# **Further Mathematics and Algorithms**

#### **Lesson 13:** Make a hash of it



Hash tables, separate chaining, open addressing, linear/quadratic probing, double hashing

#### **Outline**

- 1. Why Hash?
- 2. Separate Chaining
- 3. Open Addressing
  - Quadratic Probing
  - Double Hashing
- 4. Hash Set and Map



# **Content Addressable Memory**

- Suppose we have a list of objects which we want to look up according to its contents
- This is often referred to as associative memory structures
- A classical example would be a telephone directory
  - ★ We look up a name
  - ★ We want to know the number
- What data structure should we use?

#### **Lists and Trees**

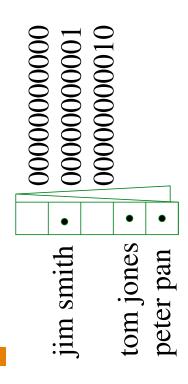
- To find an entry in a normal list takes  $\Theta(n)$  operations
- If we had a sorted list we could use "binary search" to reduce this to  $\Theta(\log(n))$ !
  - ★ We will study binary search later
  - $\star$  Maintaining an ordered list is costly  $(\Theta(n))$  insertions
- We could use a binary search tree!
  - $\star$  Search is  $\Theta(\log(n))$
  - $\star$  Insertion/deletion is  $\Theta(\log(n))$

# Thinking Outside the Box

- As with many data structures thinking about the problem differently can lead to much better solutions
- Let us consider the content we want to search on as a key!
- For telephone numbers the key would be the name of the person we want to phone
- We could get O(1) search, insertion and deletion if we used the key as an index into a big array!
- That is the key is a string of, say, 100 characters so can be represented by an 800 digit binary number.
- ullet We could look up the key in a table of  $2^{800}$  items

# **Hashing**

- This approach is slightly wasteful of memory
- Almost all memory locations would be empty!
- We can save on memory by folding up the table up onto itself



# **Hashing Codes**

- A hashing function hashCode(x) takes an object, x, and returns a positive integer, the hash code
- To turn the hash code into an address take the modulus of the table size

```
int index = abs(hashCode(x) % tableSize);
```

• If  $tableSize = 2^n$  we can compute this more efficiently using a mask

```
int index = abs(hashCode(x) & (tableSize -1));
```

### **Hashing Functions**

- Hashing functions take an object and return an integer
- Hashing functions aren't magic
  - ★ They tend to add up integers representing the parts of the object
- We want the integers to be close to random so that similar objects are mapped to different integers.
- Sometimes two objects will be mapped to the same address—this
  is known as a collision.
- Collision resolution is an important part of hashing

### **Hashing Strings**

A strings might be hashed using a function

```
unsigned long long hash(string const& s) {
  unsigned long long results = 12345;

for (auto ch = s.begin(); ch != s.end(); ++ch) {
    results = 127*results + static_cast<unsigned char>(*ch);
  }
  return results;
}
```

- The numbers 12345 and 127 is to try to prevent clashes
   —there
   are lots of alternatives
- What we want is that strings that might be similar receive very different hash codes

#### DIY

- The unordered\_set<T, Hash<T> > allows you to define your own hash function
- By default this is set to std::hash<T>(T)
- Not all classes have hash function defined so you will need to do this!
- Care is needed to make you hash function produce near random hash codes

#### **Outline**

- 1. Why Hash?
- 2. Separate Chaining
- 3. Open Addressing
  - Quadratic Probing
  - Double Hashing
- 4. Hash Set and Map



#### **Collision Resolution**

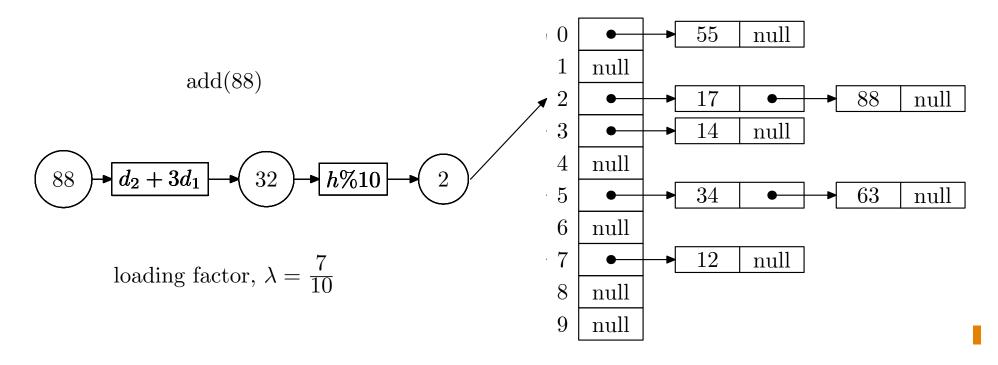
- Collisions are inevitable and must be dealt with
- There are two commonly used strategies
  - ★ Separate chaining—make a hash table of lists
  - ⋆ Open addressing—find a new position in the hash table
- Collisions add computational cost
- They occur when the hash table becomes full
- If the hash table becomes too full then it may need to be resized.

