

# CSE 332 Winter 2026

## Lecture 10: hashing

Nathan Brunelle

<http://www.cs.uw.edu/332>

# Dictionary (Map) ADT

- Contents:
  - Sets of key+value pairs
  - Keys must be comparable
- Operations:
  - **insert(key, value)**
    - Adds the (key,value) pair into the dictionary
    - If the key already has a value, overwrite the old value
      - Consequence: Keys cannot be repeated
  - **find(key)**
    - Returns the value associated with the given key
  - **delete(key)**
    - Remove the key (and its associated value)

# Dictionary Data Structures

Data Structure	Time to insert	Time to find	Time to delete
Unsorted Array	$\Theta(1)$	$\Theta(n)$	$\Theta(n)$
Unsorted Linked List	$\Theta(1)$	$\Theta(n)$	$\Theta(n)$
Sorted Array	$\Theta(n)$	$\Theta(\log n)$	$\Theta(n)$
Sorted Linked List	$\Theta(n)$	$\Theta(n)$	$\Theta(n)$
Heap	$\Theta(\log n)$	$\Theta(n)$	$\Theta(n)$
Binary Search Tree	$\Theta(\text{height})$	$\Theta(\text{height})$	$\Theta(\text{height})$
AVL Tree	$\Theta(\log n)$	$\Theta(\log n)$	$\Theta(\log n)$

Annotations:

- A red curly arrow points from the top-left corner of the table towards the "Data Structure" header.
- A red arrow points from the bottom-left corner of the table towards the "Time to insert" header.
- A red oval highlights the "Time to insert" value for the Binary Search Tree row.
- A red arrow points upwards from the bottom of the "Time to insert" column for the AVL Tree row, pointing towards the "Time to insert" header.

# BSTs and AVL Trees

- Binary Search Tree:
  - A binary tree where for each node, all keys in its left subtree are smaller and all keys in its right subtree are larger
  - Find:
    - If it matches, return the value.
    - If the search key is less than the current node, look left. If it's greater, look right.
    - If we reach an empty spot, find was unsuccessful
  - Insert:
    - Do a find, if it was successful then update the value
    - If it was unsuccessful, add a new node to the empty spot we found.
  - Delete:
    - If the deleted node is a leaf, just remove it
    - If the deleted node had one child, replace it with that one child
    - If the deleted node had 2 children, replace it with the largest key to the left
- AVL Tree:
  - A binary search tree where for each node, the height of its left subtree and the height of its right subtree are off by at most 1.
  - Find:
    - Same as BST
  - Insert:
    - Do a BST insert, then rotate if tree is unbalanced (apply one LL, RR, LR, RL case)
  - Delete:
    - Do a BST delete, then rotate if the tree is unbalanced (apply LL, RR, LR, RL cases as needed from leaf to root)

# Other Tree-based Dictionaries

- Red-Black Trees
  - Similar to AVL Trees in that we add shape rules to BSTs
  - More “relaxed” shape than an AVL Tree
    - Trees can be taller (though not asymptotically so)
    - Needs to move nodes less frequently
  - This is what Java’s TreeMap uses!
- Tries
  - Similar to a Huffman Tree
  - Requires keys to be sequences (e.g. Strings)
  - Combines shared prefixes among keys to save space
  - Often used for text-based searches
    - Web search
    - Genomes

# Next topic: Hash Tables

Data Structure	Time to insert	Time to find	Time to delete
Unsorted Array	$\Theta(n)$	$\Theta(n)$	$\Theta(n)$
Unsorted Linked List	$\Theta(n)$	$\Theta(n)$	$\Theta(n)$
Sorted Array	$\Theta(n)$	$\Theta(\log n)$	$\Theta(n)$
Sorted Linked List	$\Theta(n)$	$\Theta(n)$	$\Theta(n)$
Binary Search Tree	$\Theta(\text{height})$	$\Theta(\text{height})$	$\Theta(\text{height})$
AVL Tree	$\Theta(\log n)$	$\Theta(\log n)$	$\Theta(\log n)$
Hash Table (Worst case)	$\Theta(n)$	$\Theta(n)$	$\Theta(n)$
Hash Table (Average)	$\Theta(1)$	$\Theta(1)$	$\Theta(1)$



# Dictionary (Map) ADT

- Contents:
  - Sets of key+value pairs
  - ~~Keys must be comparable~~ Keys have a hash function
- Operations:
  - **insert(key, value)**
    - Adds the (key,value) pair into the dictionary
    - If the key already has a value, overwrite the old value
      - Consequence: Keys cannot be repeated
  - **find(key)**
    - Returns the value associated with the given key
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# The Best Dictionary Data Structure!

- Think of every key as a number
- Give each key its own index in an array

```
insert(key, value){  
    arr[key]=value;  
}  
find(key){  
    return arr[key];  
}  
delete(key){  
    arr[key] = null;  
}
```

