CS151 Intro to Data Structures

Hashmaps

Announcements & Outline

No lab today. HW6 and Lab8 due Sunday
Only two homeworks left after this one!
CS Major Info session - today at 4pm Park 231

Today:

- Map review
- Hash functions and some cool applications
- Hash Maps a magical data structure

Maps

- Also called "dictionaries" or "associative arrays"
- Similar syntax to an array:
 - m[key] retrieves a value
 - m[key] = value assigns a value
 - keys need not be ints

data structure that stores a collection of key-value pairs

Map ADT

- get (k): if the map M has an entry with key k, return its associated value; else, return null
- put (k, v): insert entry (k, v) into the map M; if key k is not already in M, then return null; else, replace old value with v and return old value associated with k
- remove(k): if the map M has an entry with key k, remove it from M and return its associated value; else, return null
- size(), isEmpty()
- keySet (): return an iterable collection of the keys in M
- values (): return an iterator of the values in M
- entrySet (): return an iterable collection of the entries in M

Map.Entry Interface

- A (Key, Value) pair
- Keys and Values can be any reference type
- Methods:
 - o getKey()
 o getValue()
 o setValue(V val)
- Implementation: SimpleEntry

ArrayMap

Last class we implemented a Map as an array of SimpleEntry s

Implementation & Runtime complexity?

- put
- get
- remove

HashMaps

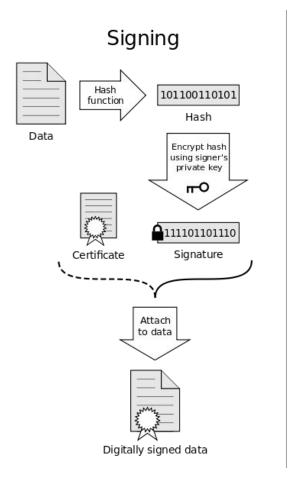
Hash Functions

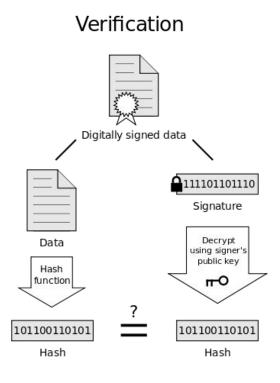
- · A hash function maps an arbitrary length input to a fixed length unique output
- https://emn178.github.io/online-tools/sha256.html
- Applications
 - data structures
 - encryption / digital signatures
 - blockchain
- Properties of a good hash function:
 - one way
 - collision resistant
 - uniformity
 - quick to compute

Digital Signatures

cryptographic method used to validate the *authenticity* and *integrity* of a digital message or document

- In public key cryptography Each user has two keys:
 - one **public key** (e, n)
 - one private key (d, n)
- RSA encryption:
 - Using the sender's private key,
 - encrypt the hash of the message or document M.
 - Signature = Hash(M)^d mod n
- RSA Decryption
 - Using the sender's public key,
 - decrypt the message
 - hash = Signature ^ e mod n





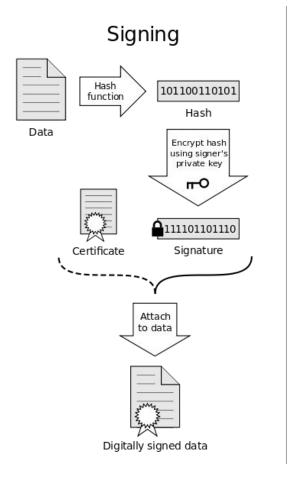
Digital Signatures

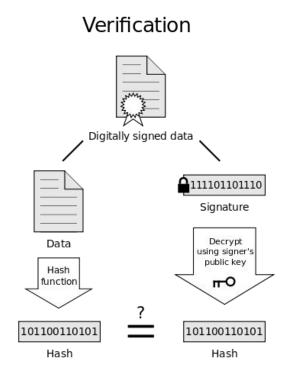
cryptographic method used to validate the *authenticity* and *integrity* of a digital message or document

Why is hashing necessary?

- 1. Fixed length output
 - a. Encryption scheme handles a small predictably sized amount of data
 - b. More efficient to sign the hash than the entire document / message
- 2. Data Integrity
 - Any small change in the message will result in a completely different hash value

Question: why do I need to send the original data **and** the signature? Why can't I recover the document from the signature?