### Resizing a Hash Table

- Resizing a hash table is easy
  - ★ Create a new hash table of, say, twice the size
  - ★ Iterate through the old hash table adding each element to the new hash table
- Note that you have to recompute all the hash codes
- Resizing a hash table has a modest amortised cost, but can give you a very hiccupy performance
- The size of a hash table is a classic example of a memory-space versus execution time trade off—using bigger (sparser) hash tables speeds up performance

# **Separate Chaining**

 In separate chaining we build a singly-linked list at each table entry

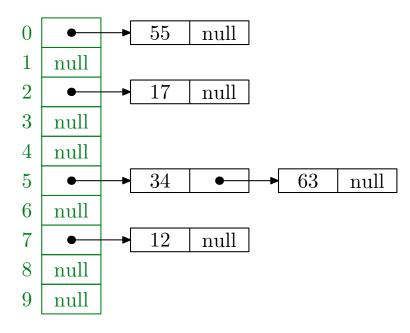


#### Search

- To find an entry in a hash table we again use the hash function on a key to find the table entry and then we search the list
- The time complexity depends on where objects are hashed.
- If the objects are evenly dispersed in the table, search (and insertion) is  $\Omega(1)$
- If the objects are hashed to the same entry in the hash table then search is  $O(n){
  m I}$
- Provided you have a good hashing function and the hash table isn't too full you can expect  $\Theta(1)$  average case performance

# **Iterating Over a Hash Table**

- To iterate over a hash table we
  - ★ Iterate through the array
  - ★ At each element we iterate through the linked list
- The order of the elements appears random
- This becomes more efficient as the table becomes fuller



55, 17, 34, 63, 12

#### **Outline**

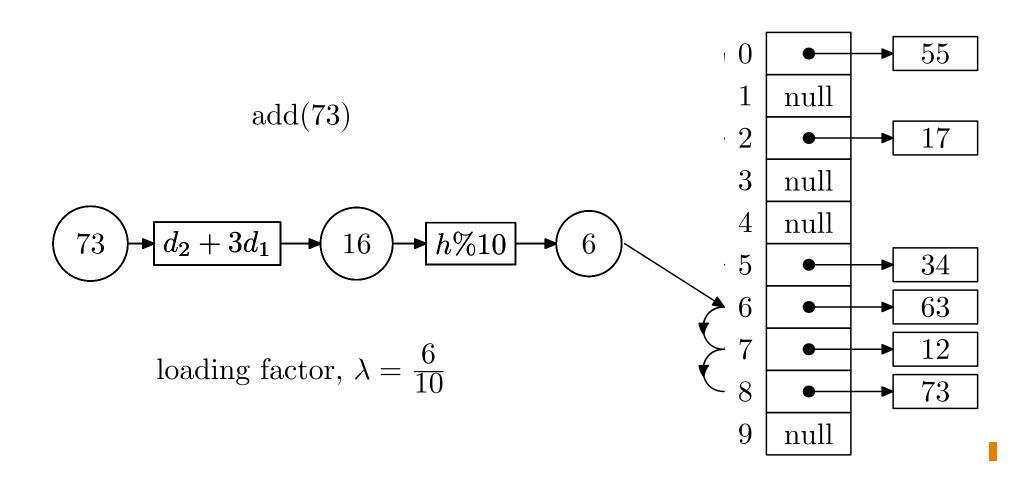
- 1. Why Hash?
- 2. Separate Chaining
- 3. Open Addressing
  - Quadratic Probing
  - Double Hashing
- 4. Hash Set and Map



# **Open Addressing**

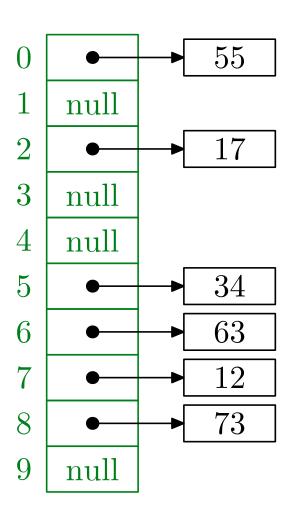
- In open addressing we have a single table of objects (without a linked-list)
- In the case of a collision a new location in the table is found
- The simplest mechanism is known as linear probing where we move the entry to the next available location.