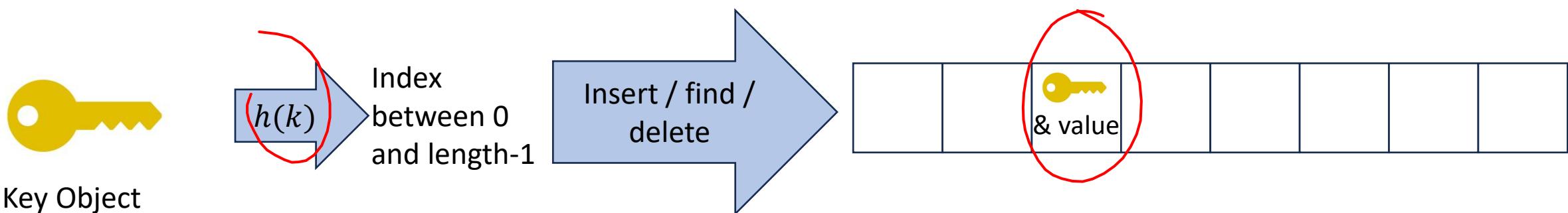


# Problem?

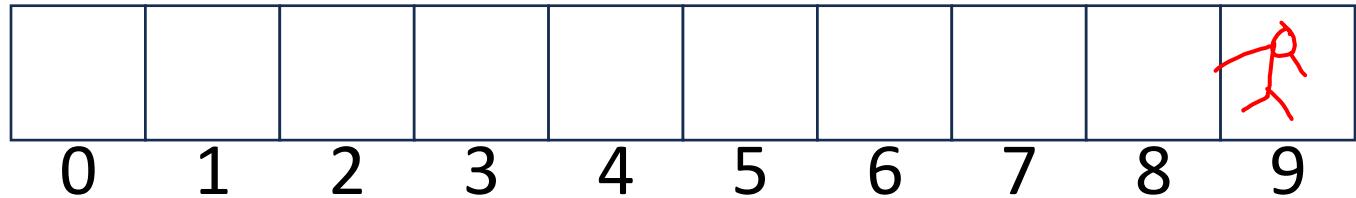


# Hash Tables

- Idea:
  - Have a small array to store information
  - Use a **hash function** to convert the key into an index
    - Hash function should “scatter” the keys, behave as if it randomly assigned keys to indices
  - Store key at the index given by the hash function
  - Do something if two keys map to the same place (should be very rare)
    - Collision resolution



# Example



- Key: Phone Number
- Value: People
- Table size: 10
- $h(phone) = \text{number as an integer \% } 10$
- $h(8675309) = 9$

$\curvearrowleft$

A red curly brace is drawn under the expression  $h(8675309) = 9$ , grouping the entire equation.

# What Influences Running time?

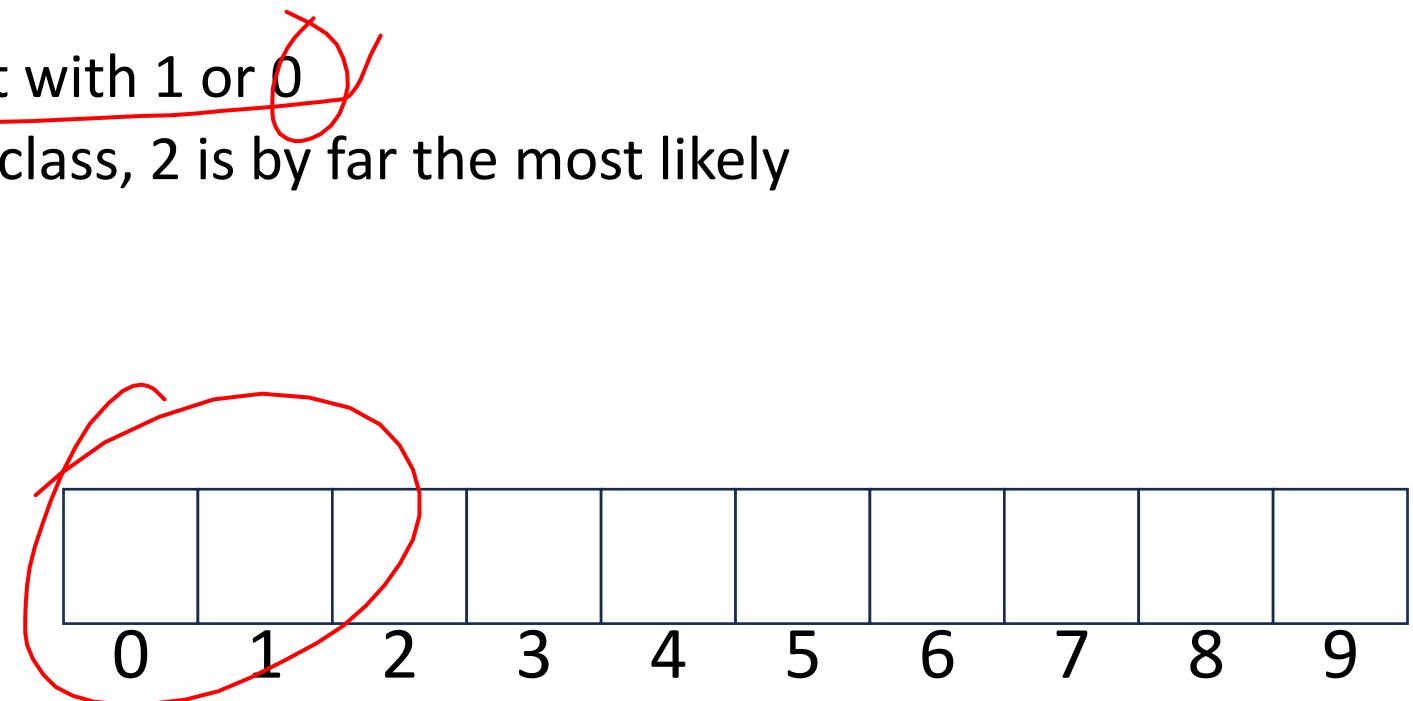
- How long hashing itself takes
- Likelihood of collisions
  - Size of the array vs number of values in the array
  - “quality” of our hash function
- What we do when we have a collision

# Properties of a “Good” Hash

- Definition: A hash function maps objects to integers
- **Consistent**
  - Objects considered “equal” should hash to the same value
  - Deterministic: running the hash function on the same object twice should yield the same result
- **Uniform**
  - Should be able to use every index in a fixed-size array
  - Should use every index at roughly equal rates
- **Effective**
  - It should be difficult to find two objects which hash to the same value
  - Given an object, it should be hard to find a different object which hashes to the same value
  - “Avalanche effect”: making a small change to the object yields big changes in the value it hashes to
- **Efficient**
  - Time to calculate the hash should be very small

# A Bad Hash (and phone number trivia)

- $h(phone)$  = the first digit of the phone number
  - Assume 10-digit format
  - ~~No US phone numbers start with 1 or 0~~
  - If we're sampling from this class, 2 is by far the most likely
- Consistent? Yes!
- Uniform? No!
- Effective? No!
- Efficient? Yes!



