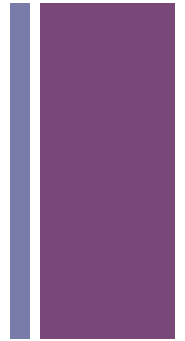


Hashtable details
Sorting Revisited

+ Map/Hashtables



- Each item is a key, value pair

- for example

■ Key	Value
■ studentId	studentRecord
■ town+state	Place
■ word	Definition

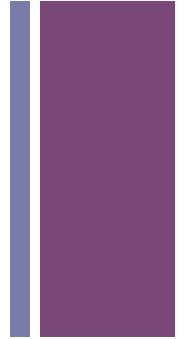
- Sometimes a Map is called a Dictionary

- There is a Set of Keys

- And a Collection of Values

- Goal: efficient add and removal (no concern for order)

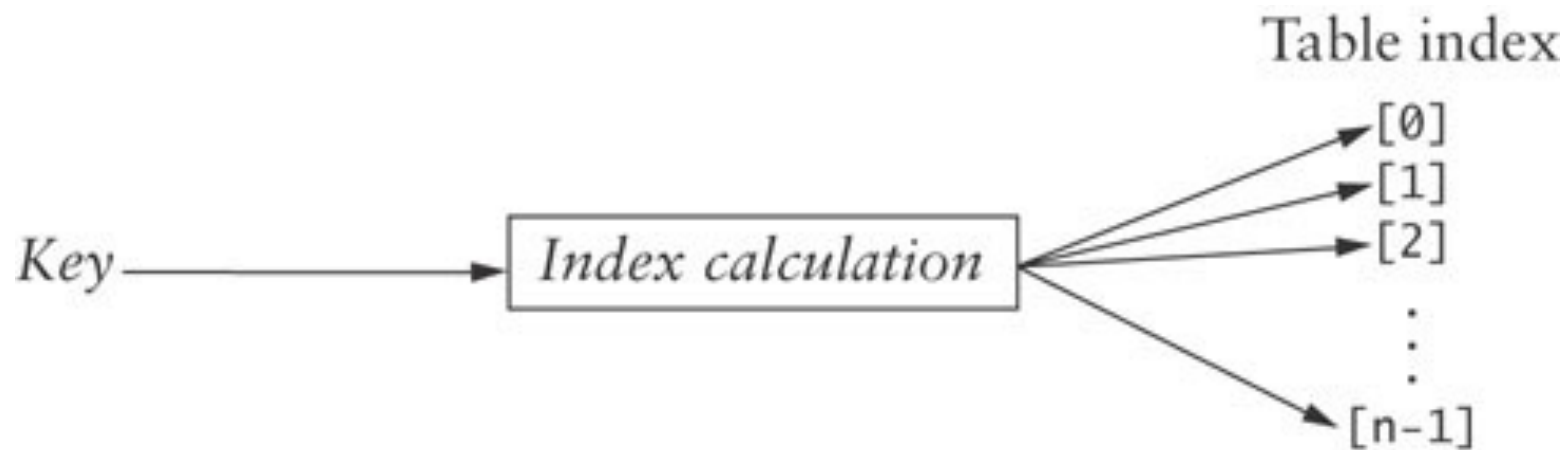
+ Map interface



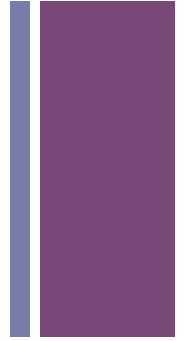
- `public V put(K key, V value)`
- `public V get(K key)`
- `public V remove(K key)`
- The expected is $O(1)$ for a Hashtable, but the worse case is $O(n)$
- A SortedHashMap has $O(\log n)$ insertion and removal

+ Hash Codes and Index Calculation

- The basis of hashing is to transform the item's key value into an integer value (its *hash code*) which is then transformed into a table index

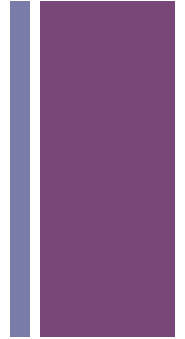


+ Valid Keys



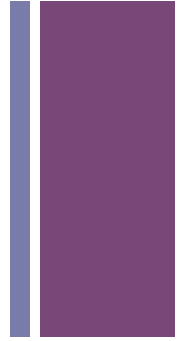
- Keys are typically
 - numbers
 - characters (also numbers)
 - Strings (sequences of characters, which are numbers)
- In Java, any Object can be a Key.
 - hashCode() can be overridden, and should be overridden if equals has been overridden.