# **Linear Probing**

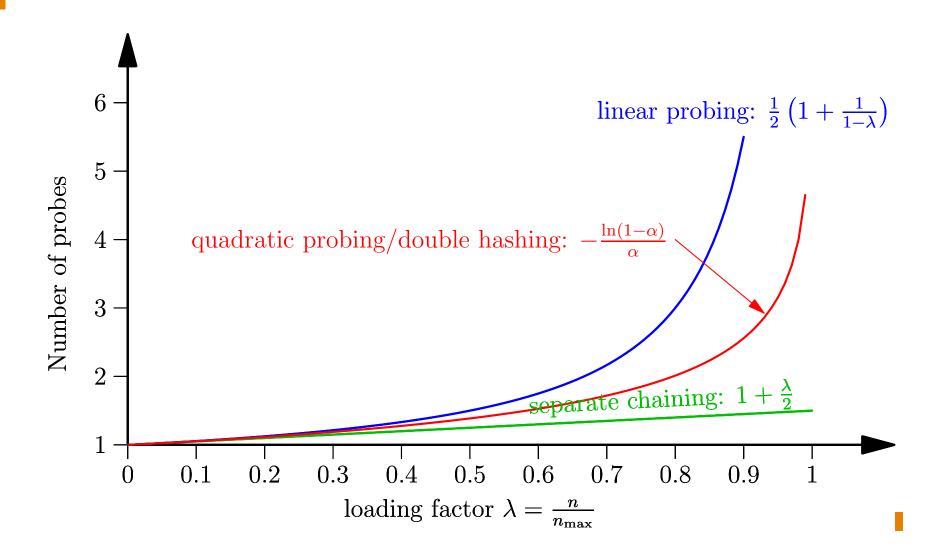


# **Linear Probing Pile Up**

- The entries will tend to pile up or cluster—this is sometimes referred to as primary clustering
- Clusters become worse as the number of entries grow!
- Clusters will increase the number of probes needed to find an insert location
- The proportion of full entries in the table is known as the loading factor



### Reducing Number of Probes



To avoid clustering we can use quadratic probing or double hashing

### **Quadratic Probing**

- In quadratic probing we try the locations  $h(x) + d_i$  where h(x) is the original hash code and  $d_i = i^2$
- That is we takes steps 1, 4, 9, 16, . . . .
- Quadratic probing prevents primary clustering so dramatically decreases the number of probes needed to find a free location when the table is reasonably full
- One problem is that if we are unlucky we might not be able to add an element to the hash table even if the table isn't full
- However, if the size of the table is prime then quadratic probing will always find a free position provided it is not more than half full

