# Data Structures and Algorithms Hashing 3

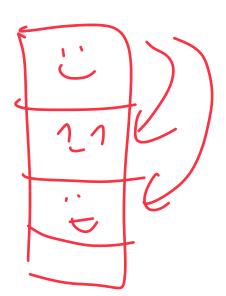
CS 225 Brad Solomon November 15, 2024



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## Learning Objectives

Review hash table implementations

Improve our closed hash implementation

Determine when and how to resize a hash table

Justify when to use different index approaches

## Simple Uniform Hashing Assumption

Given table of size m, a simple uniform hash, h, implies

$$\forall k_1, k_2 \in U \text{ where } k_1 \neq k_2 \text{ , } Pr(h[k_1] = h[k_2]) = \frac{1}{m}$$

Uniform: All keys equally likely to hash to any position

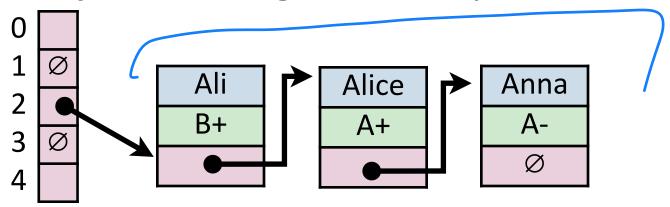
$$Pr(h[k_1]) = \frac{1}{m}$$

Independent: All key's hash values are independent of other keys

## Open vs Closed Hashing

Addressing hash collisions depends on your storage structure.

Open Hashing: store k,v pairs externally



• Closed Hashing: store k, v pairs in the hash table

0	Anna, A-
1	
2	Ali, B+
3	Alice, A+

## Separate Chaining Under SUHA



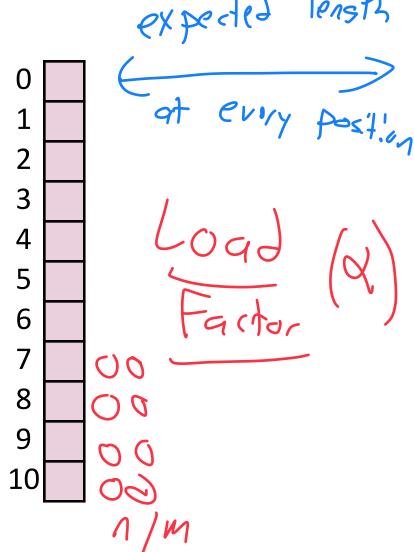
Under SUHA, a hash table of size m and n elements:

Find runs in:  $O(1+\alpha)$ .

Constant we contral

Insert runs in: O(1).

Remove runs in:  $O(1+\alpha)$ .



### Collision Handling: Linear Probing

0	22
1	8
2	16
3	29
4	4
5	11
6	13

```
h(k, i) = (k + i) \% 7

Try h(k) = (k + 0) \% 7, if full...

Try h(k) = (k + 1) \% 7, if full...

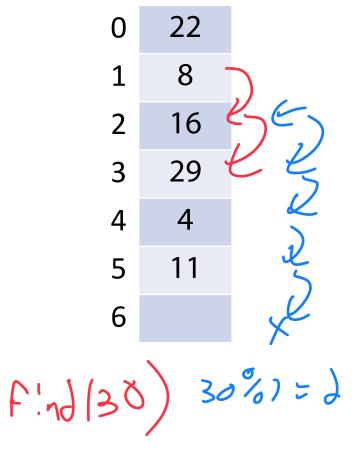
Try h(k) = (k + 2) \% 7, if full...

Try ...

Next available Space + 1
```

## Collision Handling: Linear Probing

$$S = \{ 16, 8, 4, 13, 29, 11, 22 \}$$
  $|S| = n$   
 $h(k, i) = (k + i) \% 7$   $|Array| = m$ 



### find(29)

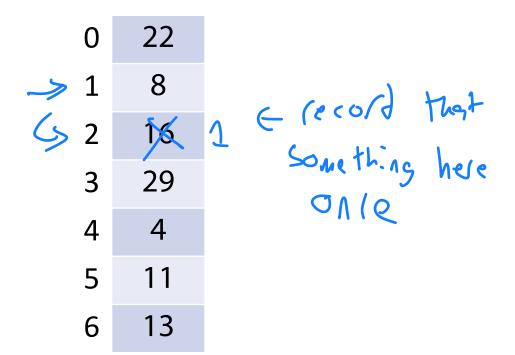
- 7 Ideal 0(1) 1) Hash the input key [h(29)=1]
- 2) Look at hash value (address) position If present, return (k, v) If not look at **next available space**

