CSE 332 Autumn 2024 Lecture 11: hashing

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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

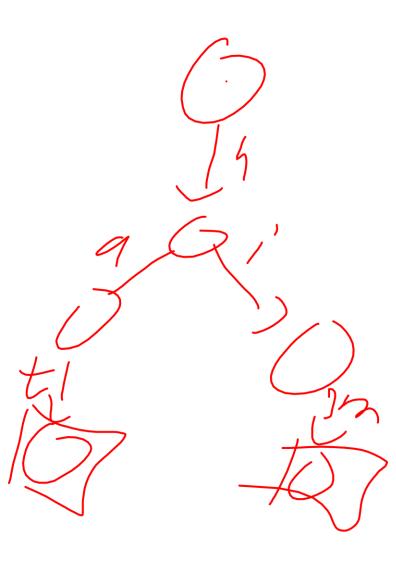
	Data Structure	Time to insert	Time to find	Time to delete
	Unsorted Array	$\Theta(1)$	$\Theta(n)$	$\Theta(n)$
	Unsorted Linked List	$\Theta(1)$	$\Theta(n)$	$\Theta(n)$
	Sorted Array	$\Theta(n)$	$\Theta(\log n)$	$\Theta(n)$
	Sorted Linked List	$\Theta(n)$	$\Theta(n)$	$\Theta(n)$
_	Неар	$\Theta(\log n)$	$\Theta(n)$	$\Theta(n)$
>	Binary Search Tree	Θ(height)	Θ(height)	Θ(height)
•	AVL Tree	$\Theta(\log n)$	$\Theta(\log n)$	$\Theta(\log n)$

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	Θ(height)	Θ(height)	Θ(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)	Θ(1)	$\Theta(1)$	$\Theta(1)$

Dictionary (Map) ADT

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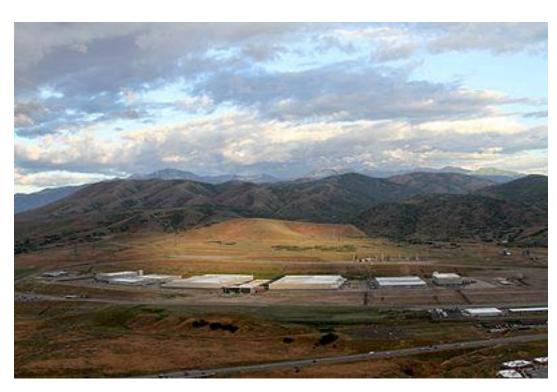
The Best Data Structure!

- 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;
}
```

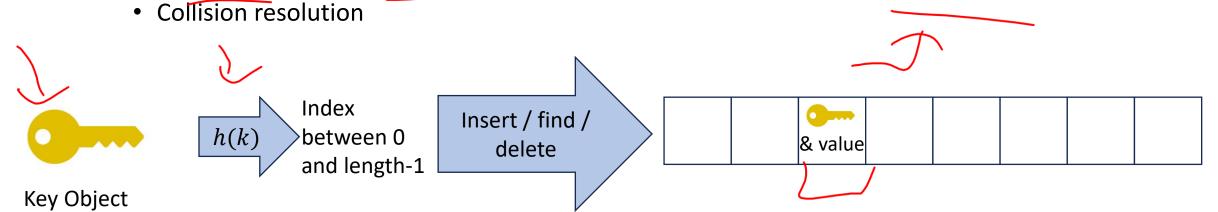
Problem?



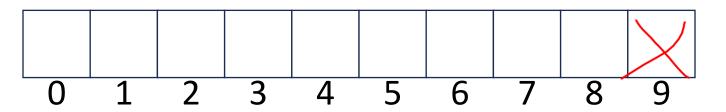


Hash Tables

- 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)



Example



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

What Influences Running time?

