## CSE 373: Hash Tables

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### **Announcements**

- HW 2 Due tonight (11PM)
- HW 3 out tomorrow (due July 18<sup>th</sup>, 11PM)

### **Hash Tables**

- Aim for constant-time (i.e., O(1)) find, insert, and delete
  - "On average" under some often-reasonable assumptions
- A hash table is an array of some fixed size

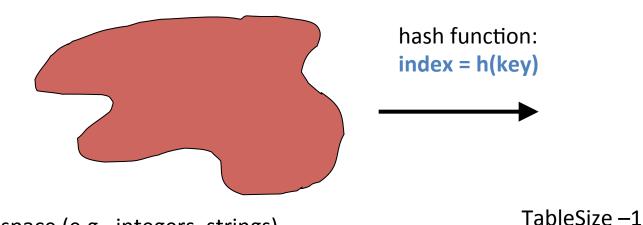
hash table 0 Basic idea: hash function: index = h(key)key space (e.g., integers, strings) TableSize -1

### Hash functions

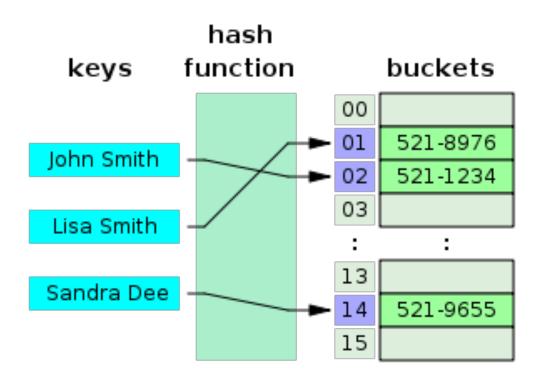
#### An ideal hash function:

- Fast to compute
- "Rarely" hashes two "used" keys to the same index
  - Often impossible in theory but easy in practice
  - Will handle collisions later

key space (e.g., integers, strings)



hash table 0



### Collision resolution

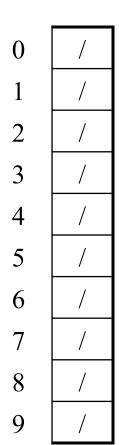
#### Collision:

When two keys map to the same location in the hash table

We try to avoid it, but number-of-keys exceeds table size

So hash tables should support collision resolution

– Ideas?

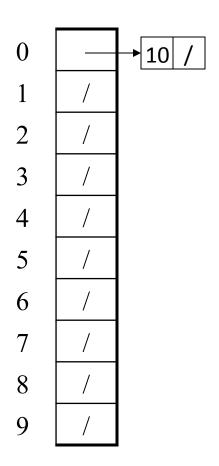


### **Chaining:**

All keys that map to the same table location are kept in a list (a.k.a. a "chain" or "bucket")

As easy as it sounds

#### **Example:**

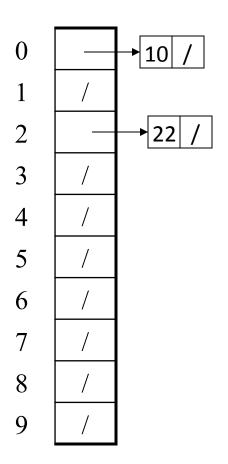


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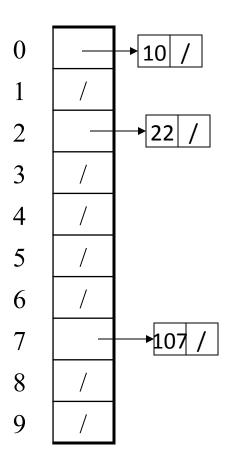


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### Example:

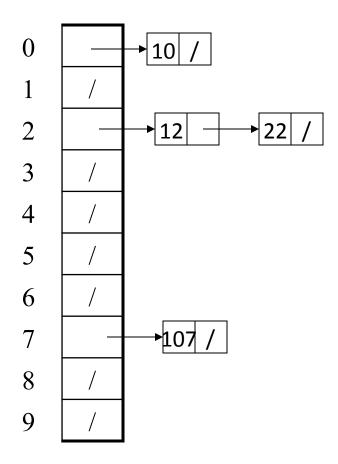


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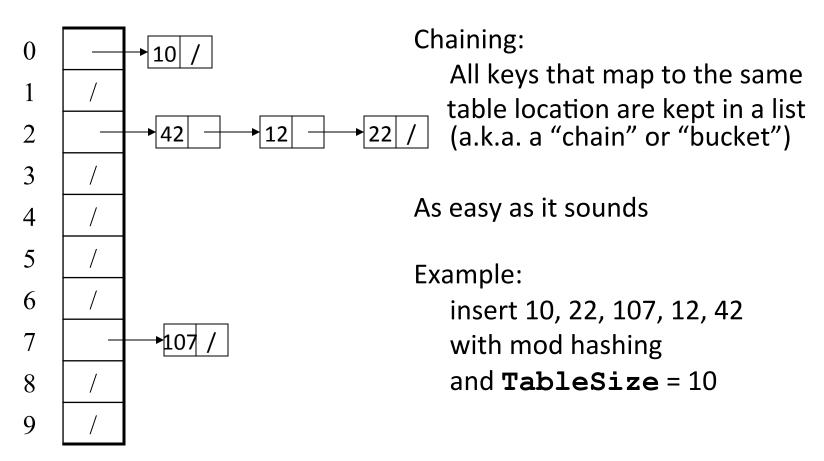


### Chaining:

All keys that map to the same table location are kept in a list (a.k.a. a "chain" or "bucket")

As easy as it sounds

### Example:



### More rigorous chaining analysis

Definition: The load factor,  $\lambda$ , of a hash table is

$$\lambda = \frac{N}{\text{TableSize}} \leftarrow \text{number of elements}$$

Under chaining, the average number of elements per bucket is  $\lambda$ 

So if some inserts are followed by *random* finds, then on average:

• Each "unsuccessful" find compares against  $\lambda$  items

So we like to keep  $\lambda$  fairly low (e.g., 1 or 1.5 or 2) for chaining

# Deleting an element using Separate Chaining

0 Another simple idea: If h (key) is already full, - try (h(key) + 1) % TableSize. If full, - try (h(key) + 2) % TableSize. If full, - try (h(key) + 3) % TableSize. If full... 3 4 Example: insert 38, 19, 8, 109, 10 5 6 38 9

- Another simple idea: If h (key) is already full,
  - try (h(key) + 1) % TableSize. If full,
  - try (h(key) + 2) % TableSize. If full,
  - try (h(key) + 3) % TableSize. If full...
- Example: insert 38, 19, 8, 109, 10

0	/
1	/
2	/
3	/
4	/
5	/
6	/
7	/
8	38

19

9

```
    Another simple idea: If h (key) is already full, 0

   - try (h(key) + 1) % TableSize. If full,
   - try (h(key) + 2) % TableSize. If full,
   - try (h(key) + 3) % TableSize. If full...
                                                     3
                                                     4
  Example: insert 38, 19, 8, 109, 10
                                                     5
                                                     6
                                                           38
                                                     9
                                                           19
```

```
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                                                     3
                                                     4
  Example: insert 38, 19, 8, 109, 10
                                                     5
                                                     6
                                                           38
                                                     9
                                                           19
```

```
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   - try (h(key) + 1) % TableSize. If full,
                                                           109
   - try (h(key) + 2) % TableSize. If full,
                                                           10
   - try (h(key) + 3) % TableSize. If full...
                                                     3
                                                     4
  Example: insert 38, 19, 8, 109, 10
                                                     5
                                                     6
                                                           38
                                                     9
                                                           19
```

