

Data Structures and Algorithms

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Part 7

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Data Structures and Algorithms

Week 7

1. Dictionaries
2. Hashing
3. Hash Functions
4. Collisions
5. Performance Analysis

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Dictionary

- *Dictionary* – a dynamic data structure with methods:
 - **Search(S, k)** – *an access operation that returns a pointer x to an element where $x.key = k$*
 - **Insert(S, x)** – *a manipulation operation that adds the element pointed to by x to S*
 - **Delete(S, x)** – *a manipulation operation that removes the element pointed to by x from S*
- An element has a *key* part and a *satellite data* part.

Dictionaries

- Dictionaries store elements so that they can be located quickly using **keys**.
- A dictionary may hold bank accounts.
 - Each account is an object that is identified by an account number.
 - Each account stores a lot of additional information.
 - An application wishing to operate on an account would have to provide the account number as a search **key**.

Dictionaries/2

- If order (methods such as *min*, *max*, *successor*, *predecessor*) is not required it is enough to check for **equality**.
- Operations that require ordering are still possible but cannot use the dictionary access structure.
 - Usually all elements must be compared, which is slow.
 - Can be OK if it is rare enough

Dictionaries/3

- Different data structures to realize dictionaries
 - arrays
 - linked lists
 - **Hash tables**
 - Binary trees
 - Red/Black trees
 - B-trees
- In Java:
 - `java.util.Map` – interface defining Dictionary ADT

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The Problem

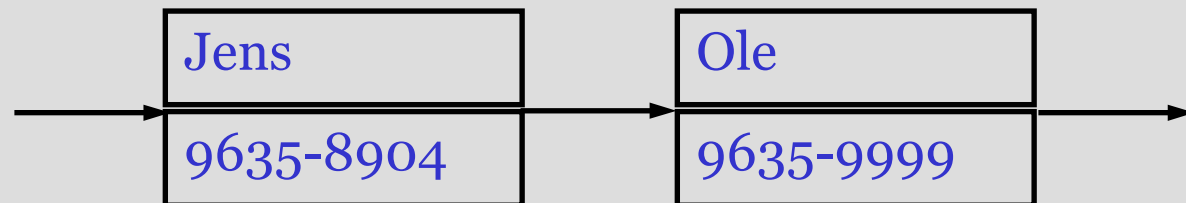
- XY Telecom, a large phone company, wants to provide a caller ID capability:
 - given a phone number, return the caller's name
 - phone numbers range from 0 to $r = 10^8 - 1$
 - want to do this as efficiently as possible

The Problem/2

- Two suboptimal ways to design this dictionary
 - direct addressing: an array indexed by key:
 - Requires $O(1)$ time,
 - Requires $O(r)$ space - huge amount of wasted space

(null)	(null)	Jens	(null)	(null)
0000- 0000	0000- 0001	9635- 8904	9635- 8905	9999- 9999

- a linked list: requires $O(n)$ time, $O(n)$ space



Another Solution: Hashing

- We can do better, with a **Hash table** of size m .
- Like an array, but with a **function** to map the large range into one which we can manage.
 - e.g., take the original key, modulo the (relatively small) size of the table, and use that as an index
- Insert (9635-8904, Jens) into a hash table with, say, five slots ($m = 5$)

- $96358904 \bmod 5 = 4$

(null)	(null)	(null)	(null)	Jens
0	1	2	3	4

- $O(1)$ expected time, $O(n+m)$ space

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Hash Functions

- Need to choose a good hash function (HF)
 - quick to compute
 - distributes keys uniformly throughout the table
- How to deal with hashing non-integer keys:
 - find some way of turning the keys into integers
 - in our example, remove the hyphen in 9635-8904 to get 96358904
 - for a string, add up the ASCII values of the characters of your string (e.g., `java.lang.String.hashCode()`)
 - then use a standard hash function on the integers

HF: Division Method

- Use the remainder: $h(k) = k \bmod m$
 - k is the key, m the size of the table
- Need to choose m
- $m = b^e$ (**bad**)
 - if m is a power of 2, $h(k)$ gives the e least significant bits of k
 - all keys with the same ending go to the same place
- m prime (**good**)
 - helps ensure uniform distribution
 - primes not too close to exact powers of 2 are best

HF: Division Method/2

- Example 1
 - hash table for $n = 2000$ character strings, ok to investigate an average of three attempts/search
 - $m = 701$
 - a prime near $2000/3$
 - but not near any power of 2
- Further examples
 - $m = 13$
 - $h(3) = 3$
 - $h(12) = 12$
 - $h(13) = 0$

HF: Multiplication Method

- Use $h(k) = \lfloor m (k A \bmod 1) \rfloor$
 - k is the key
 - m the size of the table
 - A is a constant $0 < A < 1$
 - $(k A \bmod 1)$: the fractional part of $k A$
- The steps involved
 - map $0 \dots k_{max}$ into $0 \dots k_{max} A$
 - take the fractional part (mod 1)
 - map it into $0 \dots m-1$

HF: Multiplication Method/2

- Choice of m and A
 - Value of m is not critical:
typically, for some p use $m = 2^p$
 - Optimal choice of A depends
on the characteristics of the data
 - Knuth says use $A = \frac{\sqrt{5}-1}{2} = 0.618033988$

HF: Multiplication Method/3

- Assume 7-bit binary keys, $0 \leq k < 128$
- $m = 64 = 2^6$, $p = 6$
- $A = 89/128 = .1011001$, $k = 107 = 1101011$
- Computation of $h(k)$:

$$\begin{array}{r} .1011001 \text{ A} \\ 1101011 \text{ k} \\ \hline 1001010.0110011 \text{ kA} \\ .0110011 \text{ kA mod 1} \\ \hline 011001.1 \text{ m(kA mod 1)} \end{array}$$

- Thus, $h(k) = 25$

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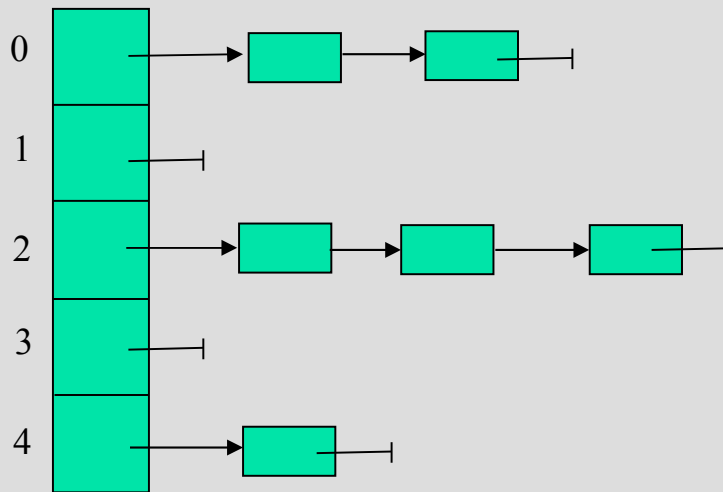
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Collisions

- Assume a key is mapped to an already occupied table location
 - what to do?
- Use a **collision handling** technique
- 3 techniques to deal with collisions:
 - chaining
 - open addressing/linear probing
 - open addressing/double hashing

Chaining

- **Chaining** maintains a table of links, indexed by the keys, to **lists** of items with the same key



Open Addressing

- All elements are stored in the hash table (can fill up), i.e., $n \leq m$
- Each table entry contains either an element or null
- When searching for an element, systematically probe table slots
- Modify hash function to take probe number i as second parameter

$$h: U \times \{ 0, 1, \dots, m-1 \} \rightarrow \{ 0, 1, \dots, m-1 \}$$

Open Addressing/2

- Hash function, h , determines the sequence of slots examined for a given key
- Probe sequence for a given key k given by

($h(k,0), h(k,1), \dots, h(k,m-1)$)

a permutation of ($0, 1, \dots, m-1$)

Linear Probing

LinearProbingInsert(k)

```
01 if (table is full) error
02 probe = h(k)
03 while (table[probe] occupied)
04     probe = (probe+1) mod m
05 table[probe] = k
```

- If the current location is used, try the next table location:
 $h(key,i) = (h_1(key)+i) \text{ mod } m$
- Lookups walk along the table
until the key or an empty slot is found
- Uses less memory than chaining
 - one does not have to store all those links
- Slower than chaining
 - one might have to probe the table for a long time

Linear Probing/2

- Problem “**primary clustering**”:
long lines of occupied slots
 - A slot preceded by i full slots has a high probability of getting filled: $(i+1)/m$
- Alternatives: (quadratic probing,) **double hashing**
- Example:
 - $h(k) = k \bmod 13$
 - insert keys: 18 41 22 44 59 32 31 73

Double Hashing

- Use two hash functions:

$$h(\text{key}, i) = (h_1(\text{key}) + i * h_2(\text{key})) \bmod m, i=0,1,\dots$$

```
DoubleHashingInsert(k)
```

```
01 if (table is full) error
```

```
02 probe = h1(k)
```

```
03 offset = h2(k)
```

```
03 while (table[probe] occupied)
```

```
04     probe = (probe + offset) mod m
```

```
05 table[probe] = k
```

- Distributes keys much more uniformly than linear probing.

Double Hashing/2

- $h_2(k)$ must be relative prime to m to search the entire hash table
 - Suppose $h_2(k) = k * a$ and $m = w * a$, $a > 1$
- Two ways to ensure this:
 - m is power of 2, $h_2(k)$ is odd
 - m : prime, $h_2(k)$: positive integer $< m$
- Example
 - $h_1(k) = k \bmod 13$, $h_2(k) = 8 - (k \bmod 8)$
 - insert keys: 18 41 22 44 59 32 31 73

Open addressing: delete

- Complex to delete from
 - A slot may be reached from different points
 - We cannot simply store “NIL”: we'd lose the information necessary to retrieve other keys
 - Possible solution: mark the deleted slot as “deleted”, insert also on “deleted”
 - Drawback: retrieval time no more depending on load factor: potentially lots of “jumps” on “deleted” slots
- When deletion admitted/frequent,
chaining preferred

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Analysis of Hashing:

- An element with key k is stored in slot $h(k)$ (instead of slot k without hashing)
- The hash function h maps the universe U of keys into the slots of hash table $T[0\dots m-1]$
 $h: U \rightarrow \{0, 1, \dots, m-1\}$
- Assumption: Each key is equally likely to be hashed into any slot (bucket):
simple uniform hashing
- Given hash table T with m slots holding n elements, the **load factor** is defined as $\alpha = n/m$

Analysis of Hashing/2

- Assume time to compute $h(k)$ is $\Theta(1)$
- To find an element
 - using h , look up its position in table T
 - search for the element in the linked list of the hashed slot
 - *uniform* hashing yields an average list length $\alpha = n/m$
 - expected number of elements to be examined α
 - search time $O(1+\alpha)$

Analysis of Hashing/3

- Assuming the number of hash table slots is proportional to the number of elements in the table
$$n = O(m)$$
$$\alpha = n/m = O(m)/m = O(1)$$
 - searching takes constant time on average
 - insertion takes $O(1)$ worst-case time
 - deletion takes $O(1)$ worst-case time (pass the element not key, lists are doubly-linked)

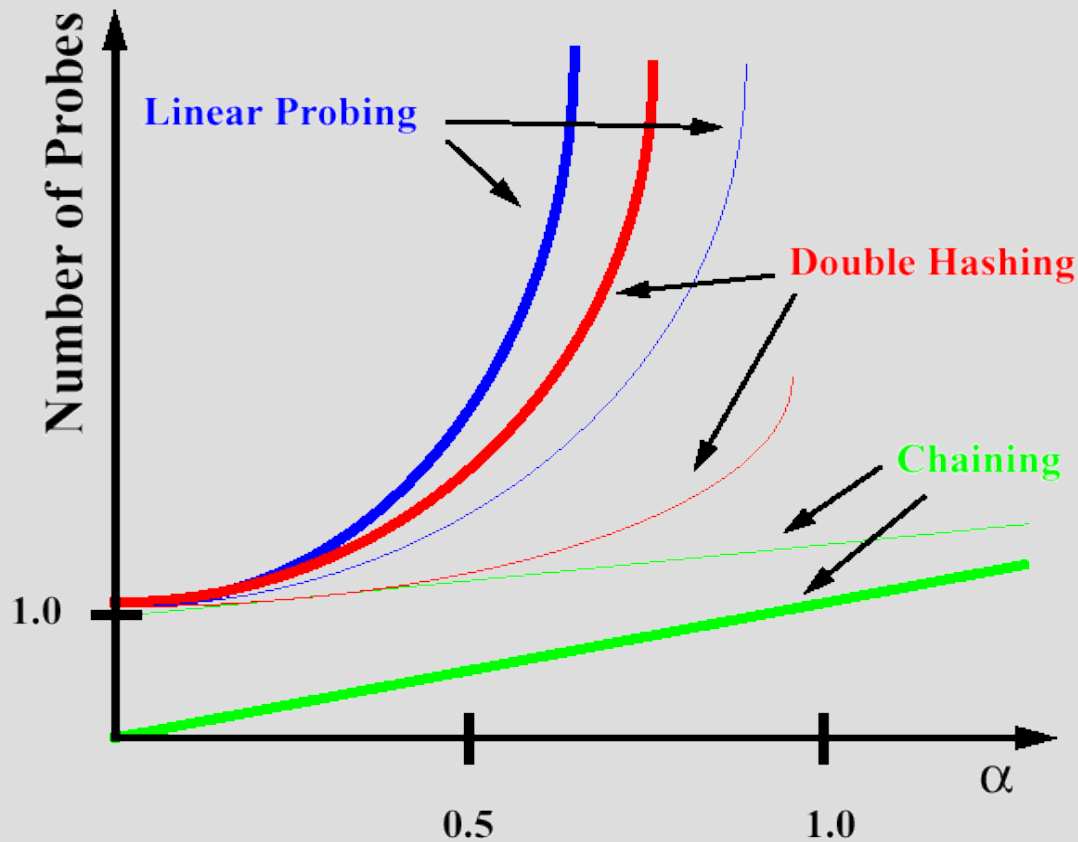
Expected Number of Probes

- Load factor $\alpha < 1$ for probing
- Analysis of probing uses *uniform hashing* assumption – any permutation is equally likely

	Unsuccessful	Successful
Chaining	$O(1 + \alpha)$	$O(1 + \alpha)$
Probing	$O\left(\frac{1}{1 - \alpha}\right)$	$O\left(\frac{1}{\alpha} \ln \frac{1}{1 - \alpha}\right)$

- Chaining: 1 ($\alpha=0\%$), 1.5 ($\alpha=50\%$), 2 ($\alpha=100\%$), n ($\alpha=n$)
- Probing, unsucc: 1.25 ($\alpha=20\%$), 2 ($\alpha=50\%$), 5 ($\alpha=80\%$), 10 ($\alpha=90\%$)
- Probing, succ: 0.28 ($\alpha=20\%$), 1.39 ($\alpha=50\%$), 2.01 ($\alpha=80\%$), 2.56 ($\alpha=90\%$)

Expected Number of Probes/2



— Unsuccessful
- - Successful

Slides by M. Böhlen and R. Sebastiani

Summary

- Hashing is very efficient
(not obvious, probability theory).
- Its functionality is limited (printing elements sorted according to key is not supported).
- The size of the hash table
may not be easy to determine.
- A hash table is not really
a dynamic data structure.

Suggested exercises

- Implement a Hash Table with the different techniques
- With paper & pencil, draw the evolution of a hash table when inserting, deleting and searching for new element, with the different techniques
- See also exercises of CLRS

Next Part

- Graphs:
 - Representation in memory
 - Breadth-first search
 - Depth-first search
 - Topological sort