Set and Map ADT Hash tables

Lecture 13

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Modeling dictionaries

sensacionalista adj. sensational sensato adj. sensible sensibilidad n. sensibility sensibilizar v. sensitize sensible adj. sensitive sensiblero adj. maudlin sensor n. sensor sensorial adj. sensory sensual adj. sensual sensualidad n. sensuality sentar v.t. sit sentencia n. sentence sentenciar v. sentence sentencioso adi. sententious sentido n. sense sentimental adj. sentimental sentimiento n. feeling sentir v. feel sentirse atrapado v. trammel seña n. sign señal n. signal señalación n. pointing señalar v. pinpoint

sepultura n. sepulture sepulturero n. undertaker sequía n. drought séquito n. entourage ser incapaz adj. unable ser mayor que v. outweigh ser mujeriego v. womanize ser suficiente v. suffice ser v be serenidad n. serenity sereno adj. serene seres queridos adj. nearest series n. series serio adj. serious, earnest sermón n. sermon sermonear v. preach serpentear v. crankle, wriggle serpentina n. streamer serpentino adj. serpentine serpiente n. snake serrar v. saw serrín n. sawdust servicio n. service

- Dictionary is a collection of pairs
- Each pair has a key and the value associated with this key
- The most important functionality of a dictionary is the search of a value by a given key

Let's assume for the simplicity of the discussion that each key is unique: no duplicate keys are allowed

Abstract Data Type: Map (Dictionary, Associative Array)

Specification

Dictionary is an Abstract Data Type which stores a collection of (key, value) pairs and supports the following operations:

- → Add (k, e) adds element e to the collection and associates it with key k
- → Remove (k) removes element with key k from the collection
- → **Get (k)** returns the element associated with key *k*

The main functionality is to quickly locate a value given the key

→ Contains (k) - returns *True* if there is an element associated with the key *k*. Returns *False* otherwise

Which data structure to use to implement Dictionary ADT?

Main goal: **locate the element by the key**

- Linked List, Array N elements are unsorted search requires O(N) time
- Sorted array N elements are sorted O(log N) binary search

It doesn't seem like we can do much better

Searching in time O(1)

- ➤ How about O(1), that is, constant-time search?
- ➤ We can do it if we store data in an array organized in a particular way

"Hash is a food, especially meat and potatoes, chopped and mixed together; a confused mess " (en.wiktionary.org/wiki/hash)

The idea of

Hashing

Example 1: First repeating character

Input: String *S* of length *N*

Output: first repeating character (if any) in S

 \succ The obvious O(N^2) solution:

for each character in order:

check whether that character is repeated

Example 1: First repeating character

Input: String S of length N

Output: first repeating character (if any) in S

	_
а	97
b	98
С	99
d	100
е	101
f	102
g	103
h	104
i	105
j	106
k	107
I	108
m	109
n	110
0	111

The number of all possible characters is 256 (ASCII characters)

- ➤ We create an array *H* of size 256 and initialize it with all zeros
- For each input character c go to the corresponding slot H[c] and set count at this position to 1
- Since we are using arrays, it takes constant time for reaching any location
- ➤ Once we find a character for which counter is already 1 we know that this is the one which is repeating for the first time

Example 1: First repeating character

Input: String *S* of length *N*

Output: first repeating character (if any) in S

cabare

а	97	1
b	98	1
С	99	1
d	100	
е	101	
f	102	
g	103	
h	104	
i	105	
j	106	
k	107	
-	108	
m	109	
n	110	
0	111	

Run-time O(N)

- ➤ Because the total number of all possible keys is small (256), we were able to map each key (character) to a single memory location
- ➤ The key tells us precisely where to look in the array!

This method of storing keys in the array is called direct addressing: store key k in position k of the array

Example 2: First repeating number

Input: Array A containing N **integers**

Output: first repeating number (if any) in A

- ➤ This very similarly looking problem is more difficult to solve with *direct addressing*
- ➤ The total number of all possible integers is 2,147,483,647. This is the universe of all possible keys thus the size of the array
- What if we have only 25 integers to store? Impractical
- > What if the keys are floats/strings/objects? Impossible
- ➤ For these cases we use a technique of *hashing*: we convert **each key into a number** using a *hash function*

Intuition: hashing inputs

- ➤ Suppose we were to come up with a "magic function" that, given a key to search for, would tell us the exact location in the array such that
 - If key is in that location, it's in the array
 - If key is not in that location, it's not in the array
- > This function would have no other purpose
- ➤ If we look at the function's inputs and outputs, the connection between them won't "make any sense"
- ➤ This function is called a *hash function* because it "makes hash" of its inputs

Case study: hashing students

- Suppose we want to store student objects in the array
- ➤ For each student we apply the following *hash function*:

```
hashCode(Student) =
   length (Student.lastName)
```

This gives us the following values:

- hashCode('Chan')=4
- hashCode('Yam')=3
- hashCode('Li')=2
- hashCode('Jones')=5
- hashCode('Taylor')=6

