Hashing

Class 19

Dictionary

- dictionary is the name for a container with the following characteristics
 - a key which identifies an entry
 - a value corresponding with the key consisting of one or more fields of data
 - i.e., it contains key-value pairs
- and with the following behaviors
 - find a specific key in the dictionary
 - Boolean find
 - retrieve the value associated with a particular key
 - insert a new key and its value into the dictionary
 - if the key exists, overwrite the value associated with it
 - remove a key and its associated value from the dictionary
- is_empty, etc

Implementation

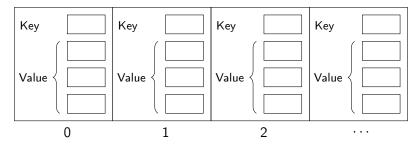
- a dictionary can be implemented in many different ways, e.g.,
 - an array of key-value pairs maintained in key order
 - a binary search tree of key-value pairs ordered by key
- but usually a dictionary implies a hashed implementation

Hashing

- there are a number of variations of hashing
 - static vs dynamic
 - open chaining vs various probing strategies
- we could spend weeks on all the variations
- we will only consider static open chaining

Hash Table

- a hash table is an array (vector) of structures
- the array size *m* is fixed (in static hashing)
- one field is the key
- the other field(s) are collectively the value
- there are *n* possible keys



Hash Function

- a hash function is used to decide where to place a key-value pair in the array
- a hash function is a function:

$$f(\text{key}) = h$$

where

$$h \in \{0, 1, \ldots, m-1\}$$

• there are many hash functions

Hash Functions

if the key is an unsigned integer, a possible hash function is

$$f(\text{key}) = \text{key mod } 33$$

here, m=33, and every possible unsigned integer is mapped to an index in the range 0...32

if the key is an ASCII string, a possible hash function is

$$f(\text{key}) = \text{ord}(\text{toupper}(\text{key.at}(0))) - 0x41$$

here, m=26, and every possible alphabetic string is mapped to an index in the range 0...25

Hash Functions

- literally thousands of hash functions have been invented
- designing them is an art form
- the primary desiderata of the hash function are:
 - it must distribute the keys as evenly as possible among all m indices
 - it must be fast to compute
- the hash function is tightly coupled to the size *m* of the hash table

Analysis

- assuming
 - a perfect hash function
 - correctly sized hash table
 - well-behaved data
- push: $T(n) \in \Theta(1)$
- pop: $T(n) \in \Theta(1)$
- find: $T(n) \in \Theta(1)$

Problem

- alas, the world is not perfect
- what if the hash table size m is smaller than the number of keys n?

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- alas, the world is not perfect
- what if the hash table size m is smaller than the number of keys n?
- by the pigeonhole principle at least two keys will be mapped to the same index
- this is a collision (aka hash clash)
- collisions can happen even if m is larger than n (how?)
- thus, we must handle collisions

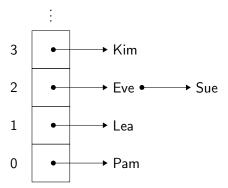
Collision Example

- keys are ASCII strings
- hash table size is 13
- hash function is sum(ord(characters)) mod table_size

Name	Function	Index
Eve	$(69 + 118 + 101) \mod 13$	2
Lea	$(76 + 101 + 97) \mod 13$	1
Kim	$(75 + 105 + 109) \mod 13$	3
Pam	$(80 + 97 + 109) \mod 13$	0
Sue	$(83 + 117 + 101) \mod 13$	2

- Eve and Sue collide
- both map to the same index

- one strategy to handle collisions is open chaining
- this is the only strategy we'll consider
- in open chaining, the hash table does not actually store entries
- rather, it stores linked lists of entries



find

- 1. compute hash of key
- 2. go to the linked list of that index
- 3. search linearly for key in the list
- push (both insert and replace)
 - 1. compute hash of key
 - 2. go to the linked list of that index
 - 3. search linearly for key in the list
 - 4. insert or update entry

pop

- 1. compute hash of key
- 2. go to the linked list of that index
- 3. search linearly for key in the list; delete entry if found

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 - can tolerate an unlimited number of collisions
- cons
 - requires the machinery of linked list
 - list could be quite long how long?

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- insertion:
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- now what is analysis?
- insertion: $T(n) \in O(n/m)$
- deletion: $T(n) \in O(n/m)$
- search: $T(n) \in O(n/m)$
- the ratio $\alpha = n/m$ is called the load factor

Load Factor

- the load factor should be close to 1
- what if it's too small (m is big compared to n)?
 lots of unused wasted space in the hash table

Rehashing

- if the table gets too full
- 70% of the entries are occupied
- running time for operations increases, performance decreases
- solution: allocate a new table of size next prime number more than twice as big as the current table
- create a new hash function appropriate for the new size
- re-hash every element in the old table to the new table
- de-allocate old table

Hash Table Size

- truism: always use hash table whose size is a prime number
- why?
- if your keys are evenly distributed, so that every possible key is equally likely, then it makes no difference
- prime or non-prime size is irrelevant
- however, in real life many key sets are not evenly distributed
- imagine you are keeping a symbol table of objects based on where they are stored in memory — their memory address
- your computer's word size is 4 bytes
- every address is a multiple of 4
- if m happens to also be a multiple of 4, $\frac{3m}{4}$ table entries are empty, and all n entries collide in the remaining m/4 slots

Prime Table Size

- every key that has a common factor with table size m will be hashed to a location that is a multiple of the common factor
- to minimize collisions we must reduce the number of common factors between m and the set of keys
- how? choose *m* to be a number that has very few factors
- what kind of number has very few factors?

Prime Table Size

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- how? choose m to be a number that has very few factors
- what kind of number has very few factors? a prime number!

Java Hashcode

- the Java Object class has the method int hashCode()
- you can generate a hash code for any object
- when hashCode() is invoked on the same object more than once during an execution of a Java application, the hashCode method must return the same integer
- if two objects are equal according to the equals() method, then hashCode() on each of the two objects must produce the same integer result
- it is not required that if two objects are unequal according to the equals() method, then calling the hashCode method on each of the two objects must produce distinct integer results — WHY?

Java String Hashcode

Java uses this as its string hash function

$$h(s) = \sum_{i=0}^{n-1} s[i] \times 31^{n-1-i}$$

where the sum is 32-bit addition and s[i] is the utf-16 character code of the ith character of the string

- what are the keys?
- how many indices can this generate?

Hash Table Implementations

- Java HashMap, uses hashCode() to place items
- C++ map
- PHP every array is actually a hash table with keys limited to integers and strings
- Python dictionary
- Perl hash
- JavaScript all object fields are associative arrays

Array vs HashMap

- what is stored
- memory used
- ordering of elements
- duplicate elements
- access strategy

Article

https://www.techrepublic.com/article/programming-languages-facebook-open-sources-its-fast-f14-hash-table-written-in-c/