### Open addressing

This is one example of open addressing

In general, open addressing means resolving collisions by trying a sequence of other positions in the table

Trying the next spot is called probing

- We just did linear probing
  - i<sup>th</sup> probe was (h(key) + i) % TableSize
- In general have some probe function f and use
  h(key) + f(i) % TableSize

Open addressing does poorly with high load factor  $\lambda$ 

- So want larger tables
- Too many probes means no more O(1)

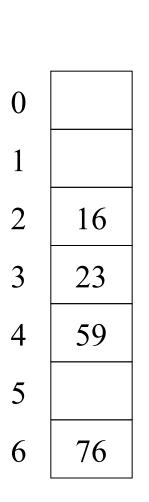
### **Open Addressing**

Write pseudocode for find(), assuming everything we've inserted is in the table.

## Deletion in open addressing

Brainstorm!

## Deletion in Open Addressing



h(k) = k % 7Linear probing

Delete(23) Find(59) Insert(30)

Need to keep track of deleted items... leave a "marker"

### **Open Addressing**

What will our pseudocode for find() look like if we're using lazy deletion?

### Other operations

insert finds an open table position using a probe
function

#### What about **find**?

- Must use same probe function to "retrace the trail" for the data
- Unsuccessful search when reach empty position

#### What about **delete**?

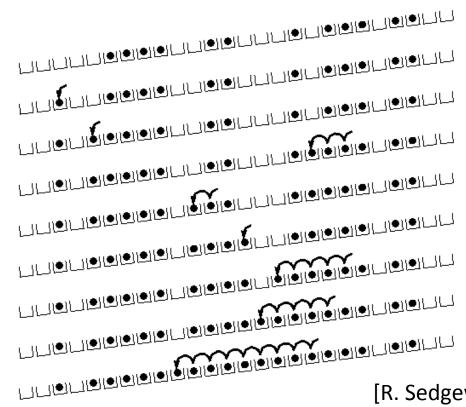
- Must use "lazy" deletion. Why?
  - Marker indicates "no data here, but don't stop probing"
- Note: delete with chaining is plain-old list-remove

## (Primary) Clustering

It turns out linear probing is a *bad idea*, even though the probe function is quick to compute (which is a good thing)

Tends to produce clusters, which lead to long probing sequences

- Called primary clustering
- Saw this starting in our example



[R. Sedgewick]

## **Analysis of Linear Probing**

- Trivial fact: For any  $\lambda < 1$ , linear probing will find an empty slot
  - It is "safe" in this sense: no infinite loop unless table is full
- Non-trivial facts we won't prove:

Average # of probes given  $\lambda$  (in the limit as **TableSize**  $\rightarrow \infty$ )

– Unsuccessful search:

Successful search:

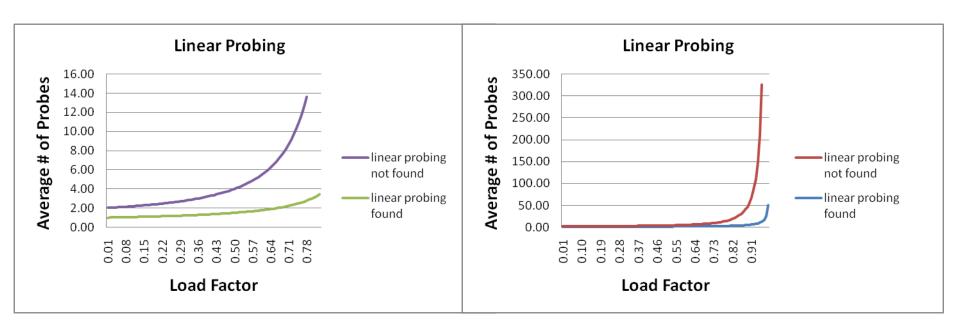
$$\frac{1}{2}\left(1+\frac{1}{(1-\lambda)^2}\right)$$