### Stop when:

- 1) We find the object we are looking for
- 2) We have searched every position in the array
- 3) We find a blank space

## Collision Handling: Linear Probing

$$S = \{ 16, 8, 4, 13, 29, 11, 22 \}$$
  $|S| = n$   
 $h(k, i) = (k + i) \% 7$   $|Array| = m$ 



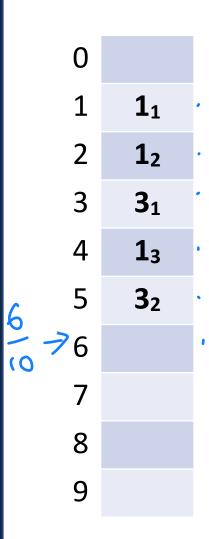
### remove (16)

- 1) Hash the input key [h(16)=2]
- 2) Find the actual location (if it exists)
- 3) Remove the (k,v) at hash value (address) Don't resize the array! Tombstone!

### A Problem w/ Linear Probing



**Primary Clustering:** "Rich get richer"



### **Description:**

Collisions create long runs of filled-in indices

Should have a 1/m chance to hash anywhere

Instead have a (size of cluster) / m chance to hash at end

If hash value is 
$$1 \rightarrow 6$$
  
 $2 \rightarrow 6$ 

### **Remedy:**

### A Problem w/ Linear Probing



**Primary Clustering:** "Rich get richer"

0	
1	11
2	12
3	31
4	1 <sub>3</sub>
5	32
6	
7	
8	
9	

### **Description:**

Collisions create long runs of filled-in indices

Should have a 1/m chance to hash anywhere

Instead have a (size of cluster) / m chance to hash at end

### Remedy:

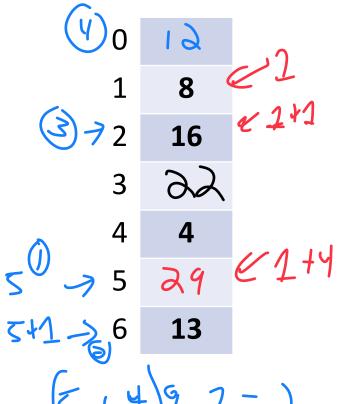
Pick a better "next available" position!

(Example of closed hashing)

1+9=10 67=3

## Collision Handling: Quadratic Probing

$$S = \{ 16, 8, 4, 13/29, 12, 22 \}$$
  
 $h(k) = k \% 7$ 



h(k, i) = (k + i\*i) % 7

Try h(k) = (k + 0) % 7, if full...

Try h(k) = (k + 1\*1) % 7, if full...

Try h(k) = (k + 2\*2) % 7, if full...

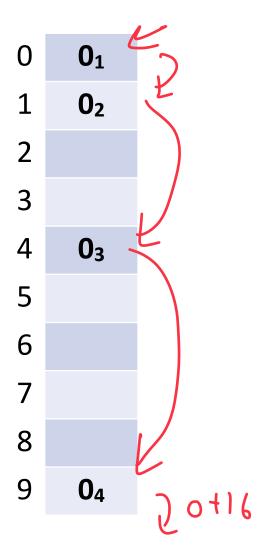
Try ...

$$3 \cdot 3$$
 $3 \cdot 3 \cdot 7 = 1$ 

$$(5+4)907=2$$
  $(5+4)907=0$ 

## A Problem w/ Quadratic Probing

Secondary Clustering: We havent solved collisions

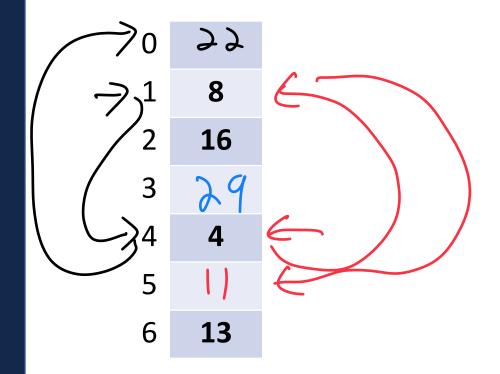


### **Description:**

De less consistent but still deterministic

(Example of closed hashing)

