Lecture 11: Dictionaries

Sequential Unfoldment of Knowledge

Wholeness Statement

The Dictionary ADT stores a searchable collection of key-value items that represents either an unordered or an ordered collection. Hashing solves the problem of item-lookup by providing a table whose size is not unreasonably large, yet it can store a large range of keys such that the value associated with each key can be found quickly (O(1)). Science of Consciousness provides systematic techniques for accessing and experiencing total knowledge of the Universe to enhance individual life. We experience this each day when we experience the silent, unbounded state of consciousness during our daily TM.

The Dictionary ADT

Our main focus in this part of the course is searching algorithms and how to organize data so it can be searched efficiently.

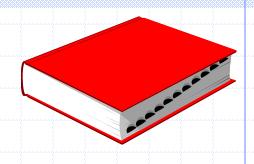
Review

- Recall that a Priority Queue contains key-value items
 - So does a Dictionary
- Priority Queue items were organized in three different ways
 - Unordered Sequence
 - Sorted Sequence
 - Binary Tree (Heap)
- Dictionaries can be organized in similar ways
 - Log file or Hashtable (unordered sequence)
 - Lookup Table (ordered sequence)
 - Binary Search Tree (ordered tree)

Two Types of Dictionaries

- 1. Unordered
- 2. Ordered
- Stores items, i.e., key-value pairs
- Both ordered and unordered Dictionaries search for a key to identify/locate the specific value(s) associated with that key
- For the sake of generality, multiple items could have the same key (e.g., log files), but generally we will require each item to have a unique key

Unordered Dictionary ADT



- The dictionary ADT models a searchable collection of key-value items
- The main operations of a dictionary are searching, inserting, and deleting items
- Multiple items with the same key are allowed in the log-file, but not generally
- Applications:
 - address book
 - credit card authorization
 - mapping host names (e.g., cs16.net) to internet addresses (e.g., 128.148.34.101)

Dictionary ADT methods:

- findValue(k): if the dictionary has an item with key k, then returns that item's value, else, returns the special value NO_SUCH_KEY
- insertItem(k, o): inserts item(k, o) into the dictionary
- removeltem(k): if the dictionary has an item with key k, removes item from the dictionary and returns its value, else returns the special element NO_SUCH_KEY
- size(), isEmpty()
- keys(), values(), items()

Log Files

- A log file (or audit trail) is a dictionary implemented by means of an unsorted sequence
 - Items are stored in the dictionary in a sequence in arbitrary order
 - Based on doubly-linked lists or a circular array

Performance:

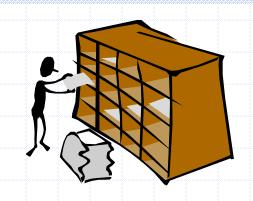
- insertItem takes O(1) time since we can insert the new item at the beginning or at the end of the sequence
- findValue and removeItem take O(n) time since in the worst case (the item is not found), we have to traverse the entire sequence to look for an item(s) with the given key
- analogous to the Priority Queue that used an unordered sequence

Log File

- For dictionaries on which insertions are the most common operations, while searches and removals are rarely performed (e.g., historical record of logins to a workstation)
- What do we do if we need to do frequent searches and removals in a large unordered dictionary?

Hash Tables

Hash Tables and Hash Functions



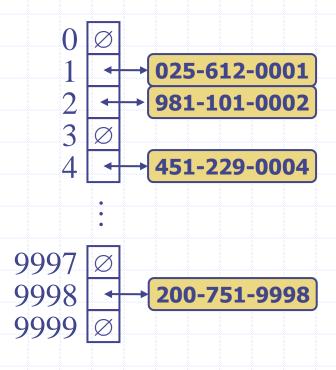
- A hash table for a given key type consists of
 - Hash function h
 - Array (called table) of size N
- lacktriangle A hash function h maps keys of a given type to integers in a fixed interval [0, N-1]
- Example: $h(k) = k \mod N$ is a hash function for integer keys
- lacktriangledark The integer h(k) is called the hash value of key k