Array of students: *hash table*

- We place the students into array slots which correspond to the computed hash values:
 - hashCode('Chan')=4
 - hashCode('Yam')=3
 - hashCode('Li')=2
 - hashCode('Jones')=5
 - hashCode('Taylor')=6

0	
1	
2	Li
3	Yam
4	Chan
5	Jones
6	Taylor
7	

Good hash function: length of the last name

- ➤ Our hash function is easy to compute
- ➤ An array needs to be of size 18 only, since the longest English surname, Featherstonehaugh (Guinness, 1996), is only 17 characters long
- ➤ We waste a little bit of space with entries 0,1 of the array, which do not seem to be ever occupied. But the waste is not that bad either

0	
1	
2	Li
3	Yam
4	Chan
5	Jones
6	Taylor
7	

Bad hash function: length of the last name

Suppose we have a new student: Smith

o hashValue('Smith')=5

➤ When several values are hashed to the same slot in the array, this is called a collision

➤ Now what?

0	
1	
2	Li
3	Yam
4	Chan
5	Jones
6	Taylor
7	

Looking for a good hash function: day of birth?

- ➤ What about the day of birth?
 - We know that this would be 365 (366) possible values, so we can have an array of size 366

Questions Responses 29 Settings

This form is no longer accepting responses. It was used for the in-class demo of the Birthday parado:

13 14 15 2 17 21 25

29 responses

What day is your birthday?

Link to Sheets

 The birth day of each student is randomly distributed across this range, and this hash function is easy to

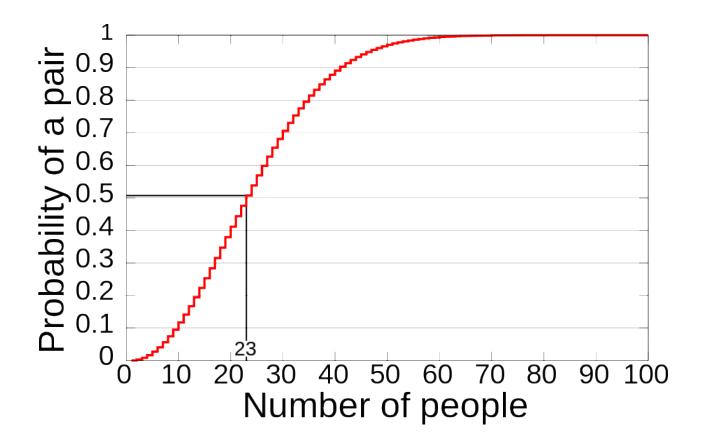
compute

Experiment:

https://forms.gle/BCNWDHvUfRuCpr5p6

Birthday paradox

- For a college with only n=24 students, the probability that any 2 of them were born on the same day is > 0.5
- ➤ Let's approximate this probability:
 - The probability of any two people not having the same birthday is: p = 364/365
 - The number of possible student pairs is $\binom{n}{2} = n(n-1)/2 = 276$
 - The probability for *n* students of not having birthday on the same date is $p^{n(n-1)/2}$. For 24 students this gives: $(364/365)^{276} \approx 0.47$.
 - \circ Then the probability of finding a pair of students colliding on their birthday is 1.00 0.47 = 0.53!
- ➤ This is called a <u>birthday paradox</u>



In search for a perfect hash function

A perfect *hash function* is a function that:

- 1. When applied to an Object, returns a number
- 2. When applied to *equal* Objects, returns the *same* number for each
- 3. When applied to *unequal* Objects returns *different* numbers for each, preventing collisions.
- 4. The numbers returned by hash function are *evenly* distributed between the range of the positions in the array
- 5. We also require for our hash function to be *efficiently* computable

non-random inputs → random numbers?

In search for a perfect hash function

- ➤ How to come up with this perfect hashing function?
- ➤ In general there is no such magic function ②
 - In a few specific cases, where all the possible values are known in advance, it is possible to define a perfect hash function. For example hashing objects by their SSN numbers. But this will require an array to be of size 109
- ➤ It seems that collisions are essentially unavoidable. That means that we cannot guarantee that the hash of a key will bring us into the exact position where this key is located
- ➤ What is the next best thing?
- A perfect hash function would have told us exactly where to look
- However, the best we can do is a function that tells us in what area of an array to start looking!



Back to students:

Hashing names by summing up their character values

➤ It seems like a good idea to map each student surname into a number by adding up the ranks (or ASCII codes) of letters in this surname.

hashCode (S) =
$$\sum_{i=0}^{len(S)} rank(S[i])$$

```
10
12
13
14
15
16
17
18
19
22
23
25
```

What a great hash function!

hashCode (S) =
$$\sum_{i=0}^{len(S)} rank(S[i])$$

- hashCode('Chan')=3+8+1+14=26
- hashCode('Yam')=24+1+13=38
- hashCode('Li')=12+9=21
- hashCode('Jones')=10+15+14+5+18=62
- hashCode('Taylor')=19+1+24+12+15+17=88
- hashCode('Smith')=18+13+9+19+8=67

```
2
7
8
10
11
12
13
14
15
16
17
18
19
20
22
23
24
25
```

Still a lot of collisions!

hashCode (S) =
$$\sum_{i=0}^{len(S)} rank(S[i])$$

- → Not only hashCode('Yam')=hashCode('May')
- → But hashCode('Chan') = hashCode('Lam') !

