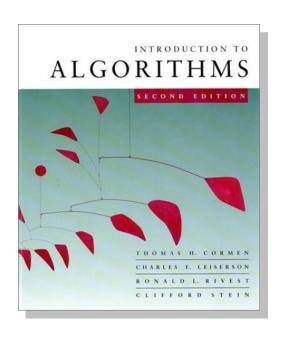
# Introduction to Algorithms 6.046J/18.401J



#### LECTURE 7

#### **Hashing I**

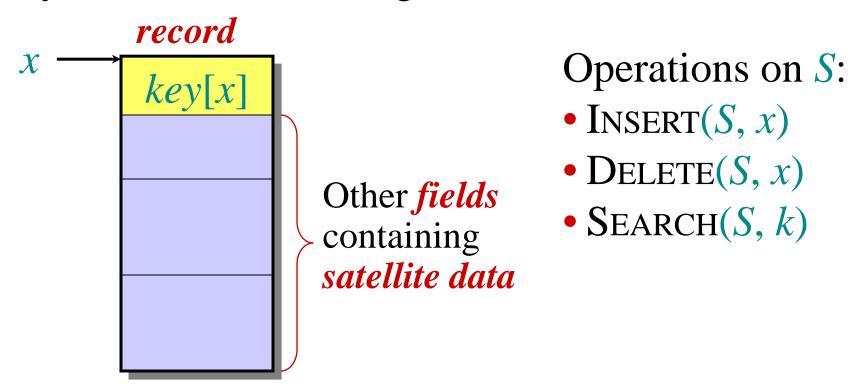
- Direct-access tables
- Resolving collisions by chaining
- Choosing hash functions
- Open addressing

#### Prof. Charles E. Leiserson



#### Symbol-table problem

Symbol table *S* holding *n* records:



How should the data structure *S* be organized?



# Direct-access table

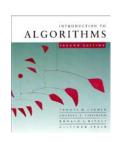
**IDEA:** Suppose that the keys are drawn from the set  $U \subseteq \{0, 1, ..., m-1\}$ , and keys are distinct. Set up an array T[0 ... m-1]:

$$T[k] = \begin{cases} x & \text{if } x \in K \text{ and } key[x] = k, \\ \text{NIL} & \text{otherwise.} \end{cases}$$

Then, operations take  $\Theta(1)$  time.

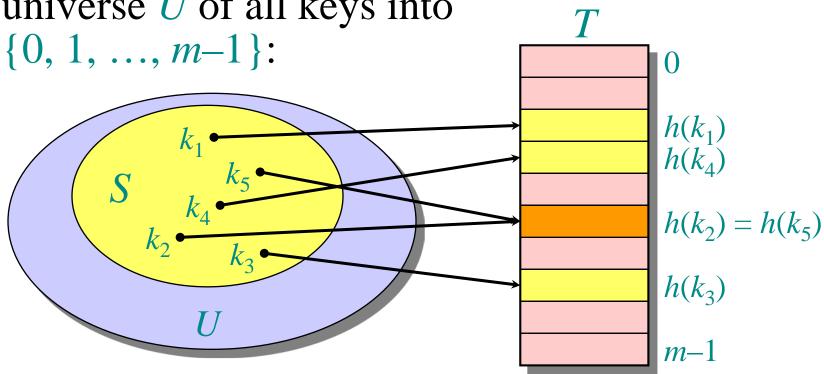
**Problem:** The range of keys can be large:

- 64-bit numbers (which represent 18,446,744,073,709,551,616 different keys),
- character strings (even larger!).

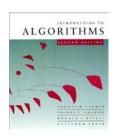


#### Hash functions

**Solution:** Use a *hash function h* to map the universe U of all keys into

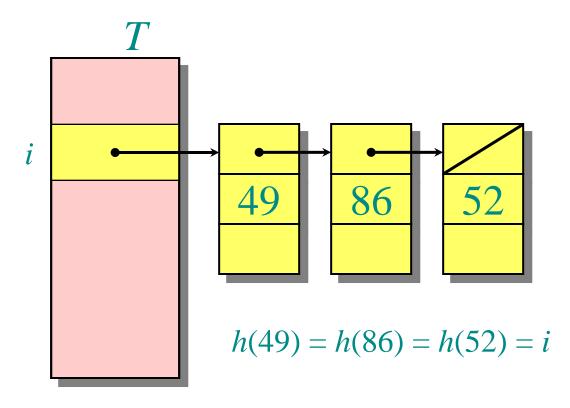


When a record to be inserted maps to an already occupied slot in T, a *collision* occurs.



