

# Unit #5: Hash functions and the Pigeonhole principle

CPSC 221: Algorithms and Data Structures

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# Unit Outline

- ▶ Constant-Time Dictionaries?
- ▶ Hash Table Outline
- ▶ Hash Functions
- ▶ Collisions and the Pigeonhole Principle
- ▶ Collision Resolution:
  - ▶ Separate Chaining
  - ▶ Open Addressing

# Learning Goals

- ▶ Provide examples of the types of problems that can benefit from a hash data structure.
- ▶ Identify the types of search problems that do not benefit from hashing (e.g. range searching) and explain why. *Alice ... Bob*
- ▶ Evaluate collision resolution policies. *or anything which requires keys to be ordered*
- ▶ Compare and contrast open addressing and chaining.
- ▶ Describe the conditions under which find using a hash table takes  $\Omega(n)$  time.
- ▶ Insert, delete, and find using various open addressing and chaining schemes.
- ▶ Define various forms of the pigeonhole principle; recognize and solve the specific types of counting and hashing problems to which they apply.

# Reminder: Dictionary ADT

## Dictionary operations

- ▶ create
- ▶ destroy
- ▶ insert
- ▶ find
- ▶ delete

key	value
Multics	MULTiplexed Information and Computing Service
Unics	single-user Multics
Unix	multi-user Unics
GNU	GNU's Not Unix

- ▶ insert(Linux, Linus Torvald's Unix)
- ▶ find(Unix)

Stores values associated with user-specified keys

# Hash Table Goal

We can do:

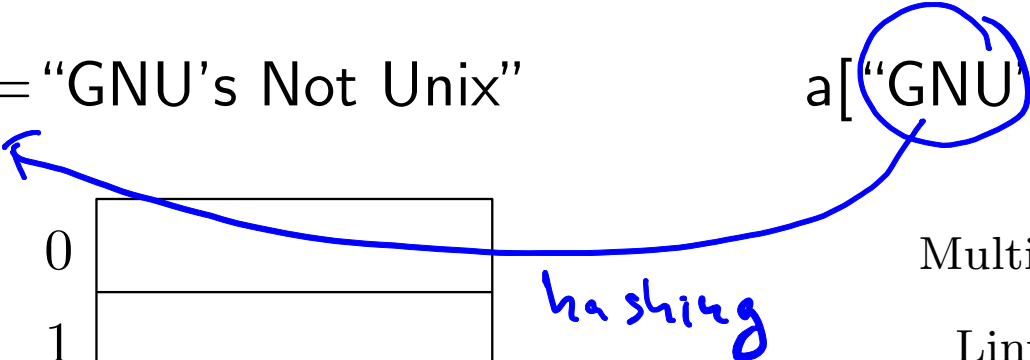
$a[2] = \text{"GNU's Not Unix"}$

0	
1	
2	GNU's Not Unix
3	
	⋮
$m - 1$	

We want to do:

$a[\text{"GNU"}] = \text{"GNU's Not Unix"}$

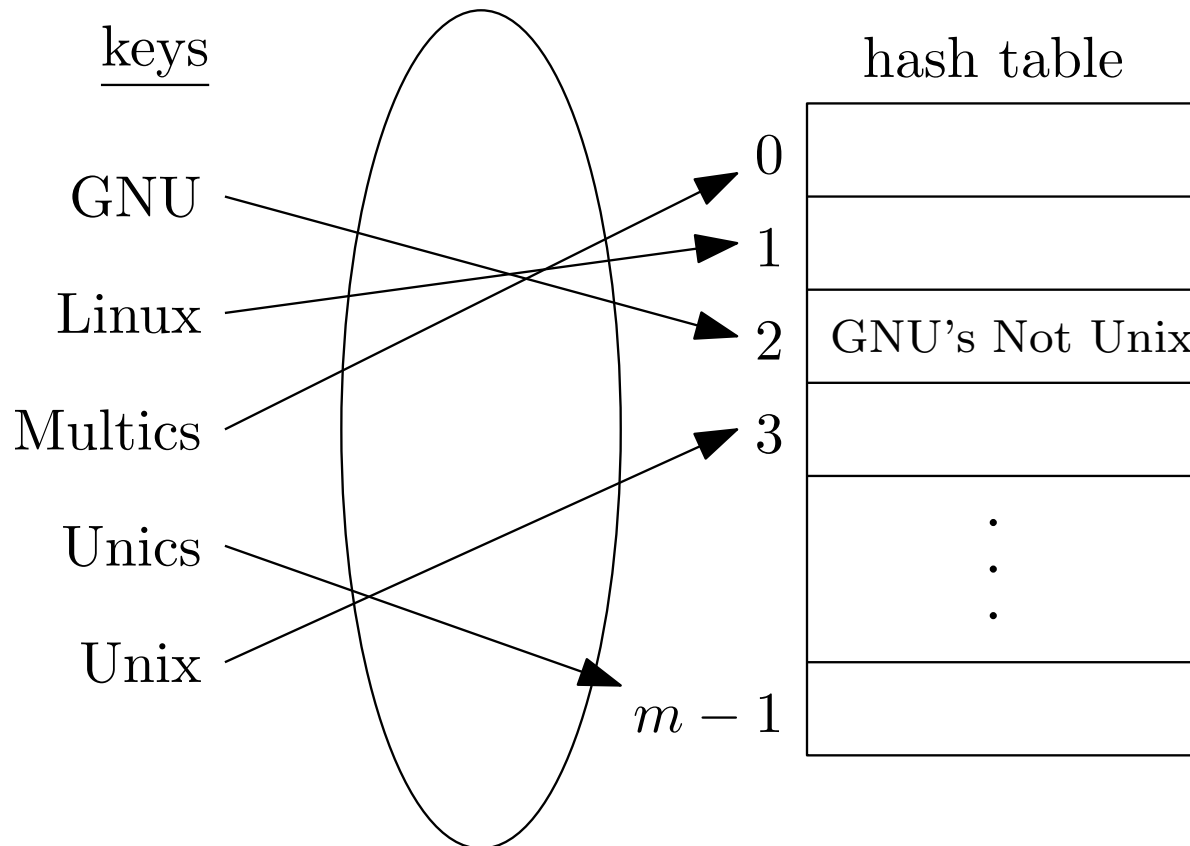
Multics	
Linux	
GNU	GNU's Not Unix
Unix	
	⋮
Unics	



associative arrays  
(also called)

# Hash table approach

Choose a **hash function** to map keys to indices.