If the hashes are equal, the signature is valid.

Hash Functions on the Blockchain

1. Digital signatures on each transaction

- 2. Data integrity and immutability
 - a. decentralized computing every machine needs to agree on the state of the program
 - b. O(n) algorithm to verify every memory slot is the same
 - c. Hash is easier to check against

Back to data structures:)

A Simple Hash Function

```
Given an int x...

h(x) = last 4 digits of x
```

- one way?
- collision resistant?
- · uniform?

Another Simple Hash Function

$$h(x) = x \% N$$

- one way?
- collision resistant?
- uniform?

HashMaps

 How can we use hash functions to improve the performance of our ArrayMap implementation?

Use hash as our array index!

- How would the following operations look?
 - put
 - get
 - remove

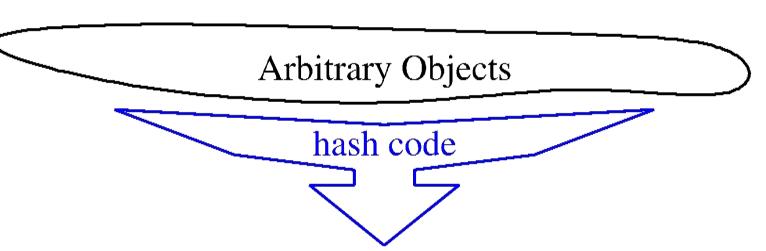
More on Hash Functions

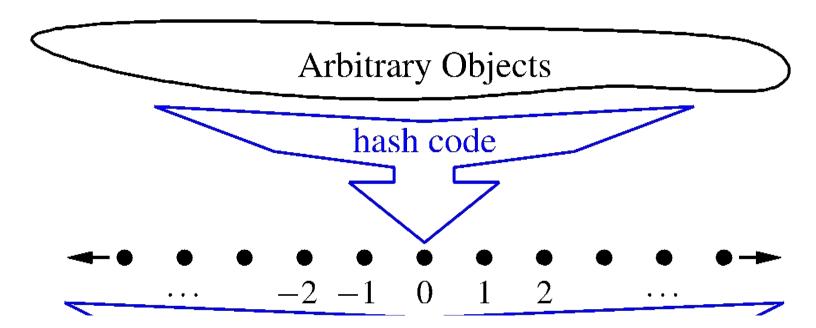
A hash function is usually specified as the composition of two functions:

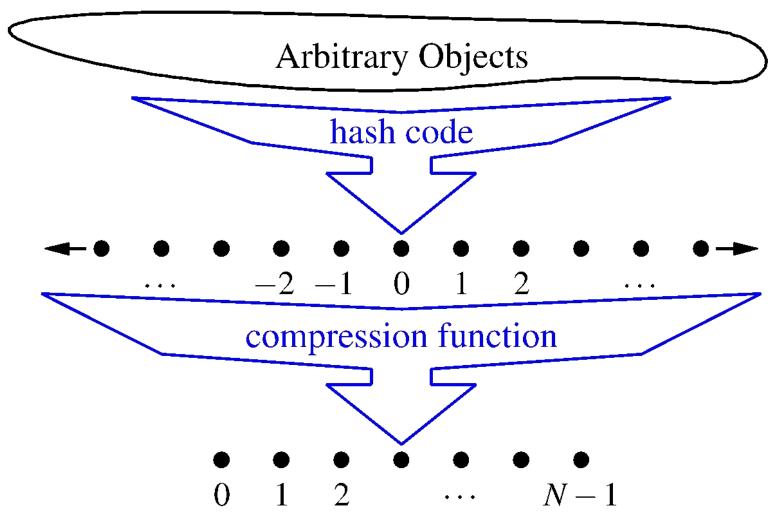
- hash code: h_1 : key \rightarrow integers
- compression: h_2 : integers $\rightarrow [0, N-1]$
- $\bullet \ h(x) = h_2(h_1(x))$