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

 This is pretty bad: need to leave sufficient empty space in the table to get decent performance

### In a chart

- Linear-probing performance degrades rapidly as table gets full
  - (Formula assumes "large table" but point remains)



• By comparison, chaining performance is linear in  $\lambda$  and has no trouble with  $\lambda > 1$ 

## Quadratic probing

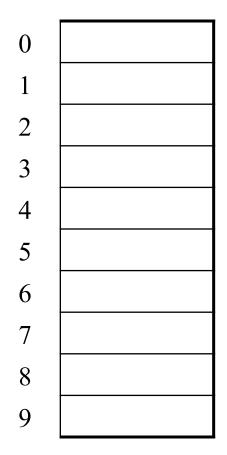
We can avoid primary clustering by changing the probe function

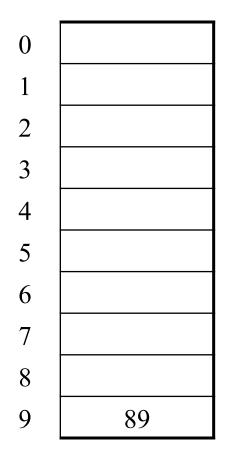
```
(h(key) + f(i)) % TableSize
```

A common technique is quadratic probing:

```
f(i) = i^2
```

- So probe sequence is:
  - 0<sup>th</sup> probe: h(key) % TableSize
  - 1st probe: (h(key) + 1) % TableSize
  - 2<sup>nd</sup> probe: (h(key) + 4) % TableSize
  - 3<sup>rd</sup> probe: (h(key) + 9) % TableSize
  - ...
  - ith probe: (h(key) + i<sup>2</sup>) % TableSize
- Intuition: Probes quickly "leave the neighborhood"





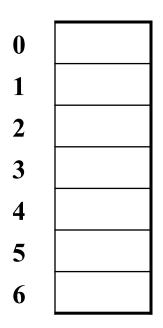
0	
1	
2	
2 3	
4	
4 5 6	
6	
7	
8	18
9	89

0	49
1	
2	
2 3	
4	
4 5 6	
6	
7	
8	18
9	89

0	49
1	
2	58
2 3	
4	
4 5 6	
6	
7	
8	18
8 9	89

0	49
1	
2	58
2 3	79
4	
4 5 6	
6	
7	
8	18
9	89

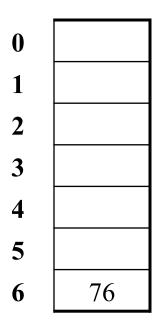
### Another Quadratic Probing Example



Insert:

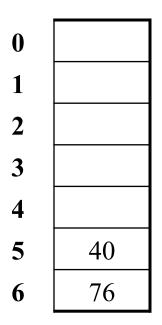
$$5 \qquad (5\%7=5)$$

$$(55 \% 7 = 6)$$



Insert:

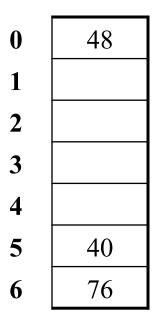
$$5 (5\%7 = 5)$$



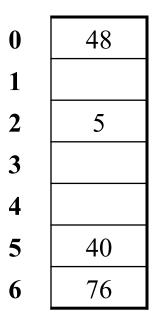
(47 % 7 = 5)

TableSize = 7

47



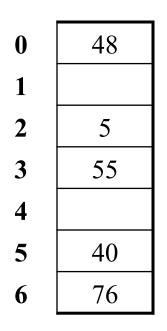
Insert:



TableSize = 7

0	48
1	
2	5
3	55
4	
5	40
6	76

Insert:



TableSize = 7

Doh!: For all n, ((n\*n) +5) % 7 is 0, 2, 5, or 6

- Excel shows takes "at least" 50 probes and a pattern
- Proof uses induction and  $(n^2+5) \% 7 = ((n-7)^2+5) \% 7$ 
  - In fact, for all c and k,  $(n^2+c)$  % k =  $((n-k)^2+c)$  % k