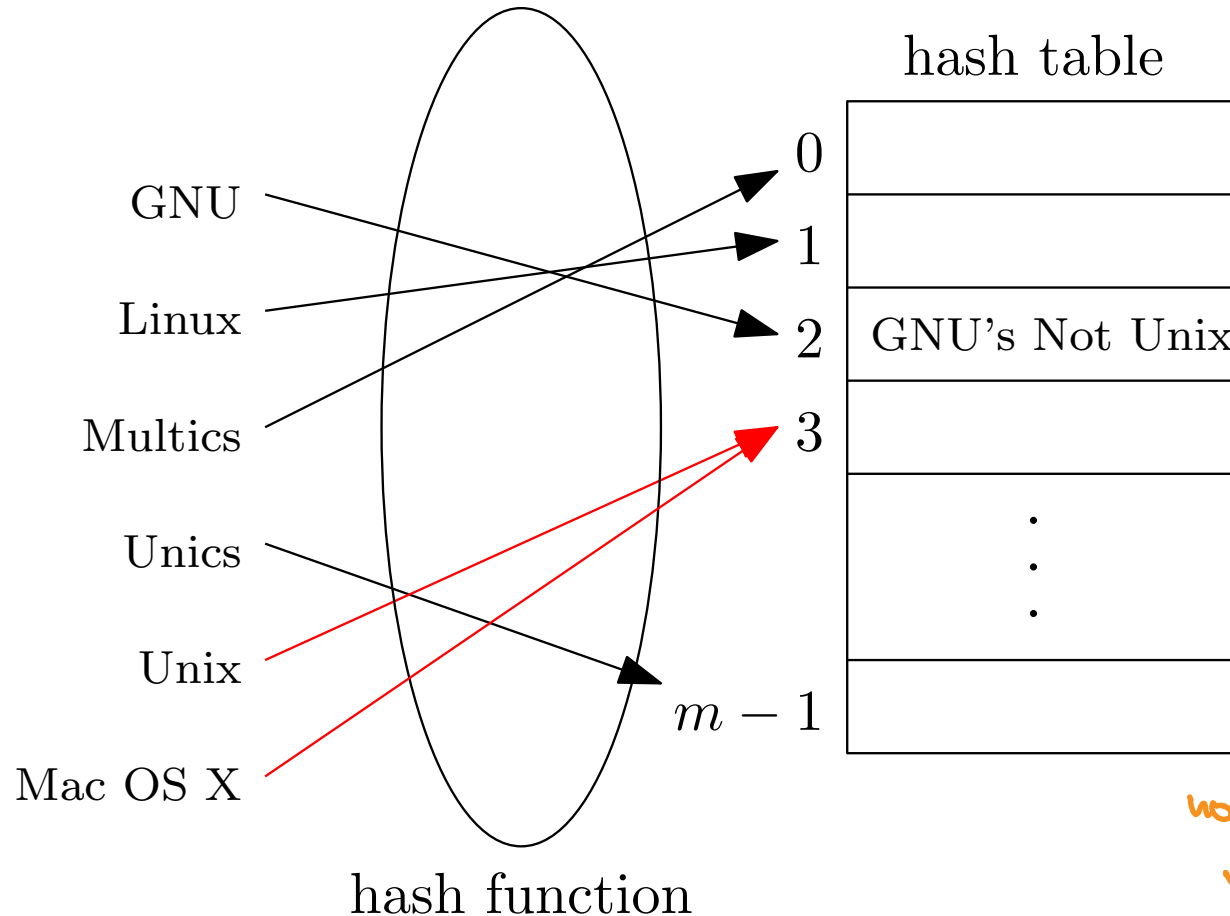
hash function

$$\text{hash}(\text{"GNU"}) = 2$$

we want hash function:  
fast computing  
few / no collisions

# Collisions

A **collision** occurs when two different keys  $x$  and  $y$  map to the same index (i.e. **slot** in table),  $\text{hash}(x) = \text{hash}(y)$ .



not practical in most cases



Can we prevent collisions?

Not completely  
(unless the size of table = # of possible keys)

# Simple, naïve hash table code

```
void insert(const Key & key ) {  
    int index = hash(key) % m;  
    HashTable[index] = key;  
}
```

size m

Problems:

.. overwrites value if  
there is a collision

```
Value & find(const Key & key ) {  
    int index = hash(key) % m;  
    return HashTable[index];  
}
```

.. doesn't  
test if key exists

What should the hash function, hash, be?

What should the table size,  $m$ , be?

What do we do about collisions?



# Good hash function properties

Using knowledge of the kind and number of keys to be stored, we should choose our hash function so that it is:

- ▶ fast to compute, and
- ▶ causes few collisions (we hope).

**Numeric keys** We might use  $\text{hash}(x) = x \bmod m$  with  $m$  larger than the number of keys we expect to store.

Example:  $\text{hash}(x) = x \bmod 7$

insert(4)

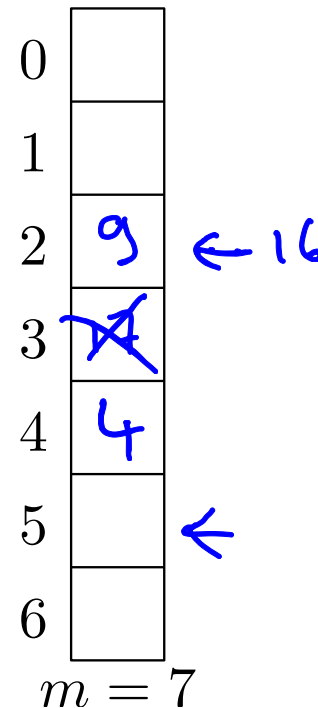
insert(17)

find(12)

insert(9)

delete(17)

insert(16)  $\Rightarrow$  collision!