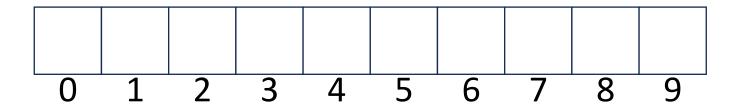
- How long hashing itself takes
- Likelihood of collisions
 - Size of the array vs number of values in the array
 - "quality" of our hash function
- What we do when we have a collision

Properties of a "Good" Hash

- Definition: A hash function maps objects to integers
- Should be very efficient
 - Time to calculate the hash should be negligible
- Should "randomly" scatter objects
 - Even similar objects should hash to arbitrarily different values
- Should use the entire table
- There should not be any indices in the table that nothing can hash to
 - Picking a table size that is prime helps with this
- Should use things needed to "identify" the object
 - Use only fields you would check for a .equals method be included in calculating the hash
 - {fields used for hashing} ⊆ {fields used for . equals}
 - More fields typically leads to fewer collisions, but less efficient calculation

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



Compare These Hash Functions (for strings)

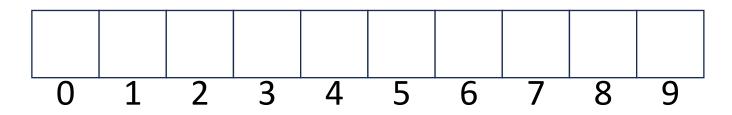
- Let $s = s_0 s_1 s_2 \dots s_{m-1}$ be a string of length m
 - Let $a(s_i)$ be the ascii encoding of the character s_i

$$\bullet \ h_1(s) = a(s_0)$$

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

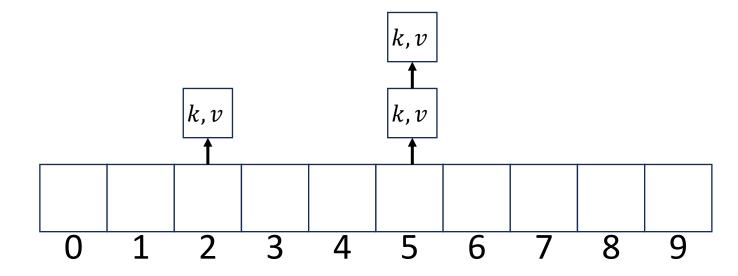
Collision Resolution

- A Collision occurs when we want to insert something into an alreadyoccupied 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



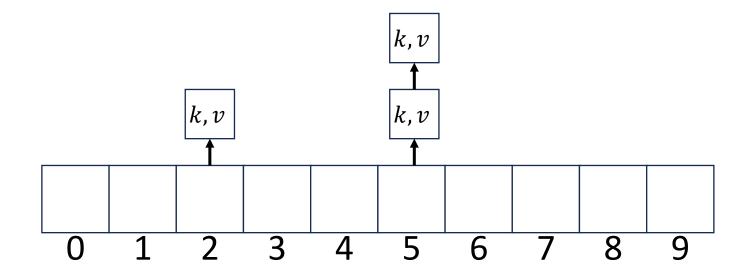
Separate Chaining Insert

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



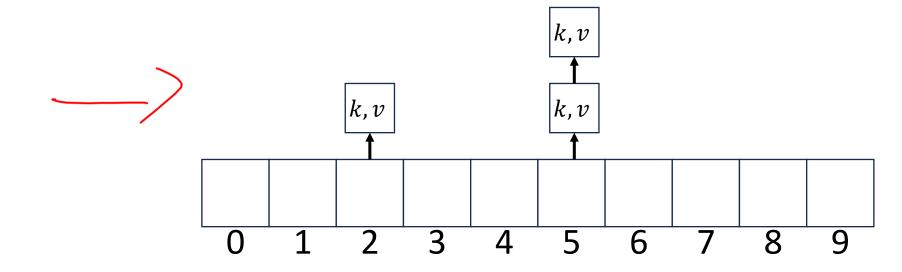