The goal is to "disperse" the keys in an appropriately random way

Arbitrary Objects







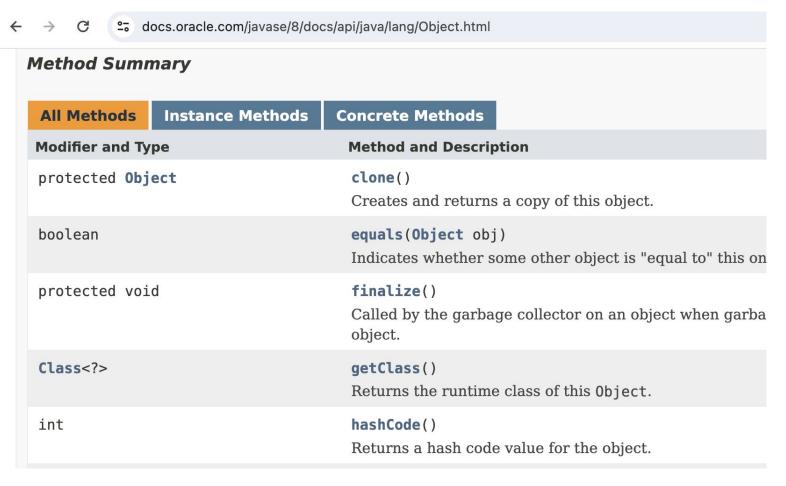
Hash Codes (h_1)

- h₁(k) takes an arbitrary key and computes an integer
 - Goal: collision resistant!
 - Need not be a fixed length or in fixed range [0, N)
 - Can even be negative

Hash Codes h₁

Idea 1: Memory Addresses

- use the memory address where the keys are stored
- default hash code for Java objects



Hash Codes (h_1)

What if our key is not an object?

- Integer cast: byte, short, int, char and float
- What about long and double??
 - Can't cast to int. We'll lose information!
 - COLLISIONS
 - Instead, partition bits into int components and combine them

Compression (h_2)

Why do we need compression?

Compression Idea 1: mod

 $h_2(x) = x \mod N$ forces output to be in range [0, N)

How should we choose N? primes!

Compression (h_2)

Compression Idea 2: Multiply Add and Divide (MAD)

```
h_2(x) = ((ax + b) \mod p) \mod N
```

```
where N is the capacity
```

p is a prime > N

a and b are [0, p)

a scales the range

b shifts the start

HashMap

Book's AbstractHashMap hash method uses:

```
h_1(k) = k.hashCode() // java memory address
```

$$h_2(x) = ((ax + b) % p) % N$$

Hash Maps

Efficient data structure that stores (Key, Value) pairs

Implements the Map ADT

- get (k): if the map M has an entry with key k, return its associated value; else, return null
- put (k, v): insert entry (k, v) into the map M; if key k is not already in M, then return null; else, replace old value with v and return old value associated with k
- remove (k): if the map M has an entry with key k, remove it from M and return its associated value; else, return null
- size(), isEmpty()
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Hash Maps

Implementation

Let's start with our ArrayMap and use hashes for indices

BE CAREFUL! % MEANS REMAINDER IN JAVA NOT MOD!

What should we do if there's a collision?

- For a first impl, let's just overwrite

Performance Analysis

	ArrayMap	Collision Resistant Hash Map
get		
put		
remove		

AbstractHashMap

```
\textbf{public abstract class } AbstractHashMap{<}K,V{>} \textbf{ extends } AbstractMap{<}K,V{>} \{
                                          // number of entries in the dictionary
      protected int n = 0;
     protected int capacity;
                                        // length of the table
     private int prime;
                                          // prime factor
      private long scale, shift;
                                         // the shift and scaling factors
      public AbstractHashMap(int cap, int p) {
        prime = p;
        capacity = cap;
        Random rand = new Random();
        scale = rand.nextInt(prime-1) + 1;
10
        shift = rand.nextInt(prime);
11
12
        createTable();
13
14
      public AbstractHashMap(int cap) { this(cap, 109345121); } // default prime
15
      public AbstractHashMap() { this(17); }
                                                                  // default capacity
     // public methods
      public int size() { return n; }
      public V get(K key) { return bucketGet(hashValue(key), key); }
      public V remove(K key) { return bucketRemove(hashValue(key), key); }
      public V put(K key, V value) {
        V answer = bucketPut(hashValue(key), key, value);
21
        if (n > capacity / 2) // keep load factor <= 0.5
          resize(2 * \text{capacity} - 1); // (or find a nearby prime)
        return answer;
25
```