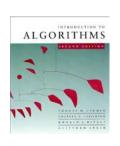
# Resolving collisions by chaining

• Link records in the same slot into a list.



#### Worst case:

- Every key hashes to the same slot.
- Access time =  $\Theta(n)$  if |S| = n



#### Average-case analysis of chaining

We make the assumption of *simple uniform hashing*:

• Each key  $k \in S$  is equally likely to be hashed to any slot of table T, independent of where other keys are hashed.

Let *n* be the number of keys in the table, and let *m* be the number of slots.

Define the *load factor* of *T* to be

 $\alpha = n/m$ 

= average number of keys per slot.



The expected time for an *unsuccessful* search for a record with a given key is

$$=\Theta(1+\alpha).$$



The expected time for an *unsuccessful* search for a record with a given key is



apply hash function and access slot

L7.8



The expected time for an *unsuccessful* search for a record with a given key is

$$=\Theta(1+\alpha)$$
. search the list

apply hash function and access slot

Expected search time =  $\Theta(1)$  if  $\alpha = O(1)$ , or equivalently, if n = O(m).



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Expected search time =  $\Theta(1)$  if  $\alpha = O(1)$ , or equivalently, if n = O(m).

A *successful* search has same asymptotic bound, but a rigorous argument is a little more complicated. (See textbook.)

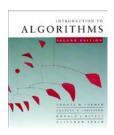


# Choosing a hash function

The assumption of simple uniform hashing is hard to guarantee, but several common techniques tend to work well in practice as long as their deficiencies can be avoided.

#### **Desirata:**

- A good hash function should distribute the keys uniformly into the slots of the table.
- Regularity in the key distribution should not affect this uniformity.



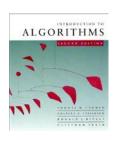
#### **Division method**

Assume all keys are integers, and define  $h(k) = k \mod m$ .

**Deficiency:** Don't pick an *m* that has a small divisor *d*. A preponderance of keys that are congruent modulo *d* can adversely affect uniformity.

Extreme deficiency: If  $m = 2^r$ , then the hash doesn't even depend on all the bits of k:

• If  $k = 1011000111010_2$  and r = 6, then  $h(k) = 011010_2$ .



#### Division method (continued)

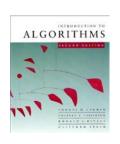
$$h(k) = k \mod m$$
.

Pick *m* to be a prime not too close to a power of 2 or 10 and not otherwise used prominently in the computing environment.

#### **Annoyance:**

• Sometimes, making the table size a prime is inconvenient.

But, this method is popular, although the next method we'll see is usually superior.



## Multiplication method

Assume that all keys are integers,  $m = 2^r$ , and our computer has w-bit words. Define

$$h(k) = (A \cdot k \mod 2^w) \operatorname{rsh} (w - r),$$

where rsh is the "bitwise right-shift" operator and A is an odd integer in the range  $2^{w-1} < A < 2^w$ .

- Don't pick A too close to  $2^{w-1}$  or  $2^w$ .
- Multiplication modulo  $2^w$  is fast compared to division.
- The rsh operator is fast.



# Multiplication method example

$$h(k) = (A \cdot k \mod 2^w) \operatorname{rsh} (w - r)$$

Suppose that  $m = 8 = 2^3$  and that our computer has w = 7-bit words:

 $\frac{1011001}{1001011} = A$   $\frac{1001001}{1001011} = k$   $\frac{10010011}{10011} = k$   $\frac{70}{1}$   $\frac{6}{2}$   $\frac{54}{3}$   $\frac{3}{4}$   $\frac{6}{2}$   $\frac{54}{3}$   $\frac{3}{4}$   $\frac{6}{2}$   $\frac{54}{3}$   $\frac{7}{4}$   $\frac{7}{4}$ 