Goals of Hash Functions

- 1. Store item (k, o) at index i = h(k) in the table
- 2. Avoid collisions as much as possible
 - Collisions occur when two different keys hash to the same index i
 - The average performance of hashing depends on how well the hash function distributes the set of keys (i.e., avoids collisions)

Example

- Design a hash table for a dictionary storing items (SSN, Name), where SSN (social security number) is a nine-digit positive integer
- Our hash table uses an array of size N = 10,000 and the hash function
 h(x) = last four digits of x



Hash Functions



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

Hash code map:

 h_1 : keys \rightarrow integers

Compression map:

 h_2 : integers $\rightarrow [0, N-1]$

The hash code map is applied first, and the compression map is applied next on the result, i.e.,

$$\boldsymbol{h}(\boldsymbol{x}) = \boldsymbol{h}_2(\boldsymbol{h}_1(\boldsymbol{x}))$$

The goal of the hash function is to "disperse" the keys in an apparently random way

Hash Code Maps



Memory address:

- We reinterpret the memory address of the key object as an integer
 - (default hash code of all Java objects)
- Good in general, except for numeric and string keys

Integer cast:

- We reinterpret the bits of the key as an integer
- Suitable for keys of length less than or equal to the number of bits of the integer type (e.g., byte, short, int, and float in Java)

Component sum:

- We partition the bits of the key into components of fixed length (e.g., 16 or 32 bits)
- Then we sum the components (ignoring overflows)
- Suitable for numeric keys of fixed length greater than or equal to the number of bits of the integer type (e.g., long and double in Java)





Polynomial accumulation:

We partition the bits of the key into a sequence of components of fixed length (e.g., 8, 16 or 32 bits)

$$\boldsymbol{a}_0 \boldsymbol{a}_1 \dots \boldsymbol{a}_{n-1}$$

• We evaluate the polynomial $p(z) = a_0 + a_1 z + a_2 z^2 + ... + a_{n-1} z^{n-1}$

at a fixed value z, ignoring overflows

Especially suitable for strings
 (e.g., the choice z = 33 gives
 at most 6 collisions on a set of
 50,000 English words)

- Polynomial p(z) can be evaluated in O(n) time using Horner's rule:
 - The following polynomials are successively computed, each from the previous one in O(1) time

$$p_0(z) = a_{n-1}$$

 $p_i(z) = a_{n-i-1} + zp_{i-1}(z)$
 $(i = 1, 2, ..., n-1)$

• We have
$$p(z) = p_{n-1}(z)$$

Compression Maps



Division:

- $\bullet h_2(y) = y \bmod N$
- The size N of the hash table is usually chosen to be a prime
 - The reason has to do with number theory and is beyond the scope of this course

- Multiply, Add and Divide (MAD):
 - $\bullet h_2(y) = (ay + b) \bmod N$
 - a and b are nonnegative integers such that $a \mod N \neq 0$
 - Otherwise, every integer would map to the same value b

Main Point

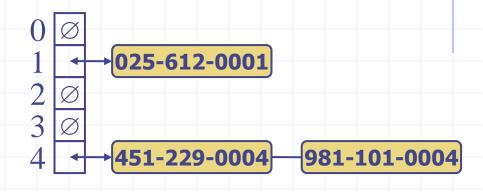
1. The hash function solves the problem of fast table-lookup, i.e., it allows the value associated with each key to be accessed quickly (in *O*(1) expected time). A hash function is composed of a hash code function and a compression function that transforms (in constant time) each key into a specific location in the table.

Science of Consciousness: Through a process of self-referral, the unified field sequentially transforms itself into all the values of creation without making mistakes.





Collisions occur when different elements are mapped to the same cell



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

 Chaining is simple, but requires additional memory outside the table

Load Factors and Rehashing

- Load factor is n/N where n is the number items in the table and N is the table size
- When the load factor goes above .75, the table is resized and the items are rehashed

Linear Probing

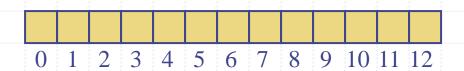


- Open addressing: the colliding item is placed in a different cell of the table
- Linear probing handles collisions by placing the colliding item in the next (circularly) available table cell
- Each table cell inspected is referred to as a "probe"
- Colliding items lump together, causing future collisions to cause a longer sequence of probes