AbstractHashMap

```
26
         private utilities
      private int hashValue(K key) {
27
28
        return (int) ((Math.abs(key.hashCode()*scale + shift) % prime) % capacity);
29
30
      private void resize(int newCap) {
31
        ArrayList < Entry < K,V >> buffer = new ArrayList <> (n);
32
        for (Entry<K,V> e : entrySet())
          buffer.add(e);
33
34
        capacity = newCap;
35
        createTable();
                                              based on updated capacity
36
        n=0:
                                              will be recomputed while reinserting entries
37
        for (Entry<K,V> e : buffer)
38
          put(e.getKey(), e.getValue());
39
40
         protected abstract methods to be implemented by subclasses
41
      protected abstract void createTable();
42
      protected abstract V bucketGet(int h, K k);
      protected abstract V bucketPut(int h, K k, V v);
43
44
      protected abstract V bucketRemove(int h, K k);
45
```

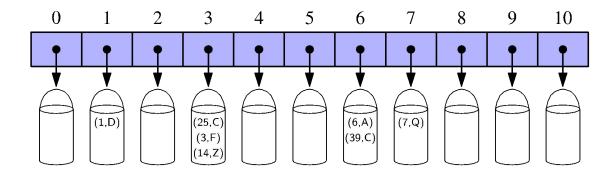
Handling Collisions

Handling Collisions

A hash function does not guarantee one-to-one mapping — no hash function does

One approach chaining:

When more than one key hash to the same index, we have a bucket Each index holds a collection of entries



Collision Handling

Collisions occur when elements with different keys are mapped to the same cell

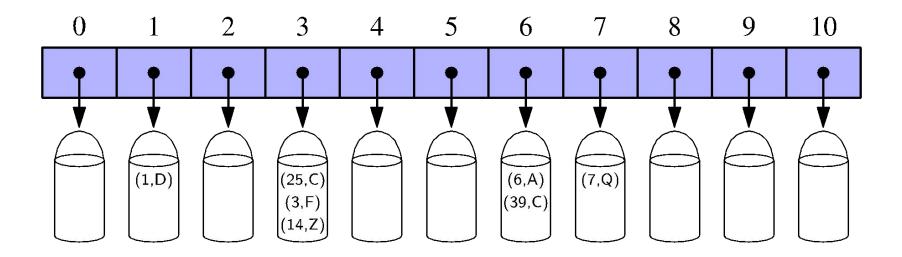
Separate Chaining: let each cell in the table point to a linked list of entries that map there

Simple, but requires additional memory besides the table

Let's implement a ChainHashMap

What data structure should we use for the buckets?

- LinkedList!



Summary

Hash Map:

- Efficient data structure with constant time* access, insertion, and removal
- * assuming no collisions or expansions

Hash Functions:

- Composition of h1 and h2
- h2 compresses output of h1 between 0 and N

Collision Handling Approach #2

Open Addressing and Probing

When a collision occurs, find an empty slot nearby to store the colliding element

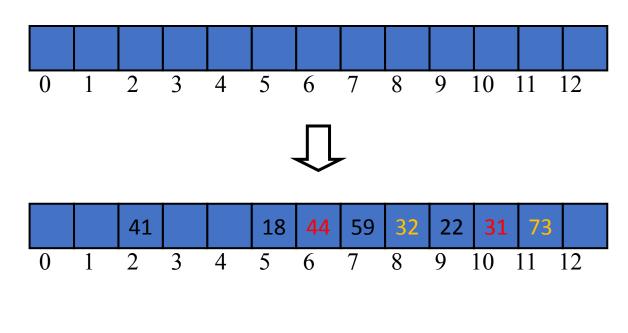
Open Addressing and Probing

- Example: h(x) = x%13
- insert 18(5), 41(2), 22(9), 44(5), 59(7), 32(6), 31(5), 73(8)

Keep "probing" (h(k)+1)%n (h(k)+2)%n

• • • •

(h(k)+i)%n until you find an empty slot!