# Resolving collisions by open addressing

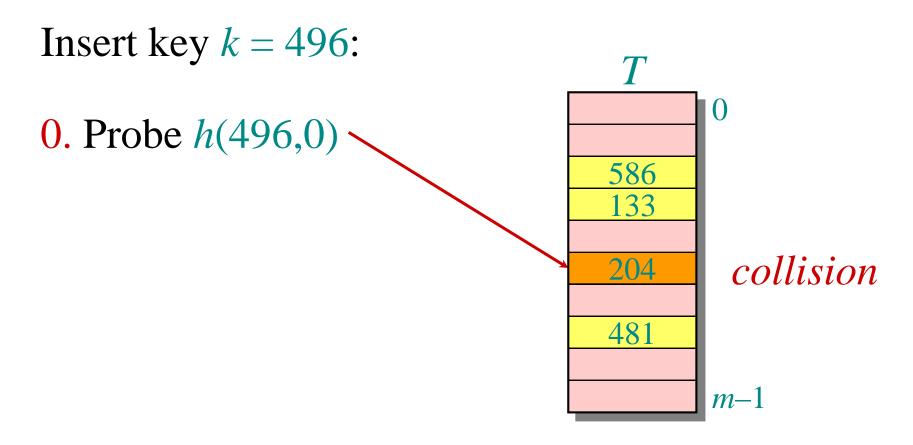
No storage is used outside of the hash table itself.

- Insertion systematically probes the table until an empty slot is found.
- The hash function depends on both the key and probe number:

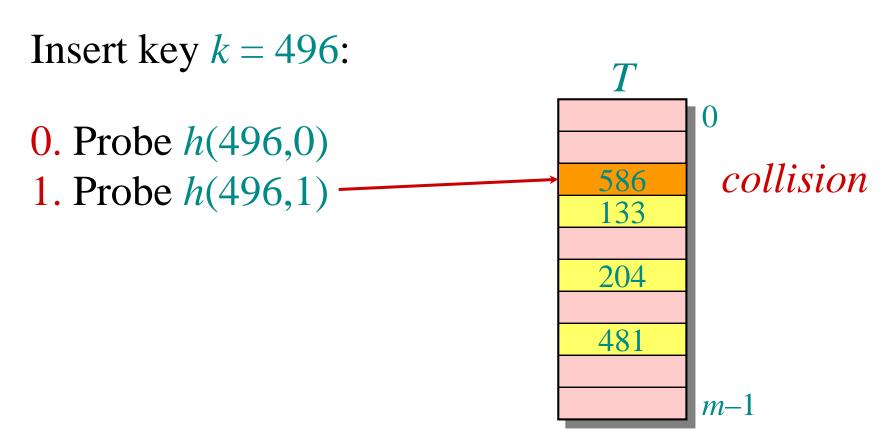
```
h: U \times \{0, 1, ..., m-1\} \rightarrow \{0, 1, ..., m-1\}.
```

- The probe sequence  $\langle h(k,0), h(k,1), ..., h(k,m-1) \rangle$  should be a permutation of  $\{0, 1, ..., m-1\}$ .
- The table may fill up, and deletion is difficult (but not impossible).