# Compare These Hash Functions (for strings)

- Let  $s = s_0 s_1 s_2 \dots s_{m-1}$  be a string of length  $m$ 
  - Let  $a(s_i)$  be the ascii encoding of the character  $s_i$
- $h_1(s) = a(s_0)$

$$\bullet h_2(s) = (\sum_{i=0}^{m-1} a(s_i))$$

$$\bullet h_3(s) = (\sum_{i=0}^{m-1} a(s_i) \cdot 37^i)$$

$$\bullet h_4(s) = (2 \cdot \sum_{i=0}^{m-1} a(s_i) \cdot 37^i)$$

P - m  
x =  $(a(s_i) \cdot x)$  } >

# Compare These Hash Functions (for strings)

- Let  $s = s_0s_1s_2 \dots s_{m-1}$  be a string of length  $m$ 
  - Let  $a(s_i)$  be the ascii encoding of the character  $s_i$
- $h_1(s) = a(s_0)$ 
  - Is: consistent, efficient
- $h_2(s) = (\sum_{i=0}^{m-1} a(s_i))$ 
  - Is: consistent, efficient, and possibly uniform
- $h_3(s) = (\sum_{i=0}^{m-1} a(s_i) \cdot 37^i)$ 
  - Is: Consistent, efficient, uniform, and effective
- $h_4(s) = (2 \cdot \sum_{i=0}^{m-1} a(s_i) \cdot 37^i)$ 
  - Is: Consistent, efficient, effective



# Ideal Insert procedure

Supposing we have a “good” hash function:

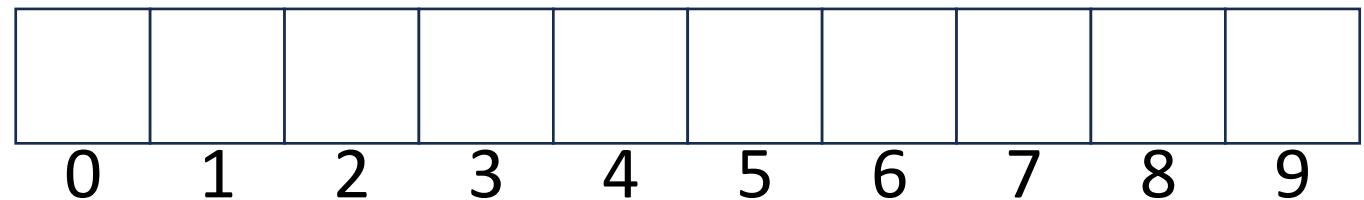
```
insert(key, value){  
    h = key.hash();  
    arr[h % table.length] = value;  
}
```

*arr*

Problem: It's possible that two different keys map to the same index!  
This is called a “collision”

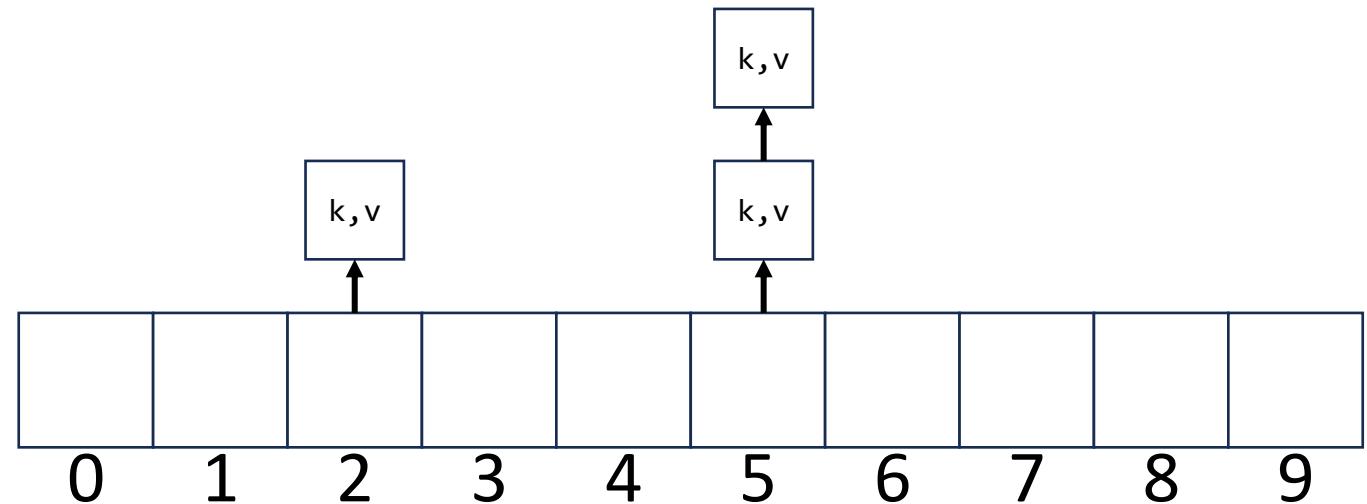
# Collision Resolution

- A Collision occurs when we want to insert something into an already-occupied position in the hash table
- 2 main strategies:
  - Separate Chaining
    - Use a secondary data structure to contain the items
      - E.g. each index in the hash table is itself a linked list
  - Open Addressing
    - Use a different spot in the table instead
      - Linear Probing
      - Quadratic Probing
      - Double Hashing