The function takes into account the value of each character in the string, but **not the order of characters**

Polynomial hashing scheme

- ➤ The summation is not a good choice for sequences of elements where the order has meaning
- ightharpoonup Alternative: choose $A \neq 1$, and use a hash function for string S of length N:

$$hashCode(S) = \sum_{i=0}^{N-1} S[i] \cdot A^{N-1-i} =$$

$$S[0] \cdot A^{N-1} + S[1] \cdot A^{N-1-1} + S[2] \cdot A^{N-1-2} + \dots + S[N-1] \cdot A^{N-1-(N-1)}$$

➤ This is a polynomial of degree N for A, and the elements (characters) of the String are the coefficients of this polynomial

Example: polynomial hashing $hashCode(S) = \sum_{i=0}^{N-1} S[i] \cdot A^{N-1-i} =$ $S[0] \cdot A^{N-1} + S[1] \cdot A^{N-1-1} + S[2] \cdot A^{N-1-2} + \dots + S[N-1] \cdot A^{N-1-(N-1)}$ S_1 = 'Yam' S_2 = 'May' A = 31 $hashCode(S_1) = 24*31^2 + 1*31^1 + 13*31^0 = 23108$ $hashCode(S_2) = 13*31^2 + 1*31^1 + 24*31^0 = 12548$ ➤ Instead of using the *summation* of all character values, the polynomial hash function introduces interactions between different bits of successive characters that will provoke or spread randomness of the result

10

12

13

14

15

16

17

18

19

20

21

22

23

24

25

How to compute polynomial of degree N in time O(N)

Horner's method:

```
egin{split} p(x) &= a_0 + a_1 x + a_2 x^2 + a_3 x^3 + \dots + a_n x^n \ &= a_0 + x igg( a_1 + x igg( a_2 + x igg( a_3 + \dots + x (a_{n-1} + x \, a_n) \dots igg) igg) \end{split}
```

```
Let x=31, and a<sub>0</sub> ... a<sub>n</sub> represent n+1 characters of string S:
public int hashCode() {
   int hash=0;
   for (int i=0; i< length(); i++)
       hash=hash*31+S[i];
   return hash;
}</pre>
```

Java String hashCode()

- ➤ Polynomial hashing is quite good: for different strings it returns mostly different values which are well spread over the range of all possible integers
- This hash function is also very efficient, since we need only n = length() steps to compute it

Implemented inside the String class

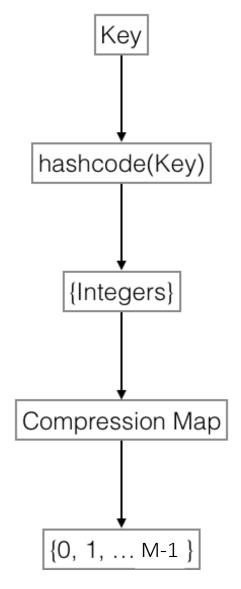
Reducing the range of hashCode to the capacity of the array

- ➤ The output of a good hash function is a number ~randomly distributed over the range of **all** integers.
 - But we need to store our objects in the array of size M
- ➤ Step 2: compression mapping
 - Converting integers in range ~ [0,40000000] to integers in range [0, M]
 - The simplest way to do it: |hashCode| MOD M
 - In practice, the MAD (Multiply Add and Divide) method:

|(A*hashCode+B) MOD M|

The best results when A, B and M are primes

Full hashing



Hashing Students to 7 slots

→ Applying polynomial hashing:

hashCode('Taylor')=-880692189 hashCode('Yam')=119397 hashCode('Li')=345 hashCode('Lee')=107020 hashCode('Lam')=106904

hashCode('Roy')=113116

→ Applying compression mapping: |(11*hashCode+13) MOD 7|

arrayIndex('Taylor')=6 arrayIndex('Yam')=2 arrayIndex('Li')=4 arrayIndex('Lee')=5 arrayIndex('Lam')=3 arrayIndex('Roy')=1

	0	
	1	Roy
	2	Yam
1	3	Lam
,	4	Li
	5	Lee
	6	Taylor

No more collisions?

 Does Java hashCode always produce different hash code for different strings?

The answer is NO.

If you run the code in the box, you will find out that

- The words Aa and BB have the same hashCode
- Words variants and gelato hash to the same value

```
• ...
```

```
public static void main(String [] args) {
   String [] words=new String[6];
   words[0]="Aa";
   words[1]="BB";
   words[2]="variants";
   words[3]="gelato";
   words[4]="misused";
   words[5]="horsemints";

for(int i=0;i<6;i++) {
    System.out.print("Hash code of "+words[i]+": ");
    System.out.println(words[i].hashCode());
   }
}</pre>
```

 We have to be prepared to deal with collisions, since they are unavoidable