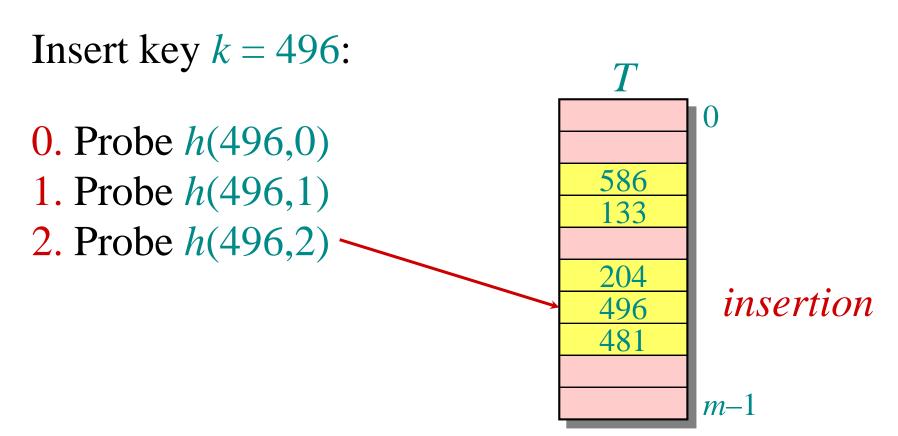




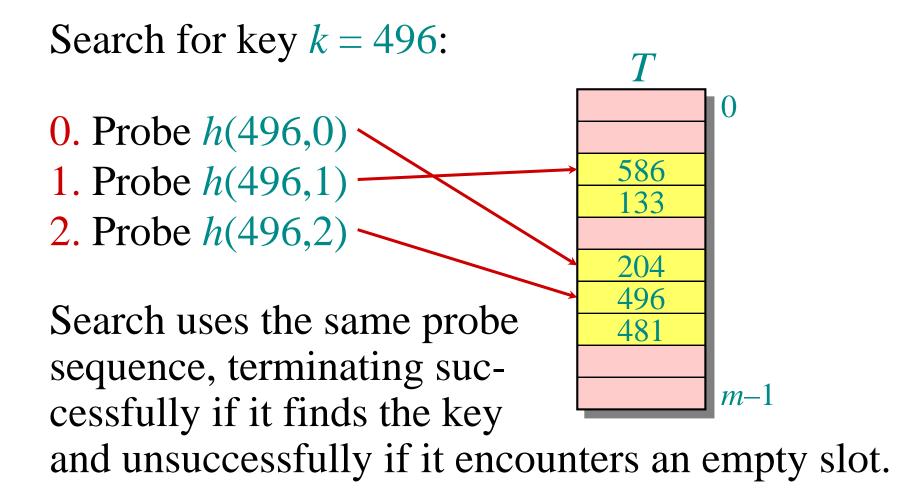














## **Probing strategies**

#### Linear probing:

Given an ordinary hash function h'(k), linear probing uses the hash function

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

This method, though simple, suffers from *primary clustering*, where long runs of occupied slots build up, increasing the average search time. Moreover, the long runs of occupied slots tend to get longer.



## **Probing strategies**

#### **Double hashing**

Given two ordinary hash functions  $h_1(k)$  and  $h_2(k)$ , double hashing uses the hash function

$$h(k,i) = (h_1(k) + i \cdot h_2(k)) \mod m$$
.

This method generally produces excellent results, but  $h_2(k)$  must be relatively prime to m. One way is to make m a power of 2 and design  $h_2(k)$  to produce only odd numbers.



# Analysis of open addressing

We make the assumption of uniform hashing:

• Each key is equally likely to have any one of the *m*! permutations as its probe sequence.

**Theorem.** Given an open-addressed hash table with load factor  $\alpha = n/m < 1$ , the expected number of probes in an unsuccessful search is at most  $1/(1-\alpha)$ .



# Proof of the theorem

#### Proof.

- At least one probe is always necessary.
- With probability n/m, the first probe hits an occupied slot, and a second probe is necessary.
- With probability (n-1)/(m-1), the second probe hits an occupied slot, and a third probe is necessary.
- With probability (n-2)/(m-2), the third probe hits an occupied slot, etc.

Observe that 
$$\frac{n-i}{m-i} < \frac{n}{m} = \alpha$$
 for  $i = 1, 2, ..., n$ .



#### **Proof (continued)**

Therefore, the expected number of probes is

$$1 + \frac{n}{m} \left( 1 + \frac{n-1}{m-1} \left( 1 + \frac{n-2}{m-2} \left( \cdots \left( 1 + \frac{1}{m-n+1} \right) \cdots \right) \right) \right)$$

$$\leq 1 + \alpha \left( 1 + \alpha \left( 1 + \alpha \left( \cdots \left( 1 + \alpha \right) \cdots \right) \right) \right)$$

$$\leq 1 + \alpha + \alpha^2 + \alpha^3 + \cdots$$

$$= \sum_{i=0}^{\infty} \alpha^{i}$$

$$= \frac{1}{1-\alpha} \cdot \square$$

The textbook has a more rigorous proof and an analysis of successful searches.



#### Implications of the theorem

- If  $\alpha$  is constant, then accessing an openaddressed hash table takes constant time.
- If the table is half full, then the expected number of probes is 1/(1-0.5) = 2.
- If the table is 90% full, then the expected number of probes is 1/(1-0.9) = 10.