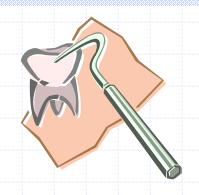
Linear Probing (§2.5.5)



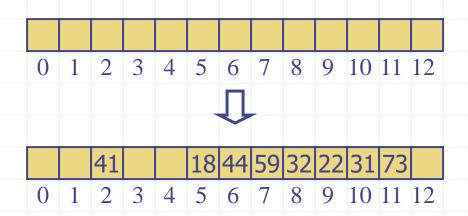
- **Exercise:**
 - $h(x) = x \mod 13$
 - Insert keys 18, 41, 22, 44, 59, 32, 31, 73, in this order



Linear Probing



- Example:
 - $h(x) = x \mod 13$
 - Insert keys 18, 41, 22, 44, 59, 32, 31, 73, in this order



Search with Linear Probing



- Consider a hash table

 A that uses linear

 probing
- findElement(k)
 - We start at cell h(k)
 - We probe consecutive locations until one of the following occurs
 - An item with key k is found, or
 - An empty cell is found, or
 - N cells have been unsuccessfully probed

```
Algorithm findElement(k)
i \leftarrow h(k)
p \leftarrow 0
do
   x \leftarrow (i + p) \mod N
   c \leftarrow A[x]
   if c = \emptyset then
      return NO_SUCH_KEY
   else if c.key() = k then
       return c.element()
   else
      p \leftarrow p + 1
while p < N
return NO SUCH KEY
```

Updates with Linear Probing

To handle insertions and deletions, we introduce a special object, called *AVAILABLE*, which replaces deleted elements

removeItem(k)

- We search for an item with key k
- If such an item (k, o) is found, we replace it with the special item AVAILABLE and we return element o
- Else, we return
 NO_SUCH_KEY

insert Item(k, o)

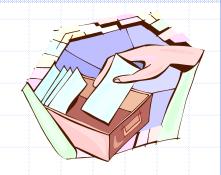
- We throw an exception if the table is full
- We start at cell h(k)
- We probe consecutive cells until one of the following occurs
 - A cell *i* is found that is either empty or stores AVAILABLE, or
 - N cells have been unsuccessfully probed
- We store item (k, o) in cell i

Quadratic Probing

- Start with the hash value i=h(k),
- Then search A[(i + j²) mod N]
 for j = 0, 1, 2, ... until an empty slot is found
- Disadvantages
 - Complicates removal even more
 - Secondary clustering

Double Hashing

- Double hashing uses a secondary hash function d(k) and handles collisions by placing an item in the first available cell of the series $(i + j*d(k)) \bmod N$ for j = 0, 1, ..., N-1
- The secondary hash function d(k) cannot have zero values
- The table size *N* must be a prime to allow probing of all the cells



Common choice of compression map for the secondary hash function:

$$\mathbf{d}_2(\mathbf{k}) = \mathbf{q} - (\mathbf{k} \bmod \mathbf{q})$$

where

- q < N
- q is a prime
- The possible values for $d_2(\mathbf{k})$ are

Example of Double Hashing

Consider a hash table storing integer keys that handles collision with double hashing

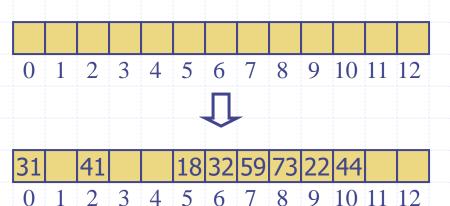
$$N = 13$$

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

$$\bullet d(k) = 7 - (k \bmod 7)$$

Insert keys 18, 41, 22, 44, 59, 32, 31, 73, in this order

	<u> </u>	4			- A
k	h(k)	d(k)	Prol	oes	
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		
18 41 22 44 59 32 31	5	4	5	9	0
73	8	4	8		