Separate Chaining Find

- To find *k*:
 - Compute the index using i = h(k) % length
 - Call find with the key on the data structure at table[i]



Separate Chaining Delete

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



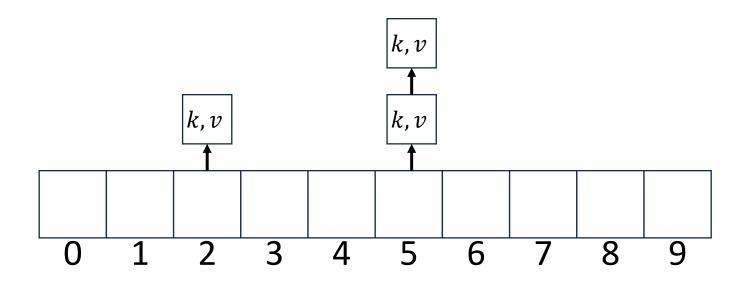
Formal Running Time Analysis

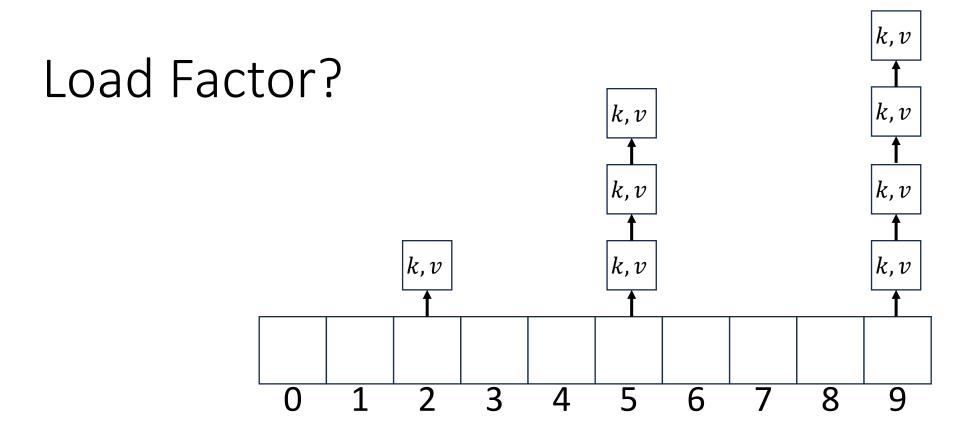
 The load factor of a hash table represents the average number of items per "bucket"

•
$$\lambda = \frac{n}{length}$$

- Assume we have a has 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 $\Theta(1)$?

Load Factor?





Load Factor? k, v|k,v||k,v||k,v||k,v|k, vk, v|k,v|k, vk, v0 2 3 4 5 6 8 9

Collision Resolution: Linear Probing

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



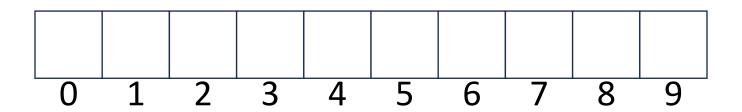
Linear Probing: Insert Procedure

- To insert k, v
 - Calculate i = h(k) % length
 - If table[i] is occupied then try (i + 1)% length
 - If that is occupied try (i + 2)% length
 - If that is occupied try (i + 3)% length

•

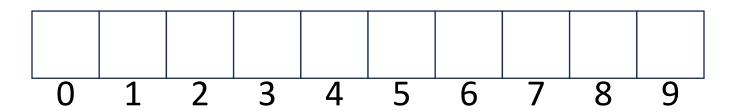


Linear Probing: Find

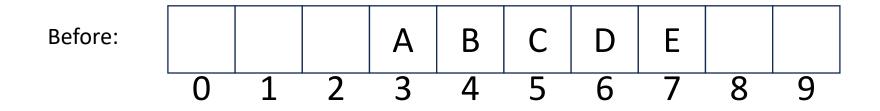


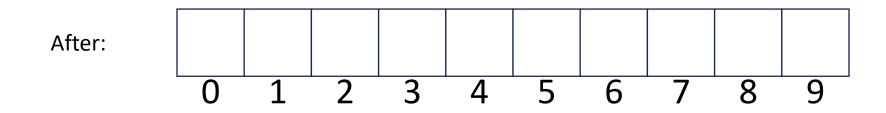
Linear Probing: Find

- To find key k
 - Calculate i = h(k) % length
 - If table[i] is occupied and does not contain k then look at (i + 1) % length
 - If that is occupied and does not contain k then look at (i+2) % length
 - If that is occupied and does not contain k then look at (i + 3) % length
 - ullet Repeat until you either find k or else you reach an empty cell in the table

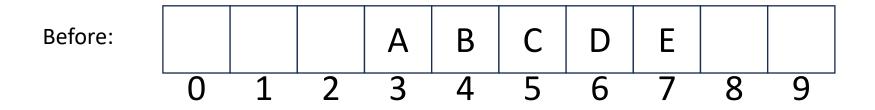


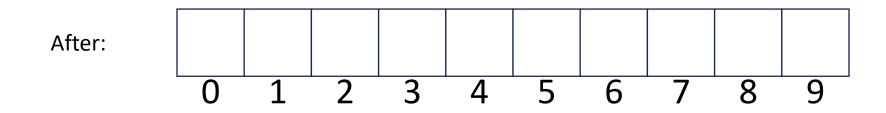