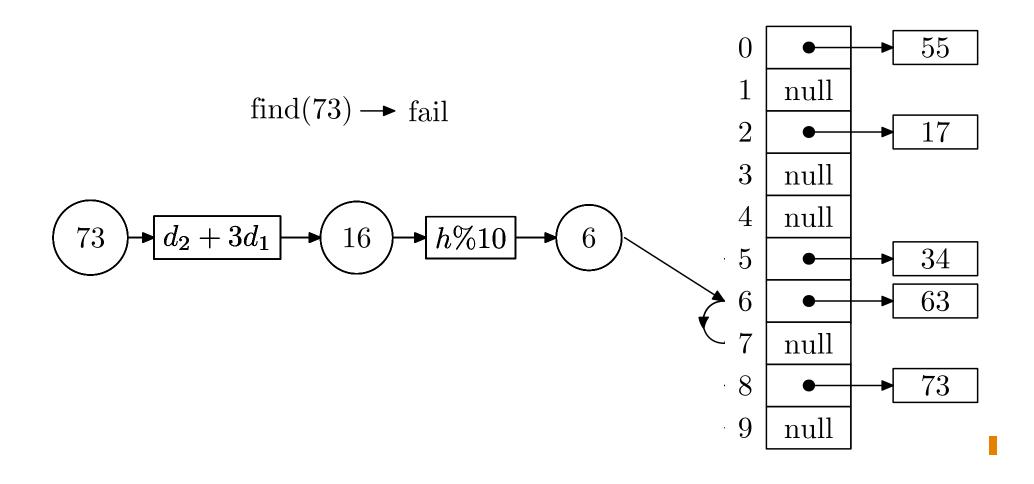
# **Double Hashing**

- An alternative strategy is to known as double hashing where the locations tried are  $h(x) + d_i$  where  $d_i = i \times h_2(x)$
- $h_2(x)$  is a second hash function that depends on the key
- A good choice is  $h_2(x) = R (x \mod R)$  where R is a prime smaller than the table size
- It is important that  $h_2(x)$  is not a divisor of the table size
  - ★ Either make sure the table size is prime or
  - $\star$  Set the step size to 1 if  $h_2(x)$  is a divisor of the table size

#### **Problems with Remove**

- For all open addressing hash systems removing an entry is a problem!
- ullet Remember our strategy to find an input x is
  - 1. Compute the array index based on the hash code of x
  - 2. If the array location is empty then the search fails
  - 3. If the array location contains the key the search succeeds
  - 4. otherwise find a new location using an open addressing strategy and go to 2
- If we remove an entry then find might reach an empty location which was previously full
- This can prevent us finding a true entry

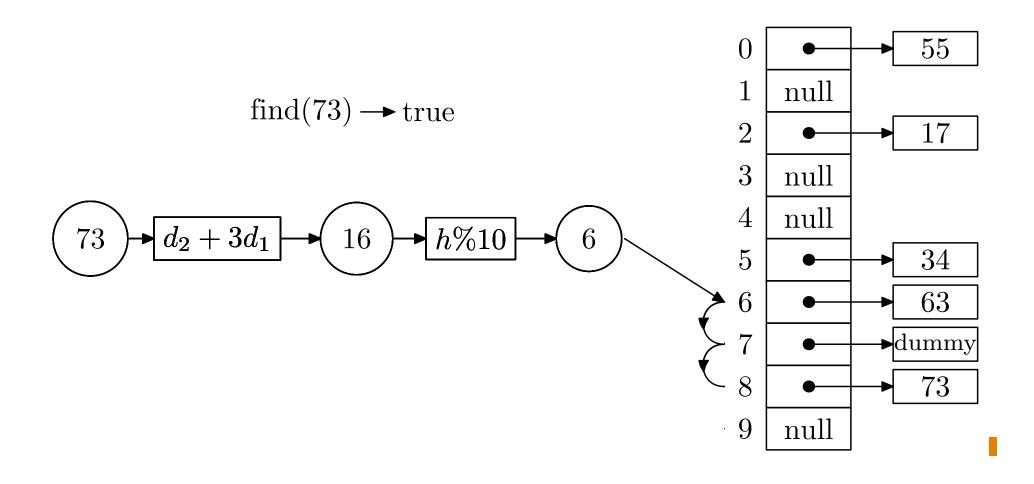
# **Linear Probing Example**



### Lazy Remove

- One easy fix is to mark the deleted table with a special entry.
- A find method would consider this entry as full
- An iterator would ignore this entry
- An insert operator could insert a new entry in these special locations

### Lazy Remove in Action



#### **Outline**

- 1. Why Hash?
- 2. Separate Chaining
- 3. Open Addressing
  - Quadratic Probing
  - Double Hashing
- 4. Hash Set and Map



### What Strategy to Use?

- Most libraries including the STL (and the Java Collection class)
  use separate chaining
- This has the advantage that its performance does not degrade badly as the number of entries increase
- This reduces the need to resize the hash table
- The C++ standard did not include a hash table until C++11
   □ lthough very good hash tables existed in C++I

# **Hash Sets and Maps**

- C++ also provides an unordered\_map<Key, V> class
- It's performance is asymptotically superior to map, O(1) rather than  $O(\log(n))$ !
- Hash functions can take time to compute (it is often  $O(\log(n))$ ) so unordered\_sets might not be faster than sets!
- One major difference is that the iterator for sets return the elements in order, undordered\_set's iterator doesn't

### **Applications**

- Hash tables are used everywhere
- E.g. most databases use hash tables to speed up search
- In many document applications hash tables will be being generated in the background
- Content addressability is ubiquitous to many application where hash tables are used as standard

#### Lessons

- Hash tables are one of the most useful tools you have available.
- They aren't particularly difficult to understand, but you need to know about
  - ★ hashing functions
  - ★ collision strategies
  - ⋆ performance (i.e. when they work)