Search



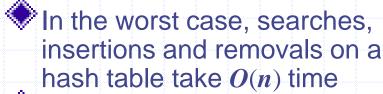
Quadratic Probing

```
Algorithm findElement(k)
i \leftarrow h(k)
p \leftarrow 0
do
   x \leftarrow (i + p^2) \mod N
   c \leftarrow A[x]
   if c = \emptyset then
      return NO SUCH KEY
   else if c.key() = k then
       return c.element()
   else
      p \leftarrow p + 1
while p < N
return NO_SUCH_KEY
```

Double Hashing

```
Algorithm findElement(k)
i \leftarrow h(k)
p \leftarrow 0
do
   x \leftarrow (i + p*d(k)) \mod N
   c \leftarrow A[x]
   if c = \emptyset then
      return NO_SUCH_KEY
   else if c.key() = k then
       return c.element()
   else
      p \leftarrow p + 1
while p < N
return NO_SUCH_KEY
```

Performance of Hashing

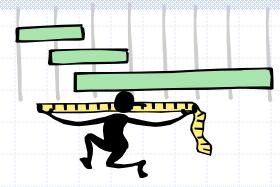


The worst case occurs when all the keys inserted into the dictionary collide

The load factor $\alpha = n/N$ affects the performance of a hash table

Assuming that the hash values are like random numbers, it can be shown that the expected number of probes for an insertion with open addressing is

 $1/(1-\alpha)$



- The expected running time of all the dictionary ADT operations in a hash table is O(1)
- In practice, hashing is very fast provided the load factor is not close to 100%
- Applications of hash tables:
 - small databases
 - compilers
 - browser caches

Universal Hashing

- If allowed to pick the keys to be hashed, then a malicious adversary can choose n keys that all hash to the same slot
 - Any fixed hash function is vulnerable to this sort of worstcase behavior
- The only effective way to improve the situation
 - Choose the hash function randomly in a way that is independent of the keys to be stored
- This approach is called universal hashing
 - The hash function is chosen randomly at beginning of execution
- Yields good performance no matter what keys are chosen by an adversary

Universal Hashing



- A family of hash functions is universal if, for any 0≤j,k≤M-1, Pr(h(j)=h(k)) ≤ 1/N.
- ◆ Keys are in the range [0, M-1]
- A hash function maps to the range [0, N-1]

- Theorem: The set of all functions, h, as defined here, is universal.
- Choose p as a prime between M and 2M.
- Randomly select 0<a<p and 0<b<p, and define h(k)=(ak+b mod p) mod N

Main Point

2. A hash table is an example of a highly efficient implementation of an unordered Dictionary ADT (its operations have expected complexity O(1)).

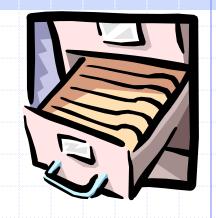
Science of Consciousness: Access to Pure Consciousness is simple, effortless, and spontaneous through the introduction of the proper techniques.

Ordered Dictionaries

Ordered Dictionaries



- Keys are assumed to come from a total order.
- Positions in a Dictionary now have two methods instead of element(), i.e., we now have
 - key()
 - value()
- Iterators return objects in order by associated key:
 - keys()
 the iterator iterates through the keys in sorted order
 - values()
 the iterator iterates through the values in the sorted order of the associated key
 - items()
 the iterator iterates through the items in the sorted order of the item's key



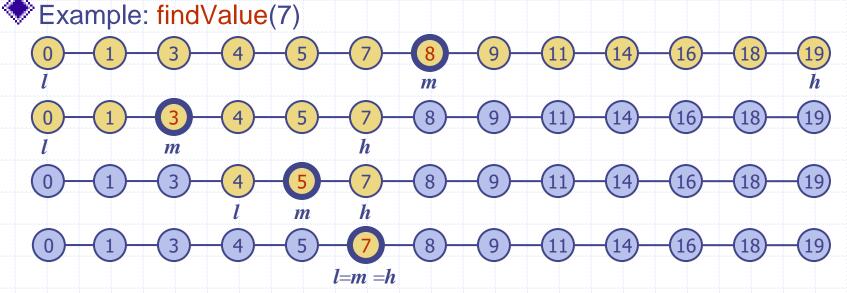
Analogous to the Priority Queue implemented as a sorted Sequence