# Hashing string keys

Example of  
a simple hash  
function:

$$s_i \in \{0, \dots, 255\}$$
$$\text{hash}'(s) = \sum_{i=0}^{k-1} s_i$$

One option

$$\text{hash}(\text{'dog'}) = \text{hash}(\text{'god'})$$
$$= \text{hash}(\text{'odg'}) = \dots$$

↑ can lead to many collisions

Let string  $s = s_0 s_1 s_2 \dots s_{k-1}$  where each  $s_i$  is an 8-bit character.

converting string to numerical value

$$\text{hash}(s) = s_0 + 256s_1 + 256^2s_2 + \dots + 256^{k-1}s_{k-1}$$

↑  
unique as long as  $s_{k-1} \neq 0$

Hash function treats string as a base 256 number.

$$\rightarrow 157 = 7 + 5 \cdot 10 + 1 \cdot 10^2$$

## Problems

- ▶  $\text{hash}(\text{"really, really big"}) = \text{well... something really, really big}$
- ▶  $\text{hash}(\text{"anything"}) \bmod 256 = \text{hash}(\text{"anything else"}) \bmod 256$

# Hashing string keys with mod and Horner's Rule

```
int hash( string s ) {
    int h = 0;
    for (i = s.length() - 1; i >= 0; i--) {
        h = (256 * h + s[i]) % m;
    }
    return h;
}
```

$$a + b \cdot 256 + c \cdot 256^2 = a + 256 \cdot (b + c \cdot 256)$$

mod m

mod

size of the hash table

Compare that to the hash function from yacc:

```
#define TABLE_SIZE 1024 // must be a power of 2
int hash( char *s ) {
    int h = *s++;
    while( *s ) h = (31 * h + *s++) & (TABLE_SIZE - 1);
    return h;
}
```

bitwise AND



s[i]

assumes null-terminated string

$$\begin{array}{r} 10110 \dots 22 \\ \underline{111 \dots 8-1} \\ 110 \dots 6 = 22 \text{ mod } 8 \end{array}$$

What's different?


# Hash Function Summary

## Goals of a hash function

- ▶ Fast to compute
- ▶ Cause few collisions

## Sample hash functions

- ▶ For numeric keys  $x$ ,  $\text{hash}(x) = x \bmod m$
- ▶  $\text{hash}(s) = \text{string as base 256 number} \bmod m$
- ▶ Multiplicative hash:  $\text{hash}(k) = \lfloor m \cdot \text{frac}(ka) \rfloor$  where  $\text{frac}(x)$  is the fractional part of  $x$  and  $a = 0.6180339887$  (for example).

  
Knutz

# Fixed hash functions are dangerous

Good hash table performance depends on few collisions.

If a user knows your hash function, she can cause many elements to hash to the same slot. Why would she want to do that?

## Denial of Service

Yacc hashes "XY" and "xy" to 769. How can you find many strings that yacc hashes to the same slot?

$$h('xY') = h('xy')$$
$$'x' + 31 \cdot 'Y' = 'x' + 31 \cdot 'y'$$

## Protection

- ▶ Choose a new hash function at random for every hash table.
- ▶ Use a cryptographically secure hash function (such as SHA-2).

for any  $k$  :  $\{XY, xy\}^k$  all mapped to same slot

for  $k=2$  :

$$h(XYXY) = h(XYxy) = h(xyXY) = h(xyxy)$$
$$= 31^2 \cdot h(XY) + h(XY)$$

# Universal hash functions

A set  $\mathcal{H}$  of hash functions is *universal* if the probability that  $\text{hash}(x) = \text{hash}(y)$  is at most  $1/m$  when  $\text{hash}()$  is chosen at random from  $\mathcal{H}$ .  $x \neq y$

Example: Let  $p$  be a prime number larger than any key. Choose  $a$  at random from  $\{1, 2, \dots, p-1\}$  and choose  $b$  at random from  $\{0, 1, \dots, p-1\}$ .

parameters

↓                      ↓

$$\text{hash}(x) = ((a \cdot x + b) \bmod p) \bmod m$$

---

# General form of hash functions

1. Map key to a sequence of bytes.
  - ▶ Two equal sequences iff two equal keys.
  - ▶ Easy. The key probably is a sequence of bytes already.
2. Map sequence of bytes to an integer  $x$ .
  - ▶ Changing bytes should cause apparently **random** changes to  $x$ .
  - ▶ Hard. May be expensive. Cryptographic hash.
3. Map  $x$  to a table index using  $x \bmod m$ .
  - ▶  $m$  is the size of hash table

# Collisions

## Birthday Paradox

With probability  $> \frac{1}{2}$ , two people, in a room of 23, have the same birthday. (Hash 23 people into  $m = 365$  slots. Collision?)

## General birthday paradox

If we *randomly* hash  $\sqrt{2m}$  keys into  $m$  slots, we get a collision with probability  $> \frac{1}{2}$ .

## Collision

Unless we know all the keys in advance and design a perfect hash function, we must handle collisions.

What do we do when two keys hash to the same slot?