#### From Bad News to Good News

#### Bad news:

 Quadratic probing can cycle through the same full indices, never terminating despite table not being full

#### Good news:

- If **TableSize** is *prime* and  $\lambda < \frac{1}{2}$ , then quadratic probing will find an empty slot in at most **TableSize/2** probes
- So: If you keep  $\lambda < \frac{1}{2}$  and **TableSize** is *prime*, no need to detect cycles
- Optional
  - Also, slightly less detailed proof in textbook
  - Key fact: For prime  $\mathbf{T}$  and  $\mathbf{0} < \mathbf{i}, \mathbf{j} < \mathbf{T}/2$  where  $\mathbf{i} \neq \mathbf{j}$ ,  $(\mathbf{k} + \mathbf{i}^2)$  %  $\mathbf{T} \neq (\mathbf{k} + \mathbf{j}^2)$  %  $\mathbf{T}$  (i.e., no index repeat)

## Quadratic Probing: Success guarantee for $\lambda < \frac{1}{2}$

Assertion #1: If T = TableSize is **prime** and  $\lambda < \frac{1}{2}$ , then quadratic probing will find an empty slot in  $\leq$  T/2 probes

Assertion #2: For prime T and all  $0 \le i,j \le T/2$  and  $i \ne j$ ,  $(h(K) + i^2) % T \ne (h(K) + j^2) % T$ 

Assertion #3: Assertion #2 proves assertion #1.

### Quadratic Probing: Success guarantee for $\lambda < \frac{1}{2}$

We can prove assertion #2 by contradiction. Suppose that for some  $i \neq j$ ,  $0 \leq i, j \leq T/2$ , prime T:  $(h(K) + i^2) % T = (h(K) + j^2) % T$ 

## Clustering reconsidered

- Quadratic probing does not suffer from primary clustering: no problem with keys initially hashing to the same neighborhood
- But it's no help if keys initially hash to the same index
  - Called secondary clustering
- Can avoid secondary clustering with a probe function that depends on the key: double hashing...

## Double hashing

#### Idea:

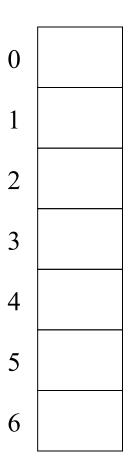
- Given two good hash functions h and g, it is very unlikely that for some key, h(key) == g(key)
- So make the probe function f(i) = i\*g(key)

#### Probe sequence:

```
0<sup>th</sup> probe: h(key) % TableSize
1<sup>st</sup> probe: (h(key) + g(key)) % TableSize
2<sup>nd</sup> probe: (h(key) + 2*g(key)) % TableSize
3<sup>rd</sup> probe: (h(key) + 3*g(key)) % TableSize
...
i<sup>th</sup> probe: (h(key) + i*g(key)) % TableSize
```

Detail: Make sure g (key) cannot be 0

## Double Hashing Example



TableSize = 7  

$$h(K) = K \% 7$$
  
 $g(K) = 5 - (K \% 5)$ 

## Double-hashing analysis

- Intuition: Because each probe is "jumping" by g (key)
  each time, we "leave the neighborhood" and "go
  different places from other initial collisions"
- But we could still have a problem like in quadratic probing where we are not "safe" (infinite loop despite room in table)
  - It is known that this cannot happen in at least one case:
    - h(key) = key % p
    - g(key) = q (key % q)
    - 2 < q < p
    - p and q are prime

## More double-hashing facts

- Assume "uniform hashing"
  - Means probability of g(key1) % p == g(key2) % p is 1/p
- Non-trivial facts we won't prove:

Average # of probes given  $\lambda$  (in the limit as **TableSize**  $\rightarrow \infty$ )