## Collision Handling: Double Hashing To work well



h(k, i) = (h<sub>1</sub>(k) + i\*h<sub>2</sub>(k)) % 7  
Try h(k) = (k + 0\*h<sub>2</sub>(k)) % 7, if full...  
Try h(k) = (k + 1\*h<sub>2</sub>(k)) % 7, if full...  
Try h(k) = (k + 2\*h<sub>2</sub>(k)) % 7, if full...  
Try ... 
$$\frac{29}{1}$$
  $\frac{1}{1}$   $\frac{2}{1}$   $\frac{1}{1}$   $\frac{2}{1}$   $\frac{1}{1}$   $\frac{1}{1$ 

Running Times (Understand why we have these rough forms)

(Expectation under SUHA)

**Open Hashing:** 

insert:  $\underline{\underline{1}}$ 

find/ remove: 1+4

Closed Hashing:

find/ remove:

## Running Times (Expectation under SUHA)



Open Hashing:  $0 \le \alpha \le \infty$  (Length of chain)

insert:  $\underline{\phantom{a}}$ 

find/ remove:  $1 + \alpha$ .

Closed Hashing:  $0 \le \alpha < 1$ 

insert:  $\frac{1-\alpha}{1}$ .

find/ remove:  $1 - \alpha$ 

### **Observe:**



(H° 271, rustine 7 20

### $\frac{4}{3}$ - If $\alpha$ is constant:

OH is constant 0(1)\*

## Running Times (Don't memorize these equations, no need.)

The expected number of probes for find(key) under SUHA

### Linear Probing:

- Successful:  $\frac{1}{2}(1 + \frac{1}{1-\alpha})$
- Unsuccessful:  $\frac{1}{2}(1 + \frac{1}{(1-\alpha)})^2$

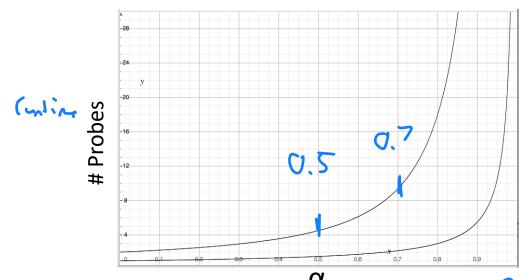
### **Double Hashing:**

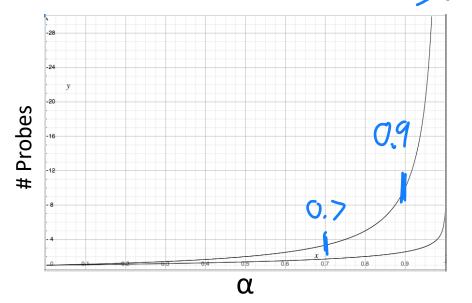
- Successful:  $1/\alpha * ln(1/(1-\alpha))$
- Unsuccessful:  $1/(1-\alpha)$