- ▶ separate chaining: store multiple items in each slot
- ▶ open addressing: pick a next slot to try

$n$  keys  
pair of keys  
# pairs

$\frac{1}{m}$

$\binom{n}{2} = \frac{n(n-1)}{2} \sim \frac{n^2}{2}$

exp. #collisions:  $\frac{n^2}{2} \cdot \frac{1}{m}$

this is 1 if  $n = \sqrt{2m}$



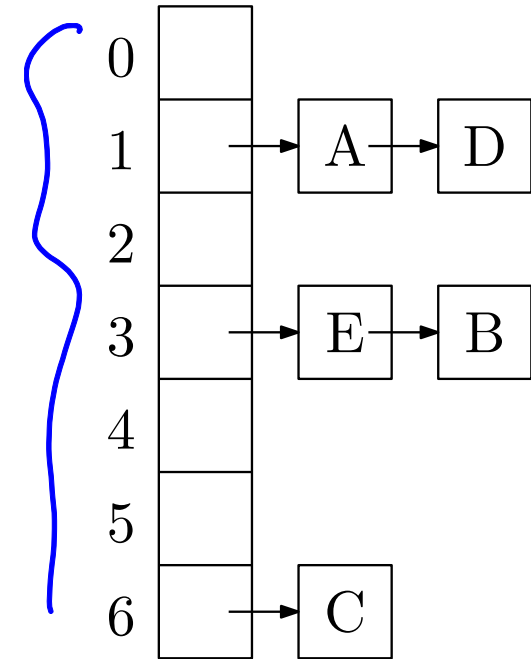
# Hashing with Chaining

## Separate Chaining

Store multiple items in each slot.

How?

- ▶ Common choice is an unordered linked list (a chain).
- ▶ Could use any dictionary ADT implementation.



Result

- + ▶ Can hash more than  $m$  items into a table of size  $m$ .
- ▶ Performance depends on the length of the chains.
- ▶ Memory is allocated on each insertion.

$n$  items  
worst-case:  $\Theta(n)$  time  $\nearrow \alpha$   
even distrib.-case:  $\Theta(\frac{n}{m})$  time } for find()

hash(A) = hash(D) = 1

hash(E) = hash(B) = 3

more exact  $\longrightarrow$

# Access time for Chaining

## Load Factor

$$\alpha = \frac{\# \text{ hashed items}}{\text{table size}} = \frac{n}{m}$$

Assume we have a uniform hash function (every item hashes to a uniformly distributed slot).

## Search cost

On average,

▶ an unsuccessful search examines  $\alpha$  items.

▶ a successful search examines  $1 + \frac{\alpha}{2} - \frac{\alpha}{2n}$  items.

to find the item  
 $n-1$  remaining items  
 $\frac{n-1}{m}$  on average in this chain  
on average  
half of them are before the value

We want the load factor to be small.

# Open Addressing

Allow only one item in each slot. The hash function specifies a *sequence* of slots to try.

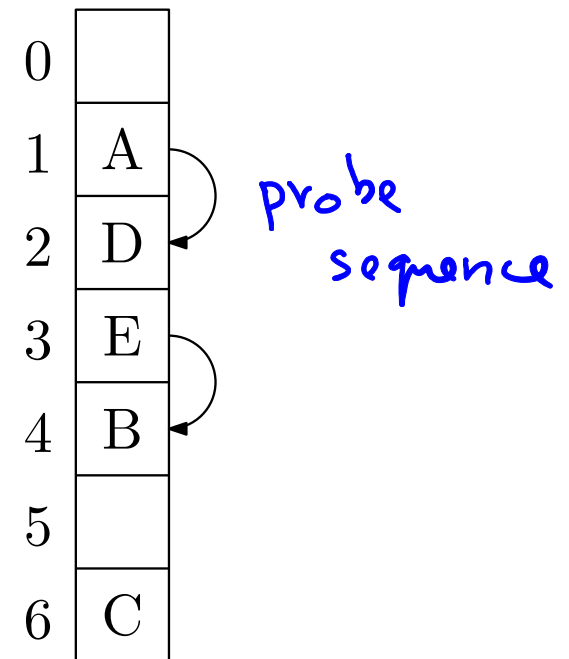
$$\begin{aligned} \text{hash}(A) &= \text{hash}(D) = 1 \\ \text{hash}(E) &= \text{hash}(B) = 3 \end{aligned}$$

**Insert** If the first slot is occupied, try the next, then the next, ... until an empty slot is found.

**Find** If the first slot doesn't match, try the next, then the next, ... until a match (found) or an empty slot (not found).

## Result

- ▶ Cannot hash more than  $m$  items into a table of size  $m$ . [Pigeonhole Principle]
- ▶ Hash table memory allocated once.
- ▶ Performance depends on number of tries.



# Probe Sequence

The sequence of slots we examine when inserting (and finding) a key.

A probe sequence is a function,  $h(k, i)$ , that maps a key  $k$  and an integer  $i$  to a table index.

Given key  $k$ :

- ▶ We first examine slot  $h(k, 0)$ .
- ▶ If it's full, we examine slot  $h(k, 1)$ .
- ▶ If it's full, we examine slot  $h(k, 2)$ .
- ▶ And so on...

} linear  
quadratic  
double hashing

If all the slots in the probe sequence are full, we fail to insert the key.

The time to insert is the number of slots we must examine before finding an empty slot.

Linear probing:  $h(k, i) = (\text{hash}(k) + i) \bmod m$

↑  
size of hash  
table

```
Entry *find( const Key & k ) {
```

```
    int p = hash(k) % size;
```

```
    for( int i=1; i<=size; i++ ) {
```

```
        Entry *entry = &(table[p]);
```

```
        if( entry->isEmpty() ) return NULL;
```

```
        if( entry->key == k ) return entry;
```

```
        p = (p + 1) % size;
```

```
    }
```

```
    return NULL;
```

```
}
```

initial value

LI:  $p = (\text{hash}(k) + i - 1) \bmod m$

empty slot  
⇒ fail

found

whole table searched ⇒ fail

Useful mod arithmetics:

$$(a+b) \bmod m = (a \bmod m + b \bmod m) \bmod m$$

$$(a \cdot b) \bmod m = (a \bmod m \cdot b \bmod m) \bmod m$$

# Linear probing example

$$h(k, i) = (k + i) \bmod 7$$

insert(76)

$$76 \% 7 = 6$$

0	
1	
2	
3	
4	
5	
6	76 ◦

insert(93)

$$93 \% 7 = 2$$

0	
1	
2	93 ◦
3	
4	
5	
6	76

insert(40)

$$40 \% 7 = 5$$

0	
1	
2	93
3	
4	
5	40 ◦
6	76

insert(47)

$$47 \% 7 = 5$$

0	47 ◦
1	
2	93
3	
4	
5	40 ●
6	76 ●

insert(10)

$$10 \% 7 = 3$$

0	47
1	
2	93
3	10 ◦
4	
5	40
6	76

insert(55)

$$55 \% 7 = 6$$

0	47 ●
1	55 ◦
2	93
3	10
4	
5	40
6	76 ●

◦ .. success

● .. fail

# Access time for linear probing

+ If  $\alpha < 1$ , linear probing will find an empty slot.