+ Goal



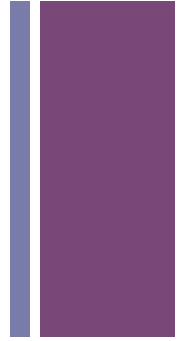
- similar (really all) keys map to different locations
- Question:
 - How do we map a large number of possible values to a much smaller table size.
 - e.g. if we use just 10 letter strings of lowercase letters, then there are 26^{10} possible keys.
- Collisions cause the $O(1)$ property to fail.

+ Potential Hashing functions



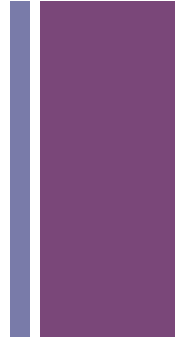
- F1: $(\text{letter1} + \text{letter2} + \text{letter 3} \dots + \text{letter n}) \% \text{tablesize}$
 - ignores order of letters
- F2 : use position and code for letter
- F3 : Java uses $31^{\text{position}} * \text{code}(\text{letter})$

+ Desirable properties



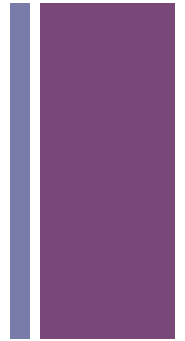
- codes generated are random
 - in int range
 - in $\text{int} \% \text{tablesize}$ range
- easy to compute

+ Handling Collisions



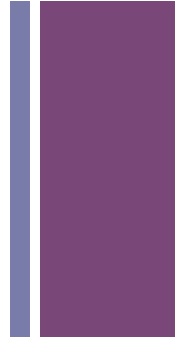
- open addressing
 - If there is a collision, check the next available spot until an open spot is found
 - linear probing
 - quadratic probing
- chaining
 - if there is a collision, add it to the bucket (a linked list)

+ Open Addressing Example (Processing Sketches)



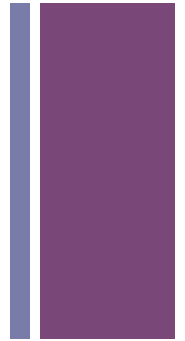
Rhoads	1847927497	4	[0]	
Thomas	1790657756	8	[1]	
Park	2480138	8	[2]	
McBride	1777921980	6	[3]	
Wofford	1116374487	6	[4]	
McPherson	64559159	8	[5]	
Vickers	2116683019	1	[6]	
McAuliffe	15268542	6	[7]	
Cassidy	2075000160	3	[8]	

+ Problem with Open Addressing



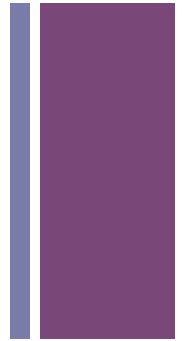
- removal
 - need to keep track of places that once held data
- The chaining strategy removes this concern.

+ Chaining



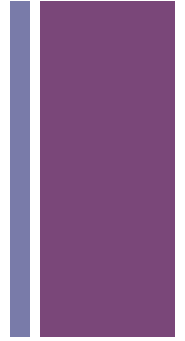
- When collide add to bucket (list)
- removal just remove the item at the hash location from the list

+ Chaining Example (on board)



Rhoads	1847927497	4	[0]	
Thomas	1790657756	8	[1]	
Park	2480138	8	[2]	
McBride	1777921980	6	[3]	
Wofford	1116374487	6	[4]	
McPherson	64559159	8	[5]	
Vickers	2116683019	1	[6]	
McAuliffe	15268542	6	[7]	
Cassidy	2075000160	3	[8]	

+ Performance of Hash tables



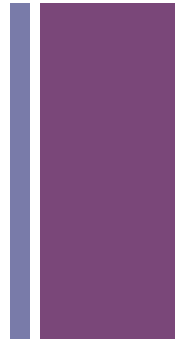
- load factor = $\text{\#filled cells} / \text{table size}$
- lower load factor leads to better performance
- higher load factor leads to worse performance.

+ Performance of Hash Tables versus Sorted Array and Binary Search Tree

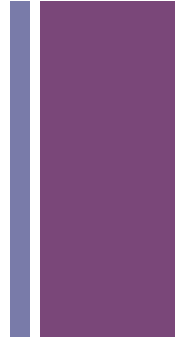
- The number of comparisons required for a binary search of a sorted array is $O(\log n)$
 - A sorted array of size 128 requires up to 7 probes (2^7 is 128) which is more than for a hash table of any size that is 90% full
 - A binary search tree performs similarly
- Insertion or removal

hash table	$O(1)$ expected; worst case $O(n)$
unsorted array	$O(n)$
binary search tree	$O(\log n)$; worst case $O(n)$

+ Questions?



+ Balancing a Binary search tree



- The key to all balanced tree methods is the concept of rotation
 - rotate left
 - rotate right

+ Sorting Review

- Selection Sort
- Insertion Sort
- Merge Sort