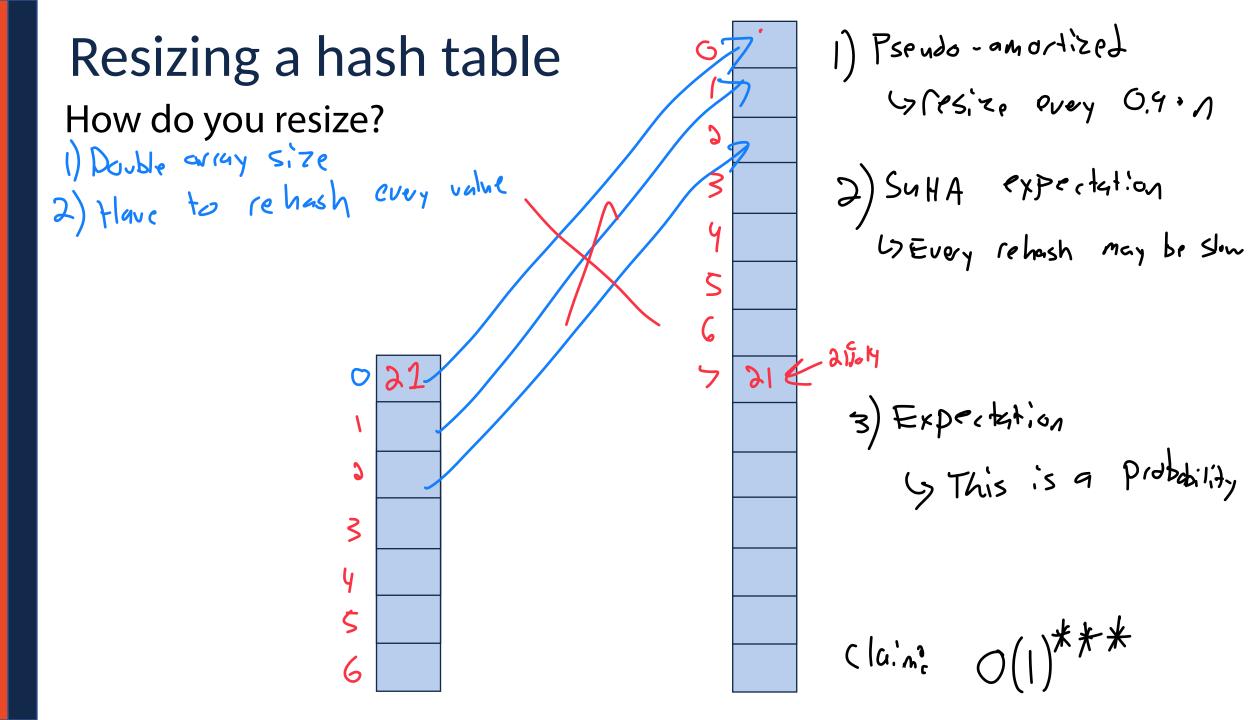


### When do we resize?

Linear ~0.7 (0.8) Double ~0.7 - (0.4)







### Which collision resolution strategy is better?



Big Records:

• Structure Speed:

What structure do hash tables implement?

What constraint exists on hashing that doesn't exist with BSTs?

Why talk about BSTs at all?

### std::map in C++

```
T& map<K, V>::operator[]
pair<iterator, bool> map<K, V>::insert()
iterator map<K, V>::erase()

iterator map<K, V>::lower_bound( const K & );
iterator map<K, V>::upper bound( const K & );
```

### std::unordered\_map in C++

```
T& unordered map<K, V>::operator[]
pair<iterator, bool> unordered map<K, V>::insert()
iterator unordered map<K, V>::erase()
iterator map<K, V>::lower bound( const K & );
iterator map<K, V>::upper bound( const K & );
float unordered map<K, V>::load factor();
void unordered map<K, V>::max load factor(float m);
```

## Running Times



	Hash Table	AVL	Linked List
Find	Expectation*: $\bigcirc(I)^{***}$ Worst Case: $\bigcirc(n)$	0(1091)	O(n)
Insert	Expectation*: $O(I)^{A \times A}$ Worst Case: $O(N)$	0(109 n)	O(1)
Storage Space	0 (n)	0(4)	0(n)

### **Bonus Slides**

### Hash Table

Worst-Case behavior is bad — but what about randomness?

1) Fix h, our hash, and assume it is good for all keys:

Simple Uniform Hashing Assumption

(Assume our dataset hashes optimally)

2) Create a *universal hash function family:* 

Given a collection of hash functions, pick one randomly

Like random quicksort if pick of hash is random, good expectation!

## Hash Function (Division Method)

Hash of form: h(k) = k % m

Pro:

Con:

### Hash Function (Mid-Square Method)

Hash of form: h(k) = (k \* k) and take b bits from middle  $(m = 2^b)$ 

## Hash Function (Mid-Square Method)

Hash of form: h(k) = (k \* k) and take b bits from middle  $(m = 2^b)$ 

## Hash Function (Multiplication Method)

Hash of form:  $h(k) = |m(kA\%1)|, 0 \le A \le 1$ 

Pro:

Con:

## Hash Function (Universal Hash Family)

Hash of form: 
$$h_{ab}(k) = \left( (ak+b) \% p \right) \% m$$
,  $a,b \in \mathbb{Z}_p^*,\mathbb{Z}_p$ 

$$\forall k_1 \neq k_2$$
,  $Pr_{a,b}(h_{ab}[k_1] = h_{ab}[k_2]) \leq \frac{1}{m}$ 

Pro:

Con: