## CSC 321: Data Structures

Fall 2018

Hash tables

- HashSet \& HashMap
- hash table, hash function
- collisions
$>$ linear probing, lazy deletion, clustering, rehashing
$>$ chaining
- Java hashCode method


## HashSet \& HashMap

recall: TreeSet \& TreeMap use an underlying binary search tree (actually, a red-black tree) to store values

- as a result, add/put, contains/get, and remove are $O(\log N)$ operations
- iteration over the Set/Map can be done in $\mathrm{O}(\mathrm{N})$
the other implementations of the set \& Map interfaces, HashSet \& Hashmap, use a "magic" data structure to provide O(1) operations*
*legal disclaimer: performance can degrade to $\mathrm{O}(\mathrm{N})$ under bad/unlikely conditions however, careful setup and maintenance can deliver $\mathrm{O}(1)$ in practice
the underlying data structure is known as a Hash Table


## Hash tables

a hash table is a data structure that supports constant time insertion, deletion, and search on average

- degenerative performance is possible, but unlikely
- it may waste some storage
- iteration order is not defined (and may even change over time)
idea: data items are stored in a table, based on a key
- the key is mapped to an index in the table, where the data is stored/accessed
example: letter frequency
- want to count the number of occurrences of each letter in a file
- have an array of 26 counters, map each letter to an index
- to count a letter, map to its index and increment



## Mapping examples

extension: word frequency

- must map entire words to indices, e.g.,

| "A" $\rightarrow 0$ | "AA" $\rightarrow 26$ | "BA" $\rightarrow 52$ | $\ldots$ |
| :---: | :---: | :---: | :---: |
| "B" $\rightarrow 1$ | "AB" $\rightarrow 27$ | "BB" $\rightarrow 53$ | $\ldots$ |
| $\ldots$ | $\ldots$ | $\ldots$ |  |
| "Z" $\rightarrow 25$ | "AZ" $\rightarrow 51$ | "BZ" $\rightarrow 77 \ldots$ |  |

- PROBLEM?
mapping each potential item to a unique index is generally not practical

```
# of 1 letter words = 26
# of 2 letter words = 26 }\mp@subsup{}{}{2}=67
# of 3 letter words = 26 }\mp@subsup{}{}{3}=17,57
```

- even if you limit words to at most 8 characters, need a table of size 217,180,147,158
- for any given file, the table will be mostly empty!


## Table size < data range

since the actual number of items stored is generally MUCH smaller than the number of potential values/keys:

- can have a smaller, more manageable table

```
e.g., table size = 26
possible mapping: map word based on first letter
```

$$
\text { "A*" } \rightarrow 0 \quad \text { "B*" } \rightarrow 1 \quad \text {... } \quad \text { " } \star \text { " } \rightarrow 25
$$

e.g., table size $=1000$ possible mapping: add ASCII values of letters, mod by 1000

$$
\begin{aligned}
& \text { "AB" } \rightarrow 65+66=131 \\
& \text { "BANANA" } \rightarrow 66+65+78+65+78+65=417 \\
& \text { "BANANABANANABANANA" } \rightarrow 417+417+417=1251 \% 1000=251
\end{aligned}
$$

- POTENTIAL PROBLEMS?


## Collisions

the mapping from a key to an index is called a hash function

- the hash function can be written independent of the table size
- if it maps to an index > table size, simply wrap-around (i.e., index \% tableSize)
since |range (hash function)| < |domain(hash function)|,
Pigeonhole Principle ensures collisions are possible ( $\mathrm{v}_{1} \& \mathrm{v}_{2} \rightarrow$ same index)

$$
\text { "ACT" } \rightarrow 67+65+84=216 \quad \text { "CAT" } \rightarrow 67+65+84=216
$$

techniques exist for handling collisions, but they are costly (LATER)
it's best to avoid collisions as much as possible - HOW?