- A dictionary implemented by means of a sorted sequence
 - store the items of the dictionary in an array-based sequence, sorted by key
 - uses an external comparator for comparing keys
- How should we lookup and find the value associated with a given key?
 - Hint: my favorite Dictionary because it uses my favorite algorithm

Binary Search



- Binary search performs operation findValue(k) on a dictionary implemented by means of an array-based sequence, sorted by key
 - similar to the high-low game (Prune and Search Strategy)
 - at each step, the number of candidate items is reduced by half
 - terminates after O(log n) steps



Binary Search Algorithm (iterative)

```
Algorithm BinarySearch(S, k):
 Input: An ordered vector S storing n items, accessed by keys()
 Output: An element of S with key k.
 low \leftarrow 0
 high \leftarrow S.size() - 1
 while low ≤ high do
    mid \leftarrow floor((low + high)/2)
     if k < key(mid) then
       high \leftarrow mid − 1
     else if k = key(mid) then // exit once the key is found
       return value(mid)
     else
       low \leftarrow mid + 1
 return NO_SUCH_KEY
```

Binary Search Algorithm (can be done recursively)

Algorithm BinarySearch(S, k, low, high):

Input: An ordered vector S storing n items, accessed by keys()

Output: An element of S with key k and rank between low & high.

```
if low > high then
  return NO_SUCH_KEY
else
  mid ← floor((low + high)/2)
  if k < key(mid) then
   return BinarySearch(S, k, low, mid-1)
  else if k = key(mid) then
  return value(mid)
  else
  return BinarySearch(S, k, mid + 1, high)</pre>
```

Binary Search Algorithm (improved, fewer key compares)

```
Algorithm BinarySearch(S, k):
 Input: An ordered vector S storing n items, accessed by keys()
 Output: An element of S with key k.
 low \leftarrow 0
 high \leftarrow S.size() - 1
 while low ≤ high do // always does log n iterations
    mid \leftarrow floor((low + high)/2)
     if k < key(mid) then // one key comparison per iteration
       high \leftarrow mid − 1
     else
       low ← mid // note that mid has not been eliminated yet
 if k = key(mid) then // check for equality after the loop
       return value(mid)
 else return NO_SUCH_KEY
```

- A dictionary implemented by means of a sorted sequence
 - store the items of the dictionary in an array-based sequence, sorted by key
 - use an external comparator for the keys
- Performance:
 - findValue
 - insertItem

removeltem

- A dictionary implemented by means of a sorted sequence
 - store the items of the dictionary in an array-based sequence, sorted by key
 - use an external comparator for the keys
- Performance:
 - find Value takes $O(\log n)$ time, using binary search
 - insertItem takes O(n) time since, in the worst case, we have to shift n/2 items to make room for the new item
 - removeItem takes O(n) time since, in the worst case, we have to shift n/2 items to compact the items after the removal

- Effective only
 - for dictionaries of small size or
 - for dictionaries on which
 - searches are the most common operation, and
 - insertions and removals are rarely performed
 - (e.g., credit card authorizations)

What do we do if this is not the case?

Main Point

3. A Lookup Table is an example of an ordered Dictionary ADT allowing elements to be efficiently accessed in order by key. When implemented as an ordered sequence, searching for a key is relatively efficient, O(log n), but insertion and deletion are not, O(n).

Science of Consciousness: The unified field of natural law always operates with maximum efficiency.

Connecting the Parts of Knowledge with the Wholeness of Knowledge

- A hash table is a very efficient way of implementing an unordered Dictionary ADT; the running time of search, insertion, and deletion is expected O(1) time.
- 2. To achieve efficient behavior of the hash table operations takes a careful choice of table size, load factor, hash function, and handling of collisions.

- 3. Transcendental Consciousness is the silent field of perfect efficiency and frictionless flow for coordinating all activity in the universe.
 - 4. Impulses within Transcendental
 Consciousness: The dynamic natural laws
 within this unbounded field create and maintain
 the order and balance in creation, all
 spontaneously without effort.
- 5. Wholeness moving within itself: In Unity Consciousness, the diversity of creation is experienced as waves of intelligence, perfectly efficient fluctuations of one's own self-referral consciousness.