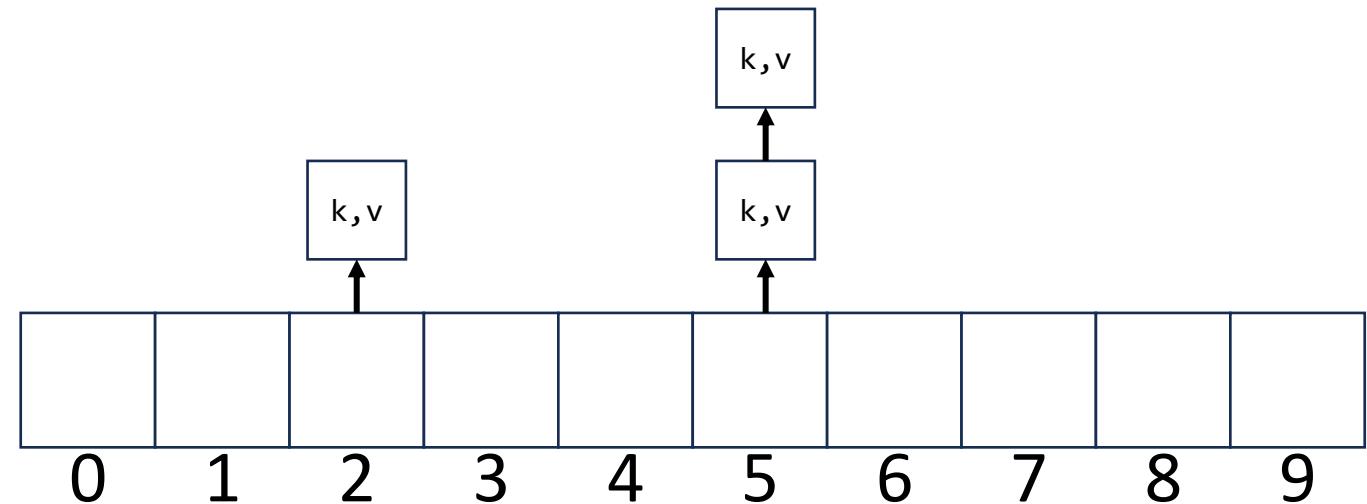
# Separate Chaining Insert

- To insert  $k, v$ :
  - Compute the index using  $i = h(k) \% \text{table.length}$
  - Add the key-value pair to the data structure at  $\text{table}[i]$



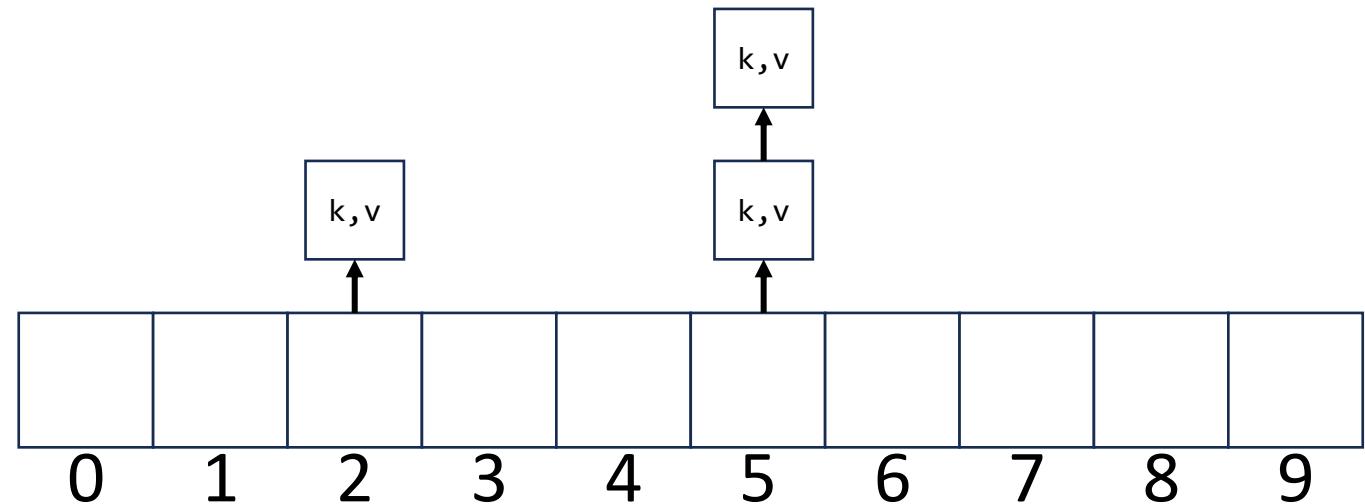
# Separate Chaining Find

- To find  $k$ :
  - Compute the index using  $i = h(k) \% \text{table.length}$
  - Call find with the key on the data structure at  $\text{table}[i]$



# Separate Chaining Delete

- To delete  $k$ :
  - Compute the index using  $i = h(k) \% \text{table.length}$
  - Call delete with the key on the data structure at  $\text{table}[i]$



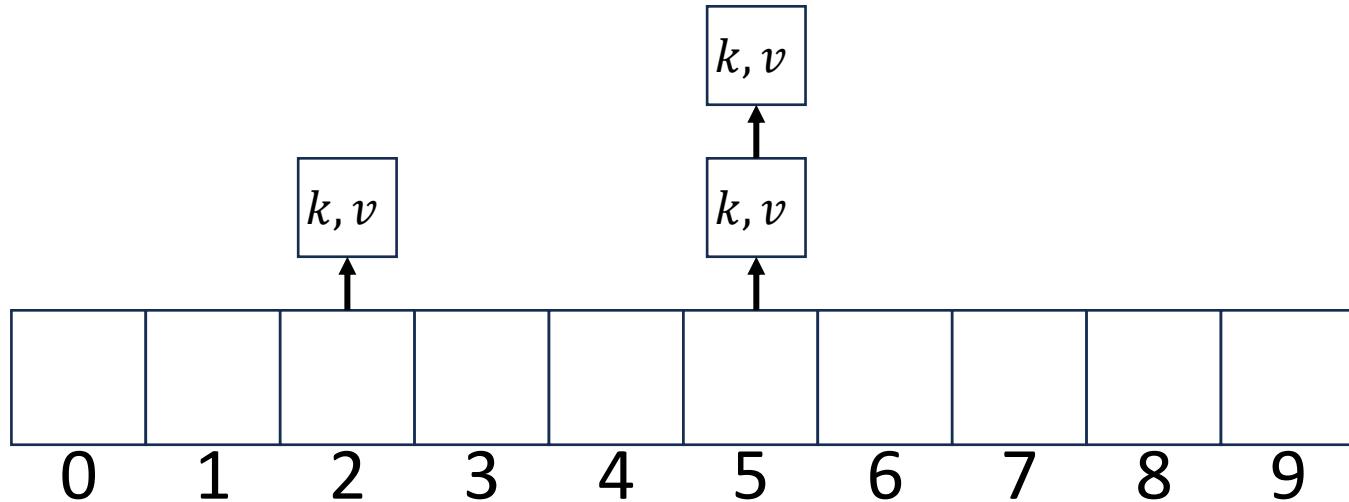
# Formal Running Time Analysis

- The **load factor** of a hash table represents the average number of items per “bucket”
  - $\lambda = \frac{n}{length}$
- Assume we have a hash table that uses a linked-list for separate chaining
  - What is the expected number of comparisons needed in an unsuccessful find?
  - What is the expected number of comparisons needed in a successful find?
- How can we make the expected running time  $\Theta(1)$ ?

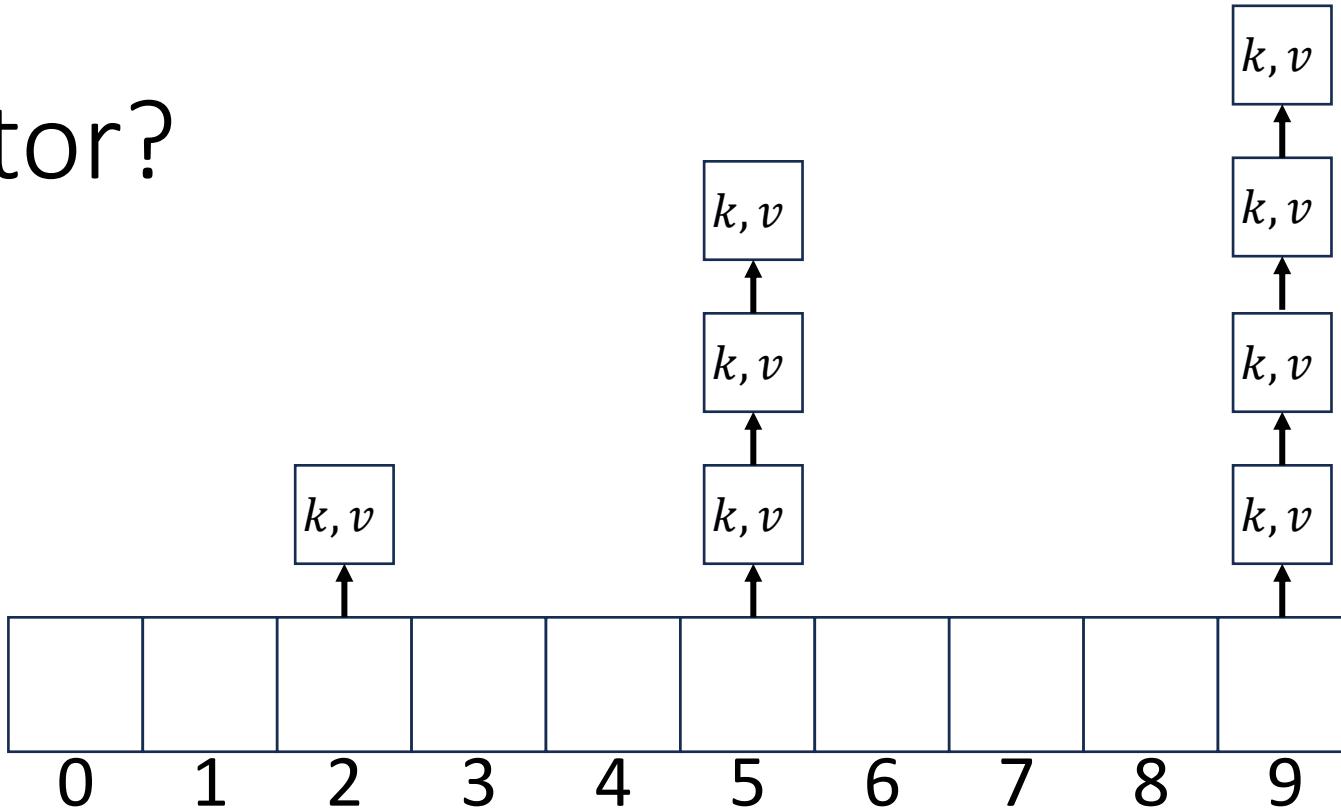
# Formal Running Time Analysis

- The **load factor** of a hash table represents the average number of items per “bucket”
  - $\lambda = \frac{n}{length}$
- Assume we have a hash table that uses a linked-list for separate chaining
  - What is the expected number of comparisons needed in an unsuccessful find?
    - $\lambda$
  - What is the expected number of comparisons needed in a successful find?
    - $\frac{\lambda}{2}$
- How can we make the expected running time  $\Theta(1)$ ?
  - Pick a constant value, resize the array whenever  $\lambda$  exceeds that constant
    - We'll talk about which constant we should pick later