- want to be sure that the hash function distributes the key evenly
- e.g., "sum of ASCII codes" hash function

OK if table size is 1000
BAD if table size is 10,000
most words are $\leq 10$ letters, so max sum of ASCII codes $=1,270$
so most entries are mapped to first $13 \%$ of table

## Better hash function

a good hash function for words should

- produce an even spread, regardless of table size
- take order of letters into account (to handle anagrams)
- the hash function used by java. util. String multiplies the ASCII code for each character by a power of 31

```
hashCode() = char }\mp@subsup{\mp@code{co}}{}{*}3\mp@subsup{1}{}{(len-1)}+\mp@subsup{\operatorname{char}}{1}{*}3\mp@subsup{1}{}{(len-2)}+\mp@subsup{\operatorname{char}}{2}{*}*3\mp@subsup{1}{}{(len-3)}+\ldots+cchar(len-1
```

Where len $=$ this.length(), chari $=$ this.charAt(i):

```
/**
* @return an int used as the hash index for this string
*/
private int hashCode() {
    int hashIndex = 0;
    for (int i = 0; i < this.length(); i++) {
        hashIndex = (hashIndex*31 + this.charAt(i));
    }
    return hashIndex;
```


## Word frequency example

returning to the word frequency problem

- pick a hash function
- pick a table size
- store word \& associated count in the table
- as you read in words, map to an index using the hash function if an entry already exists, increment otherwise, create entry with count $=1$


WHAT ABOUT COLLISIONS?

## Linear probing

linear probing is a simple strategy for handling collisions

- if a collision occurs, try next index \& keep looking until an empty one is found (wrap around to the beginning if necessary)
example: assume "first letter" hash function
- insert "BOO", "BAR", "COO", "BOW, ...
linear probing requires "lazy deletion"
- when you delete an item, you can't just empty the location, since it would leave a hole
- subsequent searches would reach that whole and stop probing
- instead, leave a marker (a.k.a a tombstone) in that spot 0 can be overwritten but not skipped when probing
example: given above insertions
- delete "BAR", search for "COO"



## Clustering and load factor

in practice, probes are not independent

- as the table fills, clusters appear that degrade performance

| maps to | $0,5-7$ require | 1 check | 4 | "DOG" |
| :--- | :--- | :--- | :--- | :--- |
| map to | 4 requires | 2 checks | 5 |  |
| map to | 3 requires | 3 checks |  | 6 |
| map to | 2 requires | 4 checks |  |  |
| map to | 1 requires | 5 checks | 7 |  |
| average $=18 / 8=2.25$ checks |  |  |  |  |

the load factor $\lambda$ is the fraction of the table that is full
empty table $\lambda=0 \quad$ half full table $\lambda=0.5 \quad$ full table $\lambda=1$

THEOREM: assuming a reasonably large table, the average number of locations examined per insertion is roughly $\left(1+1 /(1-\lambda)^{2}\right) / 2$

| empty table | $\left(1+1 /(1-0)^{2}\right) / 2=1$ |
| :--- | :--- |
| half full | $\left(1+1 /(1-.5)^{2}\right) / 2=2.5$ |
| $3 / 4$ full | $\left(1+1 /(1-.75)^{2}\right) / 2=8.5$ |
| $9 / 10$ full | $\left(1+1 /(1-.9)^{2}\right) / 2=50.5$ |

## Rehashing

as long as you keep the load factor low (e.g., $<0.75$ ), inserting, deleting and searching a hash table are all $O(1)$ operations
if the table becomes too full, then must resize

- create new table at least twice as big
- just copy over table entries to same locations???
- NO! when you resize, you have to rehash existing entries new table size $\rightarrow$ new hash function (+ different wraparound)


## LET hashCode = word.length()

ADD "UP"
ADD "OUT"
ADD "YELLOW"


| NOW <br> RESIZE <br> AND <br> REHASH | 0 |
| :---: | :---: |
|  | 1 |
|  | 2 |
|  | 3 |
|  | 4 |
|  | 5 |
|  | 6 |
|  | 7 |

## Chaining

linear probing (or variants) were initially used when memory was expensive - clustering, lazy deletion, and rehashing are all issues
modern languages like Java utilize a different approach
chaining:

- each entry in the hash table is a bucket (list)
- when you add an entry, hash to correct index then add to bucket
- when you search for an entry, hash to correct index then search sequentially



## Analysis of chaining

in practice, chaining is generally faster than probing

- cost of insertion is $\mathrm{O}(1)$ - simply map to index and add to list
- cost of search is proportional to number of items already mapped to same index e.g., using naïve "first letter" hash function, searching for "APPLE" might requires traversing a list of all words beginning with ' A '
if hash function is fair, then average size of each bucket is $\lambda$ (load factor) $\rightarrow$ average cost of a successful search is roughly $\lambda / 2$
chaining is sensitive to the load factor, but not as much as probing - WHY?
chaining uses more memory - WHY?