Linear probing suffers from **primary clustering**: creation of long consecutive sequences of filled slots. (They tend to get longer and merge.)

- Performance quickly degrades for  $\alpha > 1/2$ .

longer the sequence, higher prob. it gets hit by next insert, which will make it bigger

load f. $\alpha$	# probes for unsuc. search
0.6	3.6
0.7	6.1
0.8	13
0.9	50.5

Quadratic probing:  $h(k, i) = (\text{hash}(k) + \underline{i^2}) \bmod m$

size = m

```
Entry *find( const Key & k ) {
```

```
    int p = hash(k) % size;
```

```
    for( int i=1; i<=size; i++ ) {
```

```
        Entry *entry = &(table[p]);
```

```
        if( entry->isEmpty() ) return NULL;
```

```
        if( entry->key == k ) return entry;
```

```
        p = (p + 2*i - 1) % size;
```

```
    }  
    return NULL;
```

```
}
```

$$i^2 = \underline{(i-1)^2} + \underline{2i-1}$$

$$[\text{hash}(k) + (i-1)^2] \bmod m$$
$$[\text{hash}(k) + i^2] \bmod m$$



# Quadratic probing example

$$h(k, i) = (k + i^2) \bmod 7$$

insert(76)

$$76 \% 7 = 6$$

0	
1	
2	
3	
4	
5	
6	76 ◦

insert(40)

$$40 \% 7 = 5$$

0	
1	
2	
3	
4	
5	40 ◦
6	76

insert(48)

$$48 \% 7 = 6$$

0	48 ◦
1	
2	
3	
4	
5	40
6	76 ●

$i=1$  (at index 0)  
 $i=0$  (at index 6)

insert(5)

$$5 \% 7 = 5$$

0	48
1	
2	5 ◦
3	
4	
5	40 ●
6	76 ●

$i=2$  (at index 2)  
 $i=0$  (at index 5)  
 $i=1$  (at index 6)

insert(55)

$$55 \% 7 = 6$$

0	48 ●
1	
2	5
3	55 ◦
4	
5	40
6	76 ●

# Quadratic probing example

$m = 7$

insert(76)

$$76 \% 7 = 6$$

0	
1	
2	
3	
4	
5	
6	76

insert(93)

$$93 \% 7 = 2$$

0	
1	
2	93
3	
4	
5	
6	76

insert(40)

$$40 \% 7 = 5$$

0	
1	
2	93
3	
4	
5	40
6	76

insert(35)

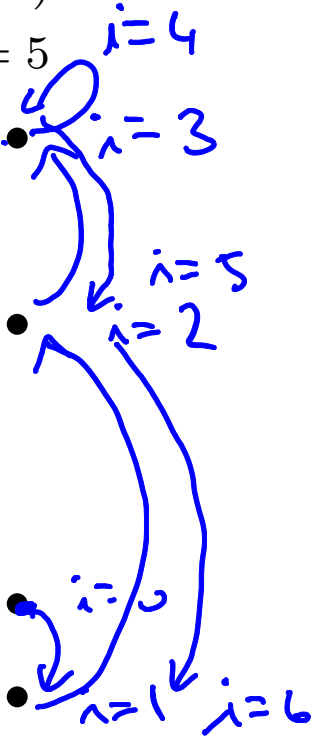
$$35 \% 7 = 0$$

0	35
1	
2	93
3	
4	
5	40
6	76

insert(47)

$$47 \% 7 = 5$$

0	35
1	
2	93
3	
4	
5	40
6	76



fail

probes

$i, m-i$

hashed to same slot

# Quadratic probing: First $\lceil m/2 \rceil$ probes are distinct

Claim: If  $m$  is prime the first  $\lceil m/2 \rceil$  probes are distinct.   
  $\rightarrow 0, \dots, \lceil m/2 \rceil - 1 \leq \lfloor m/2 \rfloor$

Proof: (by contradiction) Suppose for some  $0 \leq i < j \leq \lfloor m/2 \rfloor$ ,

$$(\text{hash}(k) + i^2) \bmod m = (\text{hash}(k) + j^2) \bmod m$$

$$\Leftrightarrow i^2 \bmod m = j^2 \bmod m$$

$$\Leftrightarrow (i^2 - j^2) \bmod m = 0$$

$$\Leftrightarrow (i - j)(i + j) \bmod m = 0$$

Since  $m$  is prime, one of  $(i - j)$  and  $(i + j)$  must be divisible by  $m$ .

But  $0 < i + j < m$  and  $-\lfloor m/2 \rfloor \leq i - j < 0$  because

$0 \leq i < j \leq \lfloor m/2 \rfloor$ . So neither can be divisible, a contradiction.

**Result** since  $m$  is odd (alternative explanation:  $i \leq \lfloor m/2 \rfloor - 1, j \leq \lfloor m/2 \rfloor$ ,  
so  $i + j \leq 2\lfloor m/2 \rfloor - 1$

If table size  $m$  is prime and there are  $< \lceil m/2 \rceil$  full slots (i.e.,  $\leq m - 1$ )  
 $\alpha < 1/2$ ), then quadratic probing will find an empty slot.

so one of those  $\lceil m/2 \rceil$  slots is empty

# Quadratic probing: Only $\lceil m/2 \rceil$ probes are distinct *bad*

**Claim:** For any  $j \in \{\lceil m/2 \rceil, \lceil m/2 \rceil + 1, \dots, m - 1\}$ , there is an  $i \in \{1, 2, \dots, \lfloor m/2 \rfloor\}$  such that  $i^2 \bmod m = j^2 \bmod m$ .

