, 49, 58, 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

# Quadratic Probing

---

- Quadratic probing uses a hash function of the form

$$h(k, i) = (h'(k) + c_1 i + c_2 i^2) \bmod B$$

- The initial position probed is  $T[h'(k)]$ ; late positions probe are offset by an amount quadratic to the probe number  $i$ .
- Works much better than linear probing but parameters  $c_1$ ,  $c_2$  and  $m$  have to be selected carefully to make full use of the hash table.
- **secondary clustering:** if  $h(k_1, 0) = h(k_2, 0)$ , then  $h(k_1, i) = h(k_2, i)$

```
static unsigned int oahash(int key, unsigned int i, unsigned int m){  
    return (hash(key, m) + i * i) % m;  
}
```



# Quadratic Probing: Example

- $h'(k) = k \bmod 10, M = 10$   
 $c_1 = 0, c_2 = 1$
- Insert 89, 18, 49, 58, 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

# Double Hashing

---

- Double hashing is one of the **best** methods for open addressing because the permutations produced is quite random.
- It uses hash functions of the form

$$h(k, i) = (h_1(k) + ih_2(k)) \bmod M$$

- Initial probe position is  $h_1(k)$  and successive probe positions are offset by  $h_2(k)$  (modulo  $m$ )
- The value  $h_2(k)$  must be relatively **prime** to the size  $M$  so that the whole table can be searched.
- **Method 1:** choose  $M = 2N$  and design  $h_2$  to return only odd numbers.

# Double Hashing (Cont')

Method 2: choose  $M$  prime and design  $h_2$  to return a positive integer  $< M$

e.g.  $h_1(k) = k \bmod M$ ,  $h_2(k) = 1 + (k \bmod (M - 1))$

```
static unsigned int hash2(int key, unsigned int m){  
    return (1 + (key % (m - 1)));  
}  
  
static unsigned int oahash(int key, unsigned int i, unsigned int m){  
    return (hash(key, m) + i * hash2(key, m)) % m;  
}
```

- Double hashing improves over linear or quadratic probing in that  $\Theta(M^2)$  sequences are used.
- It appears to be very close to the **ideal performance**.



# Double Hashing: Example

- $h_1(k) = k \bmod 10$ ,  
 $M = 10$   
 $h_2(k) = 7 - (k \bmod 7)$
- Insert  
89, 18, 49 (collide),  
58(collide), 69(collide)
- $h_1(49) + h_2(49)$   
 $= (9 + (7 - 0)) \% 10 = 6$
- $h_1(58) + h_2(58)$   
 $= (8 + (7 - 2)) \% 10 = 3$
- $h_1(69) + h_2(69)$   
 $= (9 + (7 - 6)) \% 10 = 0$

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

# Analysis of Open Addressing

---

Given an open-address hash table with load factor  $\alpha = N/M < 1$ , the expected number of probes in an **unsuccessful** search is at most  $1/(1 - \alpha)$ , assuming uniform hashing.

## Intuitive Interpretation

1 probe is always made.

With probability approx.  $\alpha$ , the first probe finds an occupied slot and we need another probe.

With probability approx.  $\alpha^2$ , the first two slots are occupied and we need the third probe.

$$\text{No. of probes} = 1 + \alpha + \alpha^2 + \dots = 1/(1 - \alpha)$$

The no. of probes needed for insertion is the same as that in an unsuccessful search.

# Analysis of Open Addressing (2)

Given an open-address hash table with load factor  $\alpha = N/M < 1$ , assuming uniform hashing, the expected number of probes in a **successful** search is at most

$$\frac{1}{\alpha} \ln \frac{1}{1 - \alpha}$$

## PROOF



If  $k$  was the  $(i + 1)^{\text{th}}$  key inserted, the expected number of

probes made in a search for  $k$  is at most  $\frac{1}{1 - i/M} = \frac{M}{M - i}$

Average over all  $n$  keys gives us the average no .of probes:

$$\begin{aligned} \frac{1}{N} \sum_{i=0}^{N-1} \frac{M}{M - i} &= \frac{M}{N} \sum_{i=0}^{N-1} \frac{1}{M - i} = \frac{1}{\alpha} \sum_{k=M-N+1}^M \frac{1}{k} \\ &\leq \frac{1}{\alpha} \int_{M-N}^M \frac{1}{x} dx = \frac{1}{\alpha} \ln \frac{M}{M - N} = \frac{1}{\alpha} \ln \frac{1}{1 - \alpha} \end{aligned}$$

# Avg. # of Probes in Double Hashing

---

We can ensure that the average cost of all searches is less than  $t$  probes by keeping the load factor less than

$$1 - 1/\sqrt{t} \quad \text{for linear probing}$$

$$1 - 1/t \quad \text{for double hashing}$$

load factor ( $\alpha$ )	1/2	2/3	3/4	9/10
search hit	1.4	1.6	1.8	2.6
search miss	1.5	2.0	3.0	5.5

# Summary

---

- Direct-address table vs Hash table
- Concept of **simple uniform hashing**
- Choice of hash functions for **integers** and **strings**
- **Collision handling**
  - **Chaining**
    - Implementation
    - Performance analysis
  - **Open addressing**: linear & quadratic probing, double hashing
    - Implementation
    - Performance analysis