

Hash Tables

CSE 373
Data Structures & Algorithms
Ruth Anderson
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5/14/2008

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Today's Outline

- **Admin:**
 - HW #5 – due Monday May 19
 - Wed at beginning of class = latest accepted
 - Midterm #2 – Friday May 23
 - Feedback Survey
- **Hash Tables** (Weiss Chapter 5)

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Dictionary Implementations

	Unsorted linked list	Sorted Array	Binary Search Tree	AVL Tree
Insert				$O(\log N)$
Find	$O(N)$			
Delete			$O(N)$	$O(\log N)$

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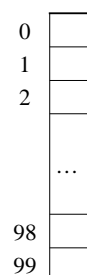
Constant Time Access

Data Set:

- 100 students
- Keys = Student numbers between 0 and 99.

Solution:

- Array of size 0-99.
- One-to-one mapping: e.g. student number 2 goes in location 2



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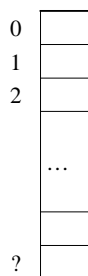
Constant Time Access?

Data Set:

- 100 students
- Keys = Student numbers between 0 and 999999999.

Solution:

- Array of size ?
- Mapping ?

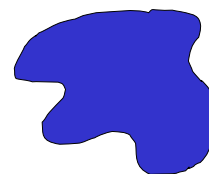


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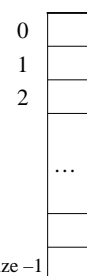
Hash Tables

- A **hash table** is an array of some fixed size.
- General idea:



Key Space (e.g., integers, strings)

Hash Table



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Example

- Key space = integers
- TableSize = 10
- $h(K) = K \bmod 10$
- **Insert:** 207, 18, 41, 194, 19, 43

0	
1	
2	
3	
4	
5	
6	
7	
8	
9	

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Another Example

- key space = integers
- TableSize = 6
- $h(K) = K \bmod 6$
- **Insert:** 7, 18, 41, 34

0	
1	
2	
3	
4	
5	

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Student Activity

Hash Functions

1. **simple/fast** to compute,
2. Avoid **collisions**
3. have keys distributed **evenly** among cells.

Perfect Hash function:

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Sample Hash Functions:

key space = strings A=0, B=1,...Z=25

$s = s_0 s_1 s_2 \dots s_{k-1}$

1. $h(s) = s_0 \bmod \text{TableSize}$
2. $h(s) = \left(\sum_{i=0}^{k-1} s_i \right) \bmod \text{TableSize}$
3. $h(s) = \left(\sum_{i=0}^{k-1} s_i \cdot 26^i \right) \bmod \text{TableSize}$

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Designing a Hash Function for web URLs

$s = s_0 s_1 s_2 \dots s_{k-1}$

Issues to take into account:

$h(s) =$

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Collision Resolution

Collision: when two keys map to the same location in the hash table.

Two ways to resolve collisions:

1. Separate Chaining
2. Open Addressing (linear probing, quadratic probing, double hashing)

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Separate Chaining

$h(K) = K \bmod 10$

0	
1	
2	
3	
4	
5	
6	
7	
8	
9	

Insert:
 10
 22
 107
 12
 42

Separate chaining:
 All keys that map to the same hash value are kept in a list (or "bucket").

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Analysis of Find

The **load factor**, λ , of a hash table is the ratio:

$$\frac{N}{\text{TableSize}} \leftarrow \begin{array}{l} \# \text{ of elements} \\ \text{table size} \end{array}$$

For separate chaining,
 $\lambda = \text{average \# of elements in a bucket}$

Average # of values needed to examine for a:

- unsuccessful find:
- successful find:

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How Big Should the Hash Table Be?

For Separate Chaining, if we want $\lambda = 1$
 (e.g. the average # of values per bucket = 1)

- How large should I make the hash table, in terms of N?

TableSize =

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tableSize: Why Prime?

- Suppose
 - data stored in hash table: 7160, 493, 60, 55, 321, 900, 810
 - tableSize = 10
 data hashes to 0, 3, 0, 5, 1, 0, 0
 - tableSize = 11
 data hashes to 10, 9, 5, 0, 2, 9, 7

Real-life data tends to have a pattern

Being a multiple of 11 is usually *not* the pattern ☺

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Open Addressing

$h(K) = K \bmod 10$

0	
1	
2	
3	
4	
5	
6	
7	
8	
9	

Insert:
 38
 19
 8
 109
 10

Linear Probing: after checking $h(k)$, try $h(k)+1$, if that is full, try $h(k)+2$, then try $h(k)+3$, etc.

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Terminology Alert!

“**Open Hashing**” “**Closed Hashing**”
 equals equals
 Weiss “**Separate Chaining**” “**Open Addressing**”

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Linear Probing

$$f(i) = i$$

- Probe sequence:

$$0^{\text{th}} \text{ probe} = h(k) \bmod \text{TableSize}$$

$$1^{\text{th}} \text{ probe} = (h(k) + 1) \bmod \text{TableSize}$$

$$2^{\text{th}} \text{ probe} = (h(k) + 2) \bmod \text{TableSize}$$

...

$$i^{\text{th}} \text{ probe} = (h(k) + i) \bmod \text{TableSize}$$

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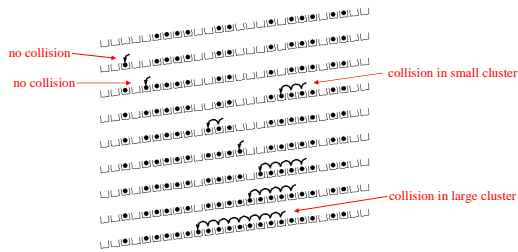
Write pseudocode for find(k) for Open Addressing with linear probing

- Find(k) returns i where $T(i) = k$

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Linear Probing – Clustering



[R. Sedgewick]

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Load Factor in Linear Probing

- For any $\lambda < 1$, linear probing will find an empty slot
- Expected # of probes (for large table sizes)

- successful search: $\frac{1}{2} \left(1 + \frac{1}{(1-\lambda)} \right)$

- unsuccessful search: $\frac{1}{2} \left(1 + \frac{1}{(1-\lambda)^2} \right)$

- Linear probing suffers from **primary clustering**
- Performance quickly degrades for $\lambda > 1/2$

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Quadratic Probing

$$f(i) = i^2$$

Less likely to encounter Primary Clustering

- Probe sequence:

$$0^{\text{th}} \text{ probe} = h(k) \bmod \text{TableSize}$$

$$1^{\text{th}} \text{ probe} = (h(k) + 1) \bmod \text{TableSize}$$

$$2^{\text{th}} \text{ probe} = (h(k) + 4) \bmod \text{TableSize}$$

$$3^{\text{th}} \text{ probe} = (h(k) + 9) \bmod \text{TableSize}$$

...

$$i^{\text{th}} \text{ probe} = (h(k) + i^2) \bmod \text{TableSize}$$

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Quadratic Probing

0		Insert:
1		89
2		18
3		49
4		58
5		79
6		
7		
8		
9		

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Quadratic Probing Example

insert(76) $76\%7 = 6$ insert(40) $40\%7 = 5$ insert(48) $48\%7 = 6$ insert(5) $5\%7 = 5$ insert(55) $55\%7 = 6$
 But... insert(47) $47\%7 = 5$

0	
1	
2	
3	
4	
5	
6	76

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Quadratic Probing:

Success guarantee for $\lambda < 1/2$

- If size is prime and $\lambda < 1/2$, then quadratic probing will find an empty slot in size/2 probes or fewer.
 - show for all $0 \leq i, j \leq \text{size}/2$ and $i \neq j$

$$(h(x) + i^2) \bmod \text{size} \neq (h(x) + j^2) \bmod \text{size}$$
 - by contradiction: suppose that for some $i \neq j$:

$$(h(x) + i^2) \bmod \text{size} = (h(x) + j^2) \bmod \text{size}$$

$$\Rightarrow i^2 \bmod \text{size} = j^2 \bmod \text{size}$$

$$\Rightarrow (i^2 - j^2) \bmod \text{size} = 0$$

$$\Rightarrow [(i + j)(i - j)] \bmod \text{size} = 0$$
 BUT size does not divide $(i-j)$ or $(i+j)$

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Quadratic Probing: Properties

- For any $\lambda < 1/2$, quadratic probing will find an empty slot; for bigger λ , quadratic probing may find a slot
- Quadratic probing does not suffer from *primary* clustering: keys hashing to the same *area* are not bad
- But what about keys that hash to the same *spot*?
 – **Secondary Clustering!**

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Double Hashing

$$f(i) = i * g(k)$$

where g is a second hash function

- Probe sequence:
 - 0th probe = $h(k) \bmod \text{TableSize}$
 - 1th probe = $(h(k) + g(k)) \bmod \text{TableSize}$
 - 2th probe = $(h(k) + 2 * g(k)) \bmod \text{TableSize}$
 - 3th probe = $(h(k) + 3 * g(k)) \bmod \text{TableSize}$
 - ...
 - j^{th} probe = $(h(k) + i * g(k)) \bmod \text{TableSize}$

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Double Hashing Example

$h(k) = k \bmod 7$ and $g(k) = 5 - (k \bmod 5)$

	76	93	40	47	10	55
0						
1				47	47	47
2		93	93	93	93	93
3					10	10
4						55
5			40	40	40	40
6	76	76	76	76	76	76
Probes	1	1	1	2	1	2

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Resolving Collisions with Double Hashing

0	
1	
2	
3	
4	
5	
6	
7	
8	
9	

Hash Functions:

 $H(K) = K \bmod M$
 $H_2(K) = 1 + ((K/M) \bmod (M-1))$
 $M =$

Insert these values into the hash table in this order. Resolve any collisions with double hashing:

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28
33
147
43

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Rehashing

Idea: When the table gets too full, create a bigger table (usually 2x as large) and hash all the items from the original table into the new table.

- When to rehash?
 - half full ($\lambda = 0.5$)
 - when an insertion fails
 - some other threshold
- Cost of rehashing?

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Hashing Summary

- Hashing is one of the most important data structures.
- Hashing has many applications where operations are limited to find, insert, and delete.
- Dynamic hash tables have good amortized complexity.

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