**Proof:** Let  $i = m - j$ .

$$i^2 = (m - j)^2 = m^2 - 2mj + j^2 = j^2 \pmod{m}.$$

**For example:**  $m = 7$

$$\text{hash}(k) + 0^2 = \text{hash}(k) + 0 \pmod{7}$$

$$\text{hash}(k) + 1^2 = \text{hash}(k) + 1 \pmod{7}$$

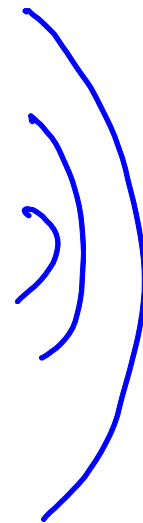
$$\text{hash}(k) + 2^2 = \text{hash}(k) + 4 \pmod{7}$$

$$\text{hash}(k) + 3^2 = \text{hash}(k) + 2 \pmod{7}$$

$$\text{hash}(k) + 4^2 = \text{hash}(k) + 2 \pmod{7}$$

$$\text{hash}(k) + 5^2 = \text{hash}(k) + 4 \pmod{7}$$

$$\text{hash}(k) + 6^2 = \text{hash}(k) + 1 \pmod{7}$$



# Access time for quadratic probing

- Only the first  $\lceil m/2 \rceil$  slots in a quadratic probe sequence are distinct — the rest are duplicates.
- + Quadratic probing doesn't suffer from primary clustering.
- Quadratic probing suffers from **secondary clustering**: all items that initially hash to the same slot follow that same probe sequence.

How could we avoid that?

Different probing sequence for different keys.

Double hashing:  $h(k, i) = (\text{hash}(k) + i \cdot \text{hash}_2(k)) \bmod m$

*inc*

```
Entry *find( const Key & k ) {  
    int p = hash(k) % size, inc = hash2(k);  
    for( int i=1; i<=size; i++ ) {  
        Entry *entry = &(table[p]);  
        if( entry->isEmpty() ) return NULL;  
        if( entry->key == k ) return entry;  
        p = (p + inc) % size;  
    }  
    return NULL;  
}
```

*< LI:  $p = [\text{hash}(k) + (i-1) * \text{hash}_2(k)] \bmod m$*

# Choosing $\text{hash}_2(k)$

$\text{hash}_2(k)$  should:

- ▶ be quick to evaluate
- ▶ differ from  $\text{hash}(k)$
- ▶ never be 0 (mod  $m$ )

We'll use:

$$\text{hash}_2(k) = r - (k \bmod r)$$

for a prime number  $r < m$ .

?

0...r-1

0...r-1

1..r

1..r

# Double hashing example

$r = 5$

$$h(k, i) = [k + i \underbrace{(5 - k \bmod 5)}_{inc}] \bmod 7$$

insert(76)

$$76 \% 7 = 6$$

0	
1	
2	
3	
4	
5	
6	76 ◦

insert(93)

$$93 \% 7 = 2$$

0	
1	
2	93 ◦
3	
4	
5	
6	76

insert(40)

$$40 \% 7 = 5$$

0	
1	
2	93
3	
4	
5	40 ◦
6	76

insert(47)

$$47 \% 7 = 5$$

$$5 - (47 \% 5) = 3$$

0	
1	47 ◦
2	93
3	
4	
5	40
6	76

$i=1$  (at index 1)  
 $i=0$  (at index 5)

insert(10)

$$10 \% 7 = 3$$

0	
1	47
2	93
3	10 ◦
4	
5	40
6	76

insert(55)

$$55 \% 7 = 6$$

$$5 - (55 \% 5) = 5$$

0	
1	47
2	93
3	10
4	55 ◦
5	40
6	76

$i=1$  (at index 4)  
 $i=0$  (at index 6)



# Access time for double hashing

+ For  $\alpha < 1$ , double hashing will find an empty slot (assuming  $m$  and  $hash_2$  are well-chosen).

$\leftarrow \neq 0$

$\uparrow$   
prime

+ No primary or secondary clustering.

- One extra hash calculation.

This is not true for double hashing  
 $\checkmark$  probing sequence, but we will assume  
that to simplify analysis.

Q. Assume prob. sequence is a random sequence  
load factor  $\alpha = \frac{n}{m}$ . We want to insert.

(1) Prob. of success of one probe?  $1 - \alpha$

(2) Exp. #probes until success?  $\frac{1}{1 - \alpha}$

$\uparrow$  similar performance  
for double hashing

# Deletion in Open Addressing

Example:  $\text{hash}(k) = k \bmod 7$ .

delete(2)

0	0
1	1
2	2
3	7
4	
5	
6	

find(7)

0	0	← not here
1	1	← not here
2		← end of search?!
3	7	
4		
5		
6		

Put a **tombstone** in the slot.

**Find** Treat tombstone as an occupied slot.

**Insert** Treat tombstone as an empty slot.

However, you may need to Find before Insert if you want to avoid duplicate keys (which you do).

# Deletion in Open Addressing

Example:  $\text{hash}(k) = k \bmod 7$ .

Example :

insert (9)

delete (2) →

insert (9)

(shows we need

to find() before insert()

if we want to avoid duplicate

keys)

Put a tombstone in the slot.

Find Treat tombstone as an occupied slot.


Insert Treat tombstone as an empty slot.

\* However, you may need to Find before Insert if you want to avoid duplicate keys (which you do).

delete(2)

0	0
1	1
2	<del>2</del> 9
3	7
4	9
5	
6	

find(7)

0	0
1	1
2	
3	7
4	
5	
6	

← not here

← not here

← keep going

← here!