ProbeHashMap

Let's implement a ProbeHashMap

Open Addressing and Probing

Linear Probing (what we just implemented):

Keep "probing" until you find an empty slot
 (h(k)+1) % n
 (h(k)+2) % n
 (h(k)+i) % n

 Colliding items cluster together – future collisions to cause a longer sequence of probes

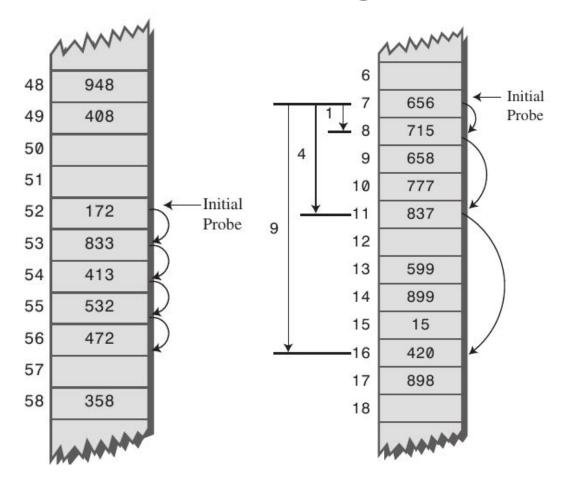
Open Addressing and Probing

Quadratic Probing:

where $f(i) = i^2$

```
    Keep "probing" until you find an empty slot
        (h(k)+f(1)) % n
        (h(k)+f(2)) % n
        ....
        (h(k)+f(i)) % n
```

Linear Probing vs Quadratic Probing



Linear Probing

Quadratic Probing

- Quadratic probing still creates large clusters!
- Unlike linear probing, they are clustered away from the initial hash position
- If the primary hash index is x, probes go to x+1, x+4, x+9, x+16, x+25 and so on, this results in Secondary Clustering

Approach #3: Double Hashing

Let's try to avoid clustering.

To probe, let's use a second hash function

 Keep "probing" until you find an empty slot (h(k)+f(1)) % n (h(k)+f(2)) % n

Where
$$f(i) = i * h'(k)$$

(h(k)+f(i)) % n

Approach #3: Double Hashing

```
Keep "probing" until you find an empty slot
      (h(k)+f(1)) % n
      (h(k)+f(2)) % n
      ....
      (h(k)+f(i)) % n
```

Where f(i) = i * h'(k)

A common choice for h'(k) = q - (k % q)where q is prime and < n

Example

Insert 18, 41, 22, 44, 59, 32, 31, 73

probe:

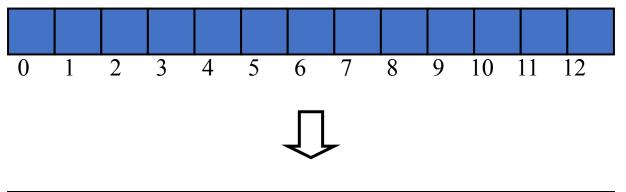
$$(h(k) + f(k)) % n$$

$$h(k) = k \% 13$$

$$f(k) = i * h'(k)$$

$$h'(k) = 7 - k \% 7$$

k	h(k)	h'(k)	Probes		
18	5	3	5		
41	2	1	2		
22	9	6	9		
44	5	5	5	10	
59	7	4	7		
32	6	3	6		
31	5	4	5	9	0
73	8	4	8		





Performance Analysis

	ChainHashMap Best Case	ChainHashMap Worst Case	ProbeHashMap Best Case	ProbeHashMap Worst Case
get				
put				
remove				

Which is better in practice?

Open Addressing vs Chaining

- Probing is significantly faster in practice
- locality of references much faster to access a series of elements in an array than to follow the same number of pointers in a linked list

Performance Analysis

	ArrayMap	HashMap with good hashing and good probing
get		
put		
remove		

Performance of Hashtable

	array	linked list	BST (balanced)	HashTable
search				
insert				
remove				

Load Factor

- HashMaps have an underlying array... what if it gets full?
 - For ChainHashMap collisions increase
 - For ProbeHashMap we need to resize!

Load Factor = # of elements stored / capacity

- A common strategy is to resize the hash map when the load factor exceeds a predefined threshold (often 0.75)
 - tradeoff between memory and runtime