## Hashtable class

Class Hashtable<K,V>

## Constructor Summary









Java provides a basic hash table implementation

- utilizes chaining
- can specify the initial table size \& threshold for load factor
- can even force a rehashing
not commonly used, instead provides underlying structure for HashSet \& HashMap



## HashSet \& HashMap

java.util.HashSet and java.util. HashMap use hash table w/ chaining

- e.g., HashSet<String> HashMap<String, Integer>

- defaults: table size = 16, max capacity before rehash $=75 \%$ can override these defaults in the HashSet/HashMap constructor call

```
note: iterating over a HashSet or HashMap is:O(num stored + table size)
WHY?
```

using HashMap instead of TreeMap

- containsKey, get \& put operations are all $\mathrm{O}(1)^{*}$
- however, iterating over the keySet (and their values) does not guarantee any order
- if you really care about speed $\rightarrow$ use HashSet/HashMap
- if the data/keys are comparable \& order matters $\rightarrow$ use TreeSet/TreeMap

```
import java.util.Map;
    mport java,util Scanner
    import java.io.File;
public class WordFreq {
    private Map<String, Integer> words;
    public WordFreq() (
        words = new HashMap<String, Integer>();
    }
    public WordFreq(String filename) {
            this();
            try {
            Scanner infile = new Scanner(new File(filename));
                while (infile.hasNext()) {
                    String nextWord = infile.next();
                    this.add(nextWord);
            }
            catch (java.io.FileNotFoundException e) {
            System.out.println("FILE NOT FOUND");
        }
    public void add(String newWord)
            String cleanWord = newWord.toLowerCase();
            if (words.containsKey(cleanWord)) (
            words.put(cleanWord, words.get(cleanWord)+1);
            else {
            words.put(cleanWord, 1)
        }
    public void showAll() {
            for (String str : words.keySet()) {
            System.out.println(str + ": " + words.get(str));
            }
        }

\section*{hashCode function}
```

import java.util.Calendar;
public class Person
private String firstName, lastName;
private Calendar birthday;
public Person(String fname, String lname, int month, int day, int year) {
this.firstName = fname;
this.lastName = lname;
this.birthday = new GregorianCalendar(year, month-1, day);
}
public String toString() {
return this.firstName + " " + this.lastName + "% " + +
this.birthday.get(Calendar.DAY OF MONTH) + "/" +
this.birthday.get(Calendar. YEA\overline{R});
}
///////////////////////////////////////////////////////////////////////
public static void main(String[] args) {
Person p1 = new Person("Chris", "Marlowe", 5, 25, 1992);
System.out.println(p1);
System.out.println(p1.hashCode());
Person p2 = new Person("Alex", "Cooper", 2, 5, 1994);
System.out.println(p2);
System.out.println(p2.hashCode());
Person p3 = new Person("Pat", "Phillips", 2, 5, 1994);
System.out.println(p3);
System.out.println(p3.hashCode());
}
}

## overriding hashCode v. 1

```
import java.util.Calendar;
import java.util.GregorianCalendar;
public class Person {
    private String firstName, lastName:
    private Calendar birthday;
    public Person(String fname, String lname, int month, int day, int year) {
        this.firstName = fname;
        this.lastName = Iname;
        this.birthday = new GregorianCalendar(year, month-1, day);
    }
    public String toString() {
            return this.firstName + " " + this.lastName + " " " +
                (this.birthday.get(Calendar.MONTH) MONTH) + + + + +
                this.birthday.get(Calendar. YEA\overline{R});
    }
                this.birthday.get(Calendar.DAY_OF_MONTH) + "/"
    public int hashCode() {
        return Math.abs((int)this.birthday.getTimeInMillis());
    }
```



```
    public static void main(String[] args) {
            Person p1 = new Person("Chris", "Marlowe", 5, 25, 1992);
            System.out.println(p1).
            System.out.println(pl.hashCode());
            Person p2 = new Person("Alex", "Cooper", 2, 5, 1994);
            System.out.println(p2);
            Person p3 = new Person("Pat", "Phillips", 2, 5, 1994);
            System.out.println(p3);
            System.out.println(p3.hashCode());
    }

\section*{overriding hashCode v. 2}
```