If same key appears multiple times, we have no idea which will be returned by find() or deleted by delete().

# Resizable hash tables

An insert using open addressing cannot succeed with a load factor of 1 or more. [Pigeonhole Principle]

An insert using open addressing with quadratic probing may not succeed with a load factor  $> 1/2$ .

Whether you use chaining or open addressing, large load factors lead to poor performance!

Hint: Think resizable arrays!

# Rehashing

$\Theta(n+m)$  for separate chaining

we need to go through hash table to find all elements in it

When the load factor gets “too large” ( $\alpha >$  some constant threshold), rehash all the elements into a new, larger table:

- ▶ takes  $\Theta(n)$  time, but amortized  $O(1)$  as long as we double table size on the resize
- ▶ spreads keys back out, may drastically improve performance
- ▶ gives us a chance to change the hash function
- ▶ avoids failure for open addressing techniques
- ▶ allows arbitrarily large tables starting from a small table
- ▶ clears out tombstones

tombstones can significantly slow down the performance, as they make probe sequences for find() very long.

# The Pigeonhole Principle

If more than  $m$  pigeons fly into  $m$  pigeonholes then some pigeonhole contains at least two pigeons.

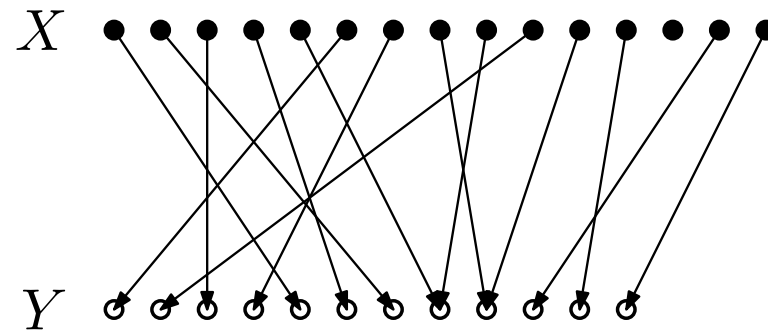
## Corollary

If we hash  $n > m$  keys into  $m$  slots, two keys will collide.

# The Pigeonhole Principle

Let  $X$  and  $Y$  be finite sets where  $|X| > |Y|$ .

If  $f : X \rightarrow Y$ , then  $f(x_1) = f(x_2)$  for some  $x_1 \neq x_2$ .



# The Pigeonhole Principle: Example #1

Suppose we have 5 colours of Halloween candy, and that there's lots of candy in a bag. How many pieces of candy do we have to pull out of the bag if we want to be sure to get 2 of the same colour?

a. 2

b. 4

c. 6

d. 8

e. None of these

pigeons = candy

holes = colors



# The Pigeonhole Principle: Example #2

## Compression

Any lossless compression algorithm (such as zip, bzip2, Huffman coding, Sequitur, etc.) will fail to compress some file.

Proof by contradiction:

How many files containing  $n$  bits are there?

$$2^n$$

How many files containing fewer than  $n$  bits are there?

$$\sum_{i=0}^{n-1} 2^i = 2^n - 1$$

What are the pigeons? pigeonholes?

↓  
all  $n$ -bit files

$$\# = 2^n$$

↓  
compressed files

$$\# = 2^n - 1$$

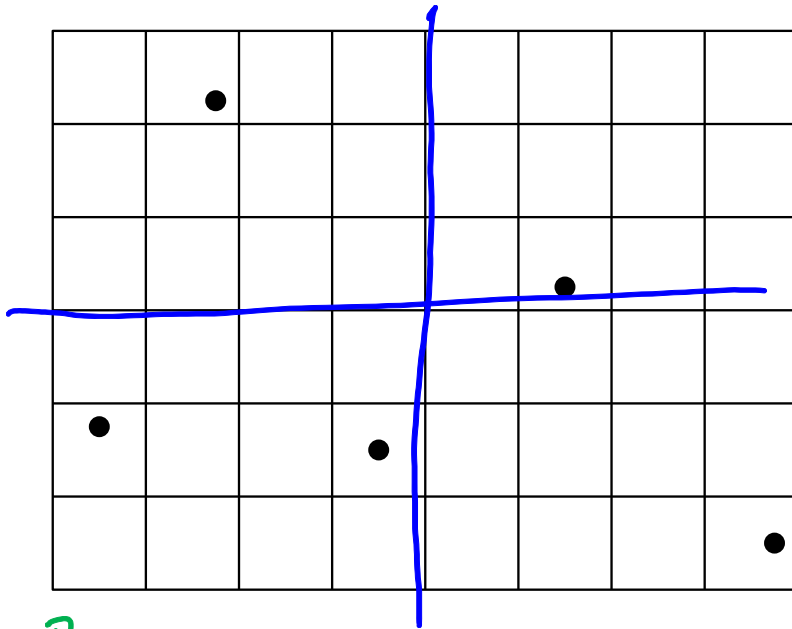
PHP: at least two files will be compressed to the same file

not lossless compression

# The Pigeonhole Principle: Example #3

— pigeons

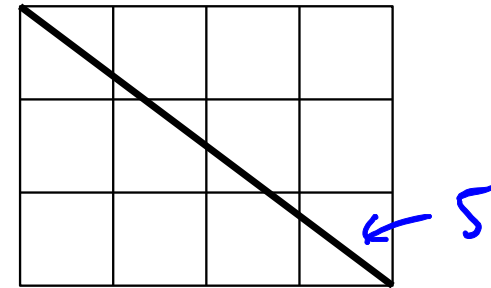
If 5 points are placed in a 6cm x 8cm rectangle, there are two points that are  $\leq 5$  cm apart.



by PHP:  
there will be  $\nearrow$  3x4 box  
containing 2 points

Example: 12x12 cm square  
we need 13 points

Hint: How long is  
this diagonal?