- Unsuccessful search (intuitive):  $\frac{1}{1-\lambda}$ 

- Successful search (less intuitive):  $\frac{1}{\lambda} \log_e \left( \frac{1}{1 \lambda} \right)$
- Bottom line: unsuccessful bad (but not as bad as linear probing), but successful is not nearly as bad

## Rehashing

- As with array-based stacks/queues/lists, if table gets too full, create a bigger table and copy everything
- With chaining, we get to decide what "too full" means
  - Keep load factor reasonable (e.g., < 1)?</p>
  - Consider average or max size of non-empty chains?
- For open addressing, half-full is a good rule of thumb
- New table size
  - Twice-as-big is a good idea, except, uhm, that won't be prime!
  - So go about twice-as-big
  - Can have a list of prime numbers in your code since you won't grow more than 20-30 times

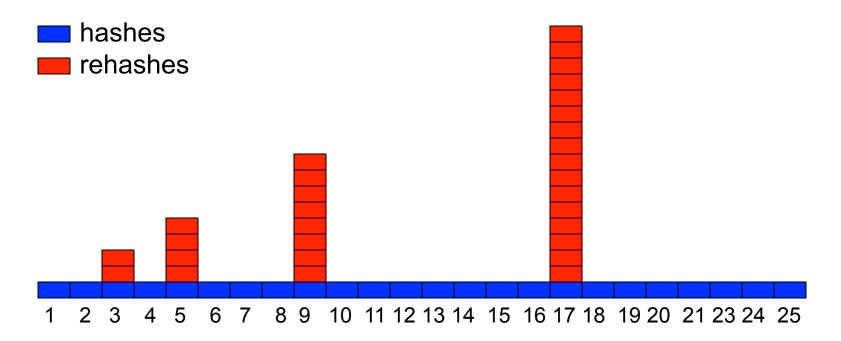
## Rehashing

When the table gets too full, create a bigger table (usually 2x as large) and hash all the items from the original table into the new table.

- When to rehash?
  - Separate chaining: full ( $\lambda$  = 1)
  - Open addressing: half full ( $\lambda = 0.5$ )
  - When an insertion fails
  - Some other threshold
- Cost of a single rehashing?

# Rehashing Picture

 Starting with table of size 2, double when load factor > 1.



# **Amortized Analysis of Rehashing**

- Cost of inserting n keys is < 3n</li>
- suppose  $2^{k} + 1 \le n \le 2^{k+1}$ 
  - Hashes = n
  - Rehashes =  $2 + 2^2 + ... + 2^k = 2^{k+1} 2^k$
  - Total = n +  $2^{k+1}$  2 < 3n

#### Example

$$- n = 33$$
, Total =  $33 + 64 - 2 = 95 < 99$ 

## **Terminology**

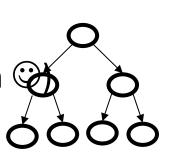
#### We and the book use the terms

- "chaining" or "separate chaining"
- "open addressing"

#### Very confusingly,

- "open hashing" is a synonym for "chaining"
- "closed hashing" is a synonym for "open addressing"

(If it makes you feel any better, most trees in CS grow upside-down @





### Equal objects must hash the same

• The Java library (and your project hash table) make a very important assumption that clients must satisfy...

```
If c.compare(a,b) == 0, then we require
h.hash(a) == h.hash(b)
```

- If you ever override equals
  - You need to override hashCode also in a consistent way
  - See CoreJava book, Chapter 5 for other "gotchas" with equals

# **Hashing Summary**

- Hashing is one of the most important data structures.
- Hashing has many applications where operations are limited to find, insert, and delete.
  - But what is the cost of doing, e.g., findMin?
- Can use:
  - Separate chaining (easiest)
  - Open hashing (memory conservation, no linked list management)
  - Java uses separate chaining
- Rehashing has good amortized complexity.
- Also has a big data version to minimize disk accesses: extendible hashing. (See book.)