import java.util.Calendar;
blic class Person {
private Calendar birthday;
public Person(String fname, String lname, int month, int day, int year) {
this.firstName = fname
this.birthday = new GregorianCalendar(year, month-1, day)
}
public String tostring() {
return this.firstName }\mp@subsup{}{~}{+
(this.birthday.get(Calendar.MONTH)+1) + "/" +
this.birthday.get(Calendar.DAY OF MONTH) + "/" +
this.birthday.get(Calendar. YEA\overline{R});
}
public int hashCode() {
return Math.abs((int)this.birthday.getTimeInMillis() +
return Math.abs((int)this.birthday.getTimeInMillis() +
}
111111111111111111111111111111111111111111111111111111111111111111111
public static void main(String[] args) {
Person p1 = new Person("Chris", "Marlowe", 5, 25, 1992);
Person pl = new Person(%)
System.out.println(p1);
Person p2 = new Person("Alex", "Cooper", 2, 5, 1994);
System.out.println(p2)
System.out.println(p2);
to avoid birthday
collisions, can also
incorporate the
names
- utilize the String
hashCode method
413568008
Alex Cooper: 2/5/1994
Person p3 = new Person(
System.out.println(p3);
System.out.println(p3.hashCode());
}

## Graphs (sneak peek)

trees are special instances of the more general data structure: graphs -informally, a graph is a collection of nodes/data elements with connections

a tree is a graph in which one node has no edges coming into it (the root) and no cycles


## Finite State Machines (FSMs)

many useful problems can be defined using simple graphs

- a Finite State Machine (a.k.a. Finite Automaton) defines a finite set of states (i.e., nodes) along with transitions between those states (i.e., edges)
e.g., the logic controlling a coin-operated turnstile

can be in one of two states: locked or unlocked
- if locked, pushing $\rightarrow$ it does not allow passage \& stays locked inserting coin $\rightarrow$ unlocks it
- if unlocked, pushing $\rightarrow$ allows passage \& then relocks inserting coin $\rightarrow$ keeps it unlocked


## Other examples

Claude Shannon used a
FSM to show constraints on
Morse code

can use a FSM to specify the behavior of a vending machine
adding a coin ( $Q, D, N$ ) changes the state

## HW6: Simulate a FSM


locked push locked locked coin unlocked unlocked push locked
unlocked coin unlocked
model a FSM by storing the edges and providing lookup methods

```
private HashMap<StateLabel, HashMap<EdgeLabel, StateLabel>> table;
```

| locked | loin | unlocked |
| :--- | :--- | :--- |
|  | push | locked |
| unlocked | coin | unlocked |
|  | push | locked |

the key to the table is the start state of an edge the value is another map, which maps edge labels to the end states
table.get("locked") $\rightarrow$
$\quad$ a map containing edges from "locked"
table.get("locked").get("coin") $\rightarrow$ "unlocked"

## HW6: Other examples



| inLetter . inLetter |
| :--- |
| inLetter - inLetter |
| inLetter - betweenLetters |
| inLetter - betweenLetters |
| betweenLetters . inLetter |
| betweenLetters - inLetter |




## HW6: PathTracer


given a start state and sequence of edges, determine the end state

```
Enter FSM file: turnstile.txt
Enter a start state (* to end): locked
Enter an edge sequence (separated with whitespace): coin push
End state: locked
Enter another start state (* to end): locked
Enter an edge sequence (separated with whitespace): push coin push push coin
End state: unlocked
Enter another start state (* to end): *
DONE
```


## HW6: PathFinder

you are given a method that finds a shortest path between to states

```
fsm.findPath("0cents","35cents") -> ["0cents","10cents","35cents"]
```

- you will write a driver class that repeatedly finds and prints paths

```
Enter FSM file: change.txt
Enter a start state (* to end): 0cents
Enter the end state: 35cents
State path: [0cents, 10cents, 35cents]
Enter another start state (* to end): 5cents
Enter the end state: 20cents
State path: [5cents, 10cents, 20cents]
Enter another start state (* to end): 10cents
Enter the end state: 5cents
No SUCH PATH
Enter another start state (* to end): *
DONE
```