$\nearrow$  3x4  
holes

# The Pigeonhole Principle: Example #4

Consider  $n + 1$  distinct positive integers, each  $\leq 2n$ . Show that one of them must divide one of the others.

For example, if  $n = 4$ , consider the following sets:

$$\{1, 2, 3, 7, 8\} \quad \{2, 3, 4, 7, 8\} \quad \{2, 3, 5, 7, 8\}$$

Hint: Any integer can be written as  $2^k \cdot q$  where  $k$  is an integer and  $q$  is odd. E.g.,  $129 = 2^0 \cdot 129$ ;  $60 = 2^2 \cdot 15$ .

holes  $\rightarrow$   $\{1, 3, 5, \dots, 2n-1\}$   $n$  possible values

there are 2 numbers with same  $q$

$2^k \cdot q$

$2^l \cdot q = 2^{l-k} \cdot 2^k \cdot q$  divides

$120 = 2^3 \cdot 15$   
 $k < l$

# General Pigeonhole Principle

Ex:  $m$  size

$2m+1$  keys

$$\lceil \frac{2m+1}{m} \rceil = 3$$

$\geq$  least keys in 1 slot

holes

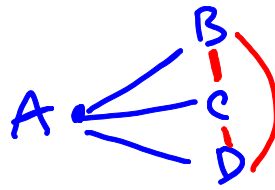
Let  $X$  and  $Y$  be finite sets with  $|X| = n$ ,  $|Y| = m$ , and  $k = \lceil n/m \rceil$ .  
If  $f : X \rightarrow Y$  then there exist  $k$  distinct values  $x_1, x_2, \dots, x_k \in X$   
such that  $f(x_1) = f(x_2) = \dots = f(x_k)$ .

Informally: If  $n$  pigeons fly into  $m$  holes, at least one hole contains at least  $k = \lceil n/m \rceil$  pigeons.

**Proof:** Assume there's no such hole. Then there are at most  $(\lceil n/m \rceil - 1)m < (n/m)m = n$  pigeons.

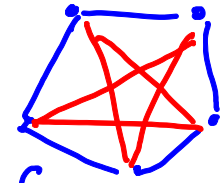
$< n/m$

# Pigeonhole Principle: Example #5



$$\lceil \frac{5}{2} \rceil = 3$$

$$R(3,3) = 6$$



## Ramsey's theorem

In any group of 6 people, where each two people are either friends or enemies (i.e., they can't be "neutral"), there must be either 3 pairwise friends or 3 pairwise enemies.

**Proof:** Let  $A$  be one of the 6 people.  $A$  has at least 3 friends or at least 3 enemies by the general pigeonhole principle because  $\lceil 5/2 \rceil = 3$ . (5 people into 2 holes (friend/enemy).)

Suppose  $A$  has  $\geq 3$  friends (the enemies case is similar) and call three of them  $B$ ,  $C$ , and  $D$ .

If  $(B, C)$  or  $(C, D)$  or  $(B, D)$  are friends then we're done because those two friends with  $A$  forms a triple of friends.

Otherwise  $(B, C)$  and  $(C, D)$  and  $(B, D)$  are enemies and  $BCD$  forms a triple of enemies.

## Pigeonhole Principle: Example #6

While on a 28-day vacation, Martina plays at least one set of tennis each day, but no more than 40 sets over all 28 days. Prove that there is a span of consecutive days in which she plays exactly 15 sets.

**Proof:** Let  $x_i$  be the total number of sets played up to and including day  $i$  (for  $i = 1, 2, \dots, 28$ ). Let  $x_0 = 0$ .

We need to show that there exist  $0 \leq i < j < 28$  such that  $x_j = x_i + 15$ .

Consider  $x_1, x_2, \dots, x_{28}, x_0 + 15, x_1 + 15, \dots, x_{27} + 15$ . These are 56 integers (pigeons) in the range  $[1, 39 + 15]$  (54 holes). Two of these integers are equal by the pigeonhole principle. Since  $x_i < x_j$  for  $i < j$  (because Martina plays  $\geq 1$  set per day), the two that are equal must be  $x_j = 15 + x_i$ . So from day  $i + 1$  to day  $j$ , Martina plays 15 sets.

# Pigeonhole Principle: Example #7

Erdős-Szekeres theorem (1935) <sup>pigeons</sup>

Any sequence  $x_1, x_2, \dots, x_n$  of  $n \geq (r-1)(s-1) + 1$  distinct numbers contains an increasing subsequence of length  $r$  or a decreasing subsequence of length  $s$ .

$r = 5$   
 $s = 5$        $n = 17$

$a_i$ : 1 2 3 1 4 4 1 3 4 2 5 5 1 2 5 5 6  
 4, 7, 12, 3, 62, 14, 2, 8, 11, 5, 20, 17, 1, 22, 15, 13, 18  
 $b_i$ : 1 1 1 2 1 2 3 3 3 4 2 3 5 2 4 5 3

in this example

Proof: Label  $x_i$  with the pair  $(a_i, b_i)$  where  $a_i$  is the length of the longest increasing subsequence ending with  $x_i$  and  $b_i$  is the length of the longest decreasing subsequence ending with  $x_i$ . No two numbers receive the same label since (for  $i < j$ ) if  $x_i < x_j$  then  $a_i < a_j$  and if  $x_i > x_j$  then  $b_i < b_j$ . If for all  $i$ ,  $a_i < r$  and  $b_i < s$ , then there are only  $(r-1)(s-1)$  labels, so by pigeonhole, two numbers receive the same label. Contradiction.

exists with same  $(a_i, b_i)$   
 $x_i \neq x_j$   
 ... contradiction  $\square$   
 PHP  $\Rightarrow$

holes

$a_i \in \{1, \dots, r-1\}$   
 $b_i \in \{1, \dots, s-1\}$   
 $(a_i, b_i) \dots (r-1)(s-1)$   
 values

$x_i < x_j$   
 $a_i < a_j$   
 $b_i > b_j$   
 $i < j$   
 should be at least  $a+1$   
 should be at least  $b+1$