- Suppose A, B, C, D, and E all hashed to 3
- Now let's delete B



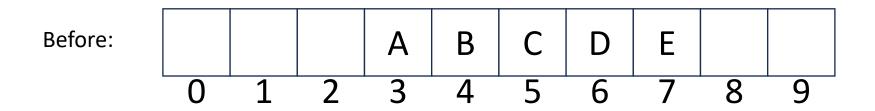


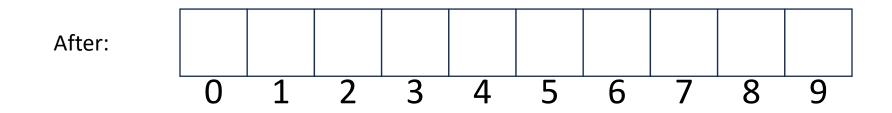
- Suppose A, B, and E all hashed to 3, and C and D hashed to 5
- Now let's delete B



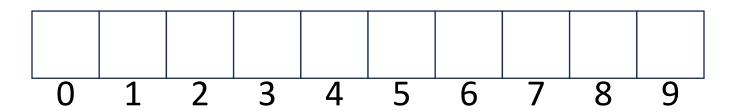


- Suppose A and E hashed to 3, and B,C, and D hashed to 4
- Now let's delete B

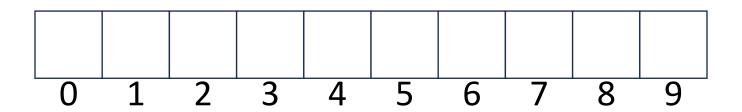




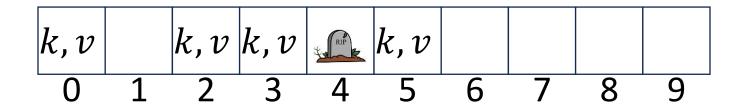
Let's do this together!



- To delete key k, where h(k) = 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 continue probing while doing the following until you find another empty index
 - If you come across a key which hashes to a value ≤ j then move that item to index j and update j.

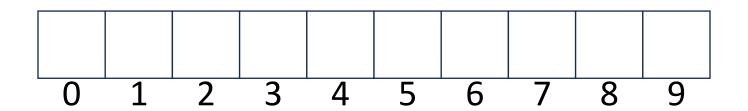


- Option 1: Fill in with items that hashed to before the empty slot
- Option 2: "Tombstone" deletion. Leave a special object that indicates an object was deleted from there
 - The tombstone does not act as an open space when finding (so keep looking after its reached)
 - When inserting you can replace a tombstone with a new item



Linear Probing + Tombstone: Find

- To find key k
 - Calculate i = h(k) % length
 - While table[i] has a tombstone or a key other than k, i = (i + 1) % length
 - If you come across k return 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) % length
 - While table[i] has a key other than k, i = (i + 1) % length
 - If 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 table[i] = k, v
 - If you come across an empty index, the find was unsuccessful
 - Set table[x] = k, v if we saw a tombstone
 - Set table[i] = k, v otherwise