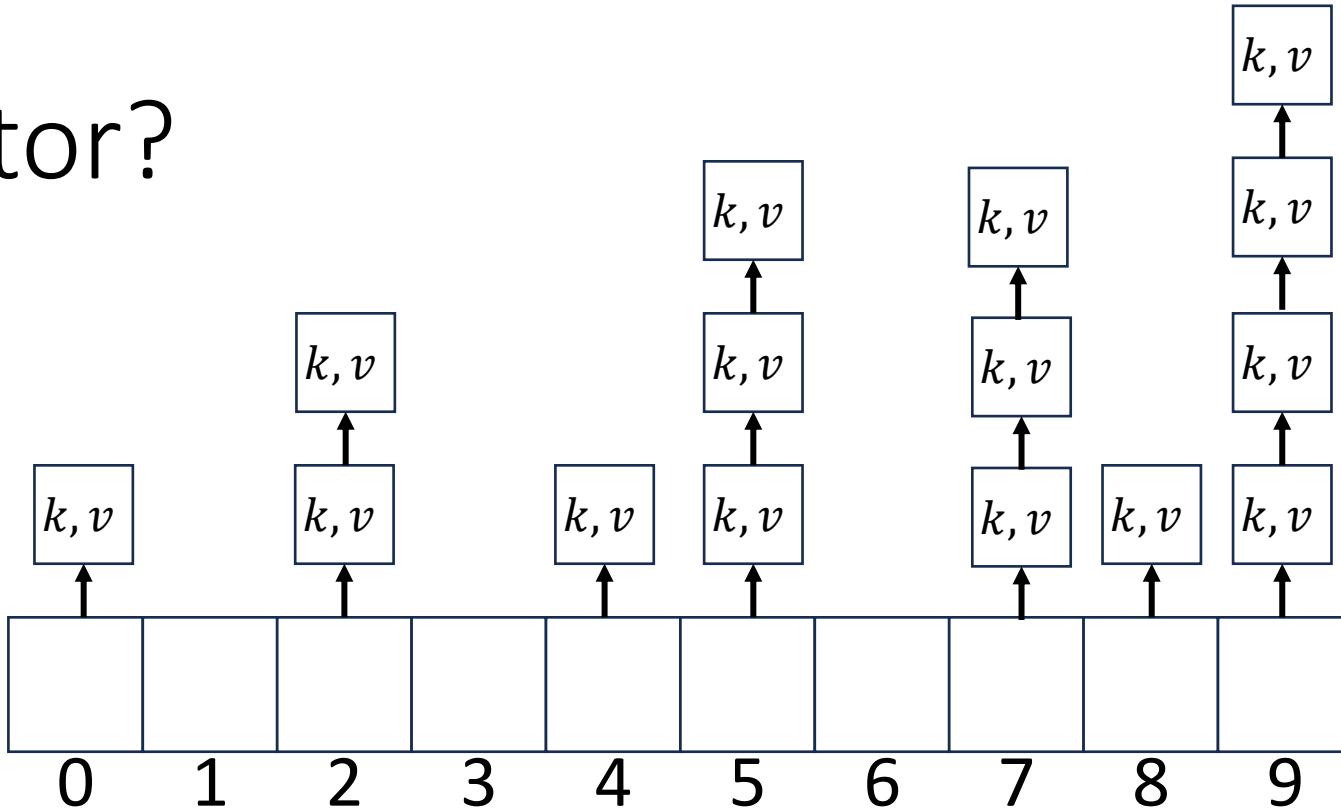
# Load Factor?



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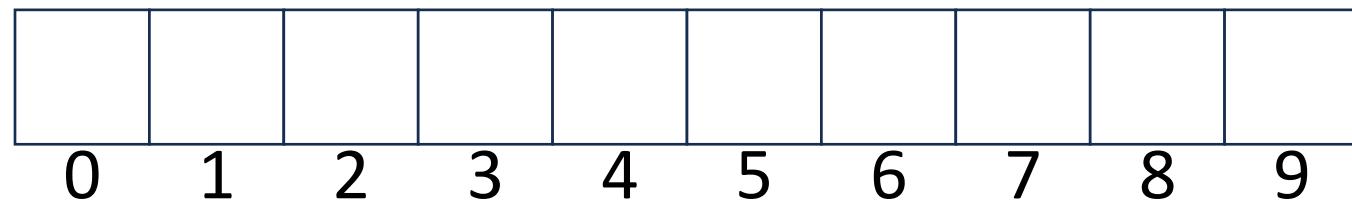


# Load Factor?



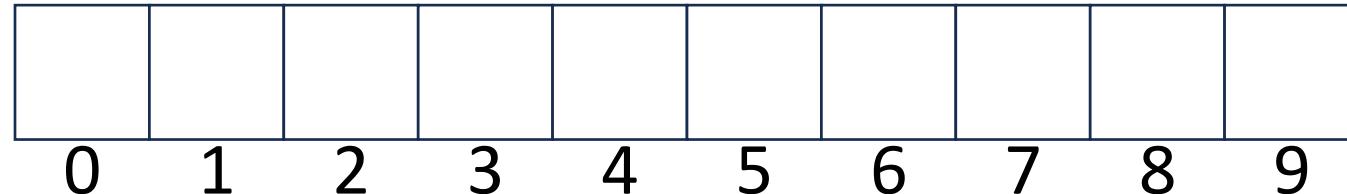
# Collision Resolution: Linear Probing

- When there's a collision, use the next open space in the table

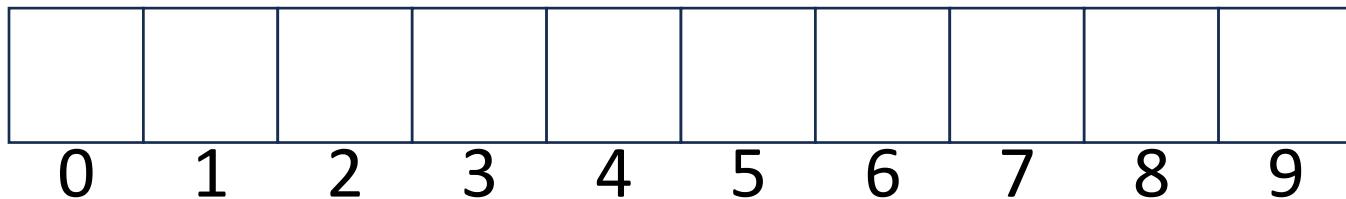


# Linear Probing: Insert Procedure

- To insert  $k, v$ 
  - Calculate  $i = h(k) \% \text{table.length}$
  - If  $\text{table}[i]$  is occupied then try index  $(i+1) \% \text{table.length}$
  - If that is occupied try index  $(i+2) \% \text{table.length}$
  - If that is occupied try index  $(i+3) \% \text{table.length}$
  - ...

