Hashing





• The Dictionary

- a dictionary (table) is an abstract model of a database
- a dictionary stores key-element pairs
- the main operation supported by a dictionary is searching by key

Hashing





- Applications
 - Telephone directory
 - Library catalogue
 - Books in print: key ISBN
 - FAT (File Allocation Table)

ADT

```
template <class K, class E>
class Dictionary {
public:
```

virtual bool IsEmpty () const = 0; virtual pair<K,E>* Get(const K&) const = 0; virtual void Insert(const pair<K,E>&) = 0; virtual void Delete(const K&) = 0;

};

Implementing a Dictionary with a Sequence

- unordered sequence
 - searching and removing takes O(n) time
 - inserting takes O(1) time
 - applications to log files (frequent insertions, rare searches and removals) 34 14 12 22 18

Implementing a Dictionary with a Sequence

- *array-based ordered sequence* (assumes keys can be ordered)
- - searching takes O(log n) time *(binary search)*
 - inserting and removing takes O(n) time
 - application to look-up tables(frequent searches, rare insertions and removals)

Other Implementations?

- Binary search tree $- O(h) \rightarrow O(n)$
- Balanced search trees
 O(logN)
- Key comparison based
- Can we do better?

Application

- China Telecom is a large phone company, and they want to provide enhanced caller ID capability:
 - given a phone number, return the caller's name
 - phone numbers are in the range 0 to $R = 10^{10}$ –1
 - n is the number of phone numbers used
 - want to do this as efficiently as possible

Bucket Array

- Each cell is thought of as a bucket or a container
 - Holds key element pairs
 - In array A of size N, an element e with key k is inserted in A[k].

	(null)	(null)		Roberto	 (null)	
000	00000000		4109321568		9999999999	

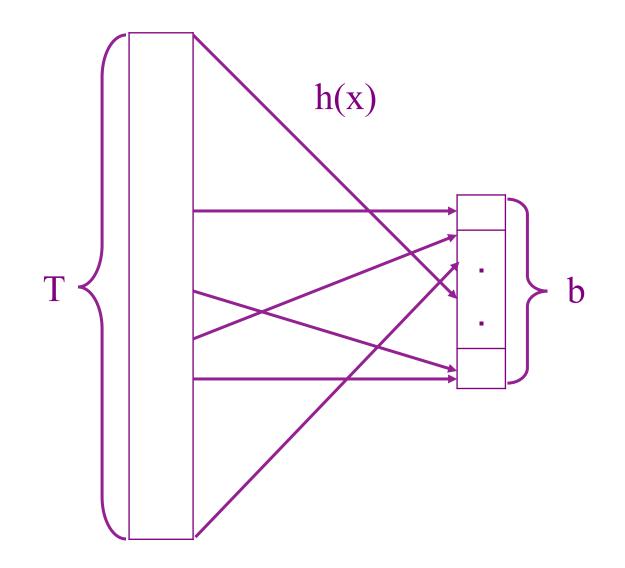
- A bucket array indexed by the phone number has optimal O(1) query time
- There is a huge amount of wasted space

Bucket Array

- A data structure
- The location of an item is determined by:
 - directly as a function of the item itself: f(key)=key
 - Not by a sequence of trial and error comparisons
- Commonly used to provide faster searching
 - O(n) for linear searches
 - O (logn) for binary search
 - -O(1) for hash table

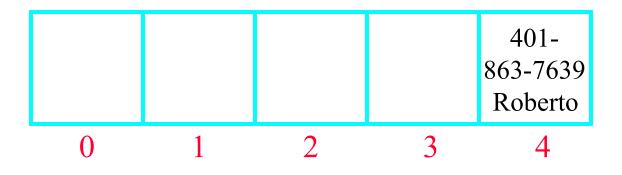
Space Solution

- *A Hash Table* is an alternative solution with O(1) expected query time and O(n + N) space, where N is the size of the table
- Like an array, but with a function to map the large range of keys into a smaller one
 - e.g., take the original key, *mod* the size of the table, and use that as an index



Example

- Insert item (401-863-7639, Roberto) into a table of size 5
- 4018637639 mod 5 = 4, so item (401-863-7639, Roberto) is stored in slot 4 of the table
- A lookup uses the same process: map the key to an index, then check the array cell at that index



Static hashing

- dictionary pairs are stored in a table, ht, called hash table
- **ht** is partitioned into **b** buckets: ht[0:b-1]
- **ht** is maintained in sequential memory
- each bucket holds \$ slots, each slot holds one pair, usually, s = 1
- the address of a pair with key k is determined by a hash function h, h(k) is the hash or home address of k, h(k)∈{0, 1,...,b-1}

Notations

- **T** --- the total number of possible keys.
- **n** --- the number of pairs in the hash table.

Definition:

The key density of a hash table is the ratio n/T.
The loading density (or factor) of a hash table is α=n/(s×b).

Usually, n<<T, and b<<T.

Notations

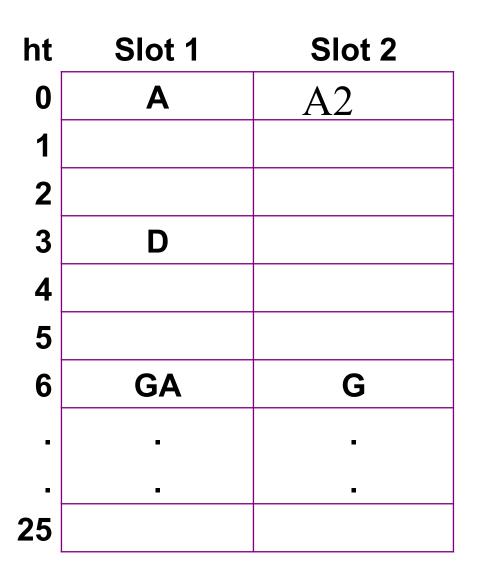
- 2 keys k₁ and k₂ are said to be synonyms with respect to h if h(k1) = h(k2).
- a **collision** occurs when the home bucket for the new pair is not empty.
- an **overflow** occurs when a new pair is hashed into a full bucket.
- when s=1, collisions and overflows occur simultaneously.

An Example

- b=26, s=2, n=10, hence $\alpha = ?$ -10/52 = 0.19
- Keys: GA, D, A, G, L, A2, A1, A3, A4, E
- h(k) = the first character of k

– A to Z corresponds to 0 to 25

- GA, D, A, G, L entered
- A2 entered
 - Collision
- A1 entered
 - Collision
 - Overflow



Analysis

- No overflow
- Performance of insert, delete, search
 - Hash function
 - Searching within a bucket
 - Independent of **n**
- However, Overflow is happening

-T >> b

From Keys to Indices

- The mapping of keys to indices of a hash table is called a hash function
- A hash function is usually the composition of two maps:
 - hash code map: key \$\laph\$ integer

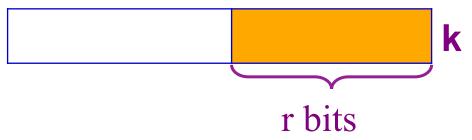
- **compression map**: integer & [0, N - 1]

Hash function

- Essential requirement of the hash function map equal keys to equal indices
- A "good" hash function
 - minimizes the **probability of collisions**
 - -Easy to compute
- uniform hash function
 - If k is a key chosen at random from the key space, then the probability that h(k)=i to be 1/b for all buckets i

Hash function **compression map**

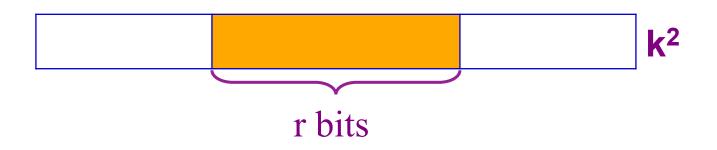
- Division
 - $-h(\mathbf{k}) = |\mathbf{k}| \mod N$
 - Selection of N is critical
 - $-N=2^{r}$ is bad because not all the bits are taken into account



the table size N is usually chosen as a prime number

Hash function hash code map

- Mid-square
 - h(k) is computed by using an appropriate number of bits from the middle of k² to obtain the bucket address.
- If r bits used, $b=2^{r}$



Hash function hash code map

- Folding
 - k is partitioned into several parts, all but the last being of the same length
 - These partitions are then added together to obtain the hash address for k.

k=12320324111220 is partitioned into parts that are 3 decimal digits long.

 $P_1=123, P_2=203, P_3=241, P_4=112, P_5=20.$

• shift folding

h(k)=123+203+241+112+20=699

• folding at the boundaries

h(k)=123+302+241+211+20=897

Hash function hash code map

- Digit Analysis
 - each k is interpreted as a number using some radix r
 - the digits of each k are examined
 - digits having the most skewed distribution are deleted
 - until the number of digits left is small enough to give an address

Hash function hash code map

- Converting Keys to integers
 - for strings of a natural language, combine the character values (ASCII or Unicode) $a_0 a_1 \dots a_{n-1}$ by viewing them as the coefficients of a polynomial: $a_0 + a_1 x$ $+ \dots + x_{n-1} a_{n-1}$

Overflow handling

- A key is mapped to an already occupied table location
 - what to do?!?
- Use a collision handling technique
 - Open Addressing
 - Linear Probing
 - Quadratic probing
 - Double Hashing

- Chaining

Linear Probing

- $h_i(K) = (hash(K) + i) \mod m$
- Insertion:
 - Let K be the new key to be inserted, compute hash(K)
 - For i = 0 to m-1
 - compute $L = (hash(K) + I) \mod m$
 - T[L] is empty, then we put K there and stop.
 - If we cannot find an empty entry to put K, it means that the table is full and we should report an error.

Quadratic Probing

- $h_i(K) = (hash(K) + i^2) \mod m$
- Insertion:
 - Let K be the new key to be inserted, compute hash(K)
 - For i = 0 to m-1
 - compute $L = (hash(K) + i^2) mod m$
 - T[L] is empty, then we put K there and stop.
 - If we cannot find an empty entry to put K, it means that the table is full and we should report an error.

Double Hashing

• Hash1(), Hash2(),,HashN()

An Open Hash Table

Hash (key) produces an index in the range 0 to 6. That index is the "home address"

> Some insertions: K1 --> 3 K2 --> 5 K3 --> 2



key value

Handling Collisions

Some more insertions: K4 --> 3 K5 --> 2 K6 --> 4

Linear probing collision resolution strategy



Search Performance



Average number of probes needed to retrieve the value with key K?

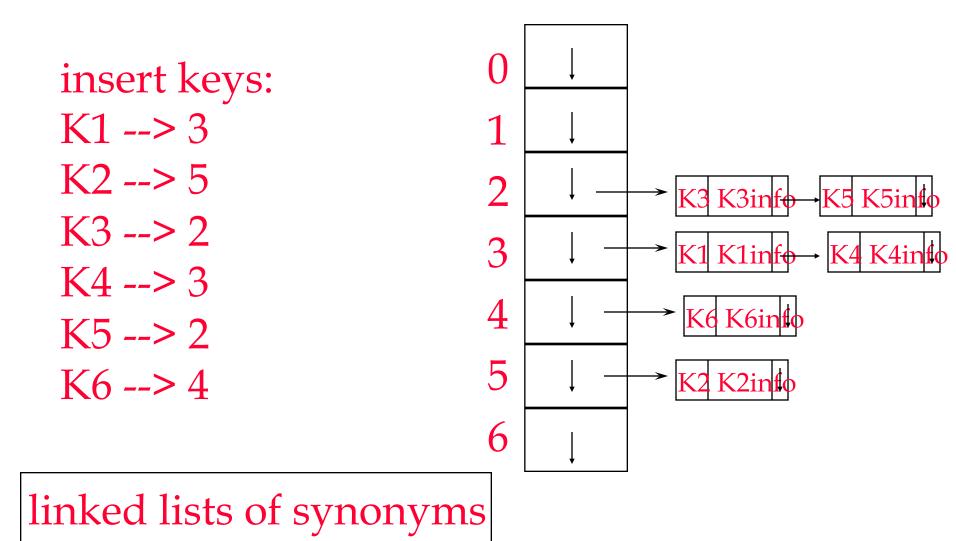
K	hash(K)	#probes
K1	3	1
K2	5	1
K3	2	1
K4	3	2
K5	2	5
K6	4	4

14/6 = 2.33 (successful) unsuccessful search?

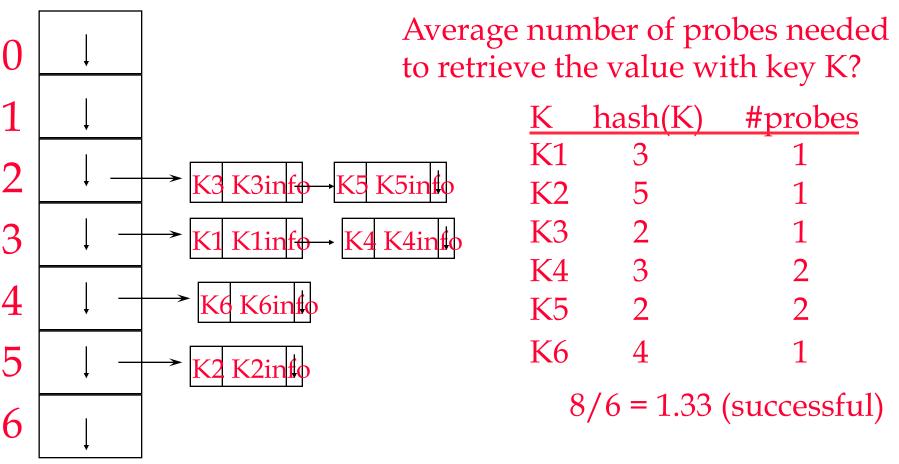
Chaining

- Linear probing performs poorly
 - the search for a key involves comparison with keys having different hash values
 - making a local collision a global one

A Chained Hash Table



Search Performance



unsuccessful search?

successful search performance

	open addressing (linear probing)	open addressing (double hashing)	chaining
load factor			
0.5	1.50	1.39	1.25
0.7	2.17	1.72	1.35
0.9	5.50	2.56	1.45
1.0			1.50
2.0			2.00

Exercises: P475-3, 6