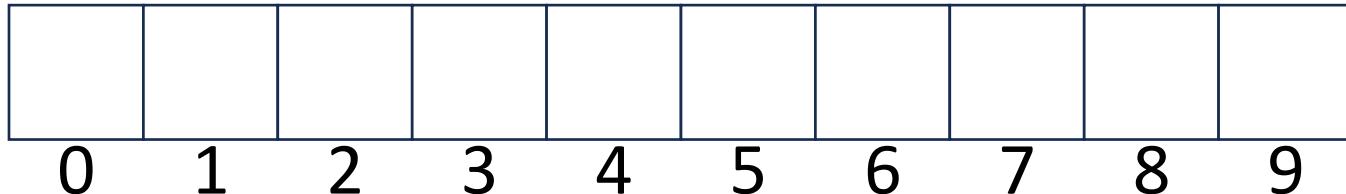


# Linear Probing: Find



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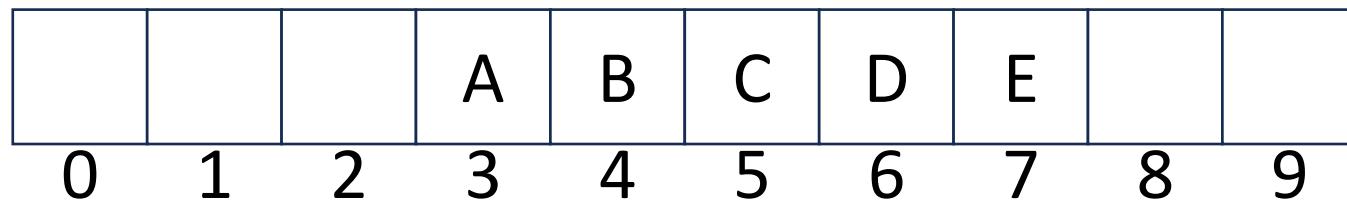
- To find key  $k$ 
  - Calculate  $i = h(k) \% \text{table.length}$
  - If  $\text{table}[i]$  is occupied but doesn't have  $k$ , check  $(i+1) \% \text{table.length}$
  - If that is occupied and doesn't contain  $k$ , check  $(i+2) \% \text{table.length}$
  - If that is occupied and doesn't contain  $k$ , check  $(i+3) \% \text{table.length}$
  - Repeat until you either find  $k$  or else you reach an empty cell in the table



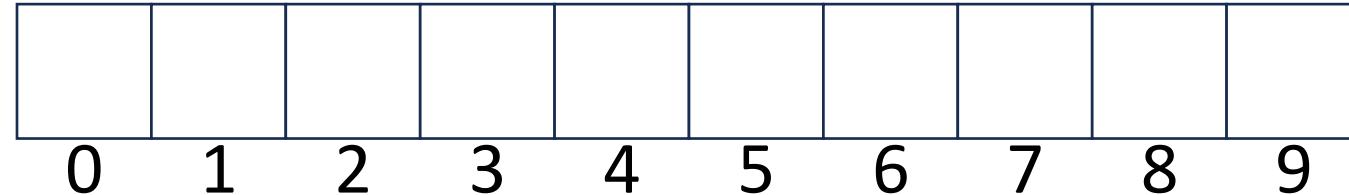
# Linear Probing: Delete

- Suppose A, B, C, D, and E all hashed to 3
- Now let's delete B

Before:



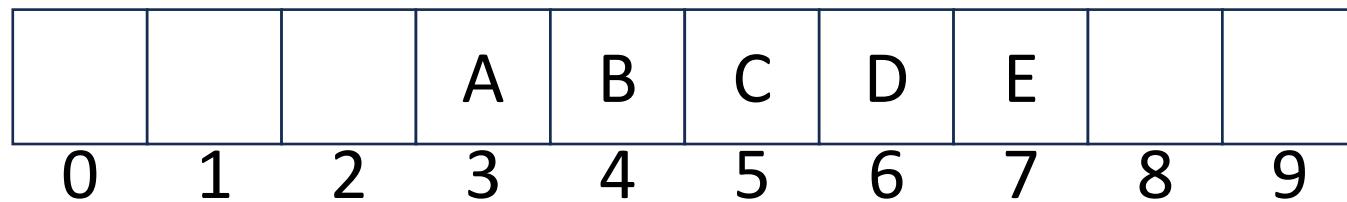
After:



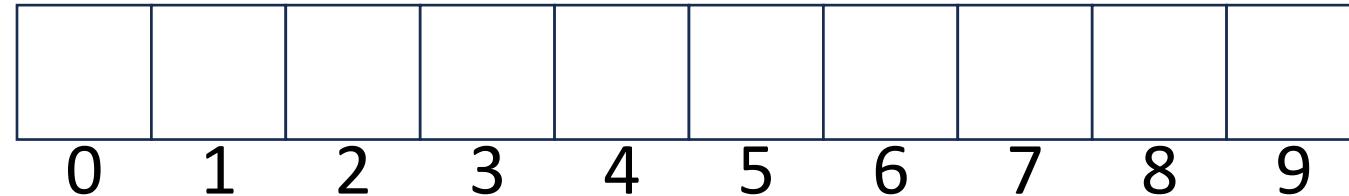
# Linear Probing: Delete

- Suppose A, B, and E all hashed to 3, and C and D hashed to 5
- Now let's delete B

Before:



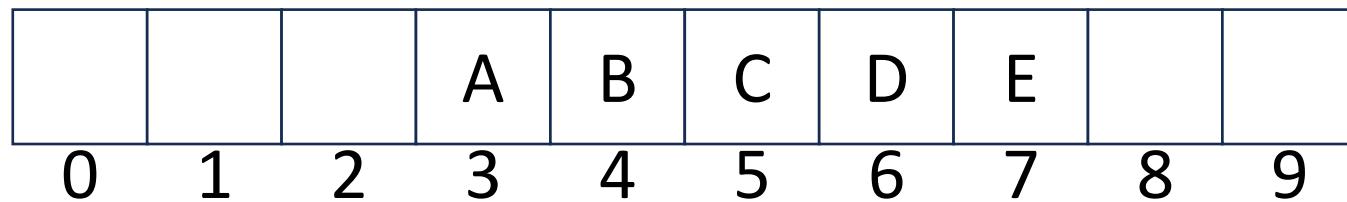
After:



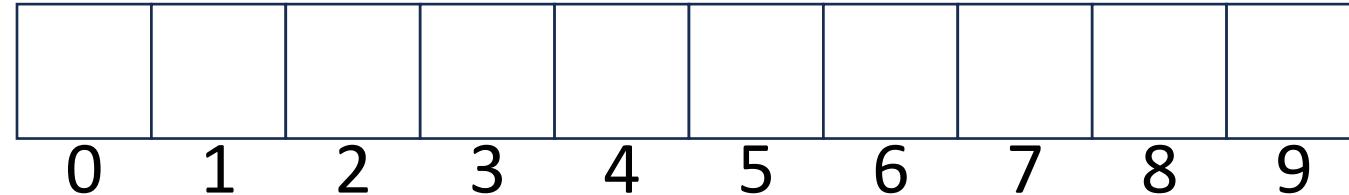
# Linear Probing: Delete

- Suppose A and E hashed to 3, and B,C, and D hashed to 4
- Now let's delete B

Before:

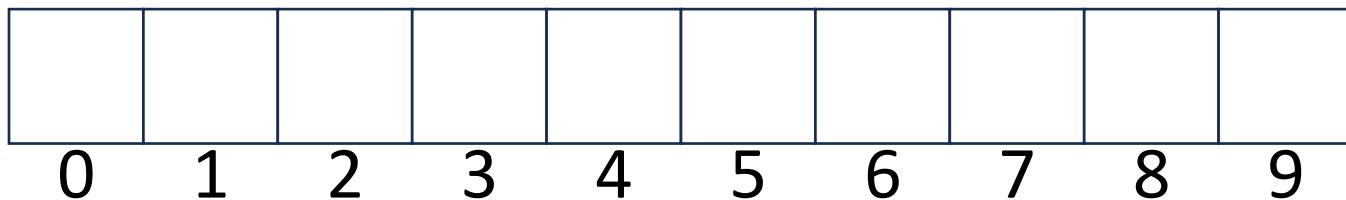


After:



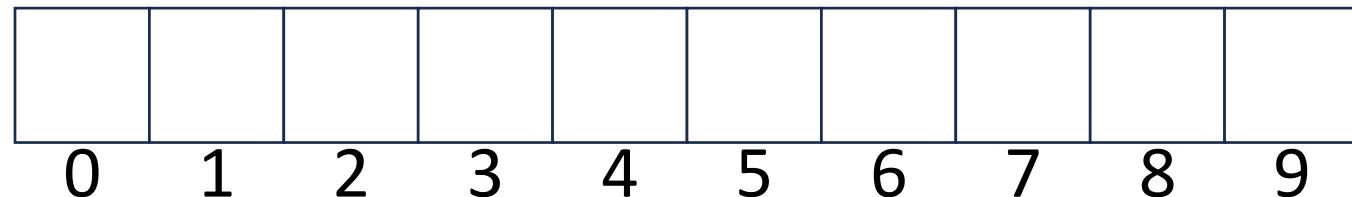
# Linear Probing: Delete

- Let's do this together!



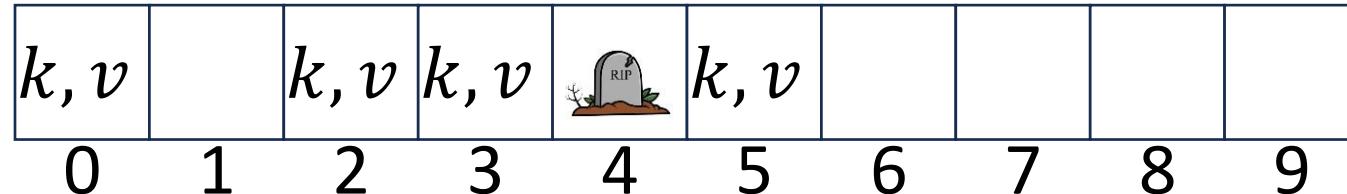
# Linear Probing: Delete

- To delete key  $k$ , where  $h(k) \% \text{table.length} = i$ 
  - Assume it is present
- Beginning at index  $i$ , probe until we find  $k$  (call this location index  $j$ )
- Mark  $j$  as empty (e.g. null), then...
  - Challenge: we need to make sure future finds could be successful
  - What if there were values that mapped to index  $i$  that appeared after  $j$ ?
  - What if there were items that hashed to a value between  $i$  and  $j$  and appeared after  $j$  due to probing?



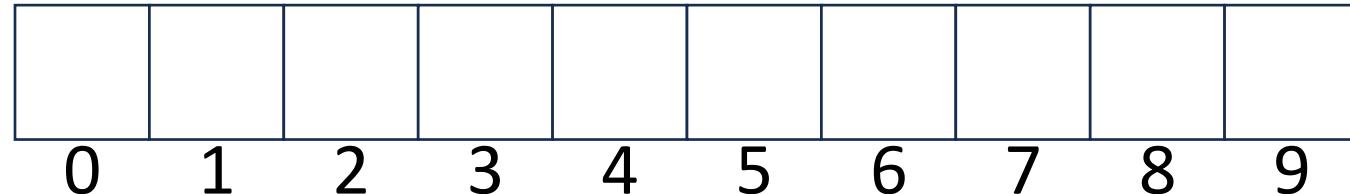
# Linear Probing: Delete

- **Option 1 (harder):** Plug the hole with other items in a way that makes probes behave correctly
- **Option 2 (easier):** “Tombstone” deletion. Leave a special object that indicates something was deleted from there
  - The tombstone does not act as an open space when finding (so keep looking after it's reached)
  - When inserting you can replace a tombstone with a new item



# Linear Probing + Tombstone: Find

- To find key  $k$ 
  - Calculate  $i = h(k) \% \text{table.length}$
  - While  $\text{table}[i]$  has a key other than  $k$ , set  $i = (i+1) \% \text{table.length}$
  - If you come across  $k$  return  $\text{table}[i]$
  - If you come across an empty index, the find was unsuccessful



# Linear Probing + Tombstone: Insert

- To insert  $k, v$ 
  - Calculate  $i = h(k) \% \text{table.length}$
  - While  $\text{table}[i]$  has a key other than  $k$ , set  $i = (i+1) \% \text{table.length}$ 
    - If  $\text{table}[i]$  has a tombstone, set  $x = i$ 
      - That is where we will insert if the find is unsuccessful
  - If you come across  $k$ , set  $\text{table}[i] = k, v$
  - If you come across an empty index, the find was unsuccessful
    - Set  $\text{table}[x] = k, v$  if we saw a tombstone
    - Set  $\text{table}[x] = k, v$  otherwise

