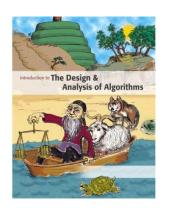




Introduction to

Algorithm Design and Analysis

[9] Hashing



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In the Last Class...

- The searching problem
 - o "Architecture" of data
- logn search
 - o Binary search
 - In a more general sense
 - o Red-black tree: balanced BST
 - Definition
 - o Black height constraint for balance
 - o Color constraint for low maintenance cost
 - Operation
 - o Insertion, deletion



Hashing

- The searching problem
 - o The ambition of hashing
- Hashing
 - o Brute force table: direct addressing
 - o Basic idea of hashing
- Collision Handling for Hashing
 - o Closed address hashing
 - o Open address hashing
- Amortized Analysis
 - o Array doubling



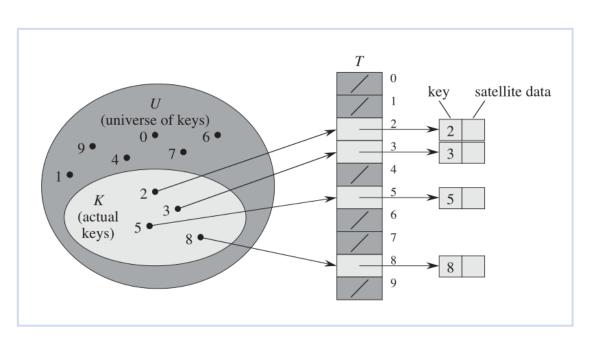
Cost for Searching

- Brute force
 - \circ O(n)
- Balanced BST
 - o O(logn)
- Hashing almost constant time
 - \circ O(1+ α)
- "Mission impossible"
 - $\circ O(1)$



Searching - a Brute Force Approach

- Direct-address table
 - o Take into account the *whole universe* of keys



Direct-address Table

DIRECT-ADDRESS-SEARCH (T, k) return T[k]

DIRECT-ADDRESS-INSERT (T, x)

T[key[x]] := x

DIRECT-ADDRESS-DELETE (T, x)

T[key[x]] := NIL

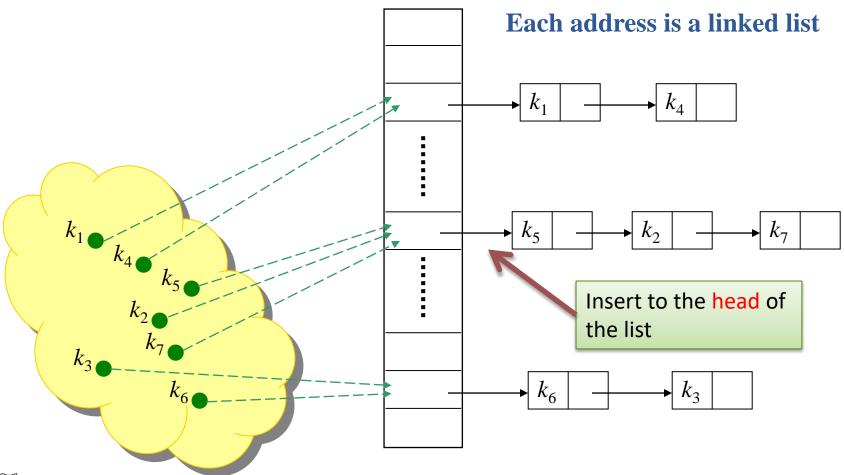


Hashing: the Idea

quite large, but only a **Hash Table** (in feasible size) small part is used in an application E[0]• Index distribution E[1] Collision handling χ Hash **Key Space Function** E[k]H(x)=kValue of a A calculated specific key array index for the key E[m-1]



Collision Handling: Closed Address





Closed Address - Analysis

- Assumption simple uniform hashing
 - o For j=0,1,2,...,m-1, the average length of the list at E[j] is n/m.
- The average cost for an unsuccessful search
 - o Any key that is not in the table is equally likely to hash to any of the *m* address.
 - o Total cost $\Theta(1+n/m)$
 - The average cost to determine that the key is not in the list E[h(k)] is the cost to search to the end of the list, which is n/m.



Closed Address - Analysis

- For successful search (assuming that x_i is the ith element inserted into the table, i=1,2,...,n)
 - o For each *i*, the probability of that x_i is searched is 1/n.
 - o For a specific x_i , the number of elements examined in a successful search is t+1, where t is the number of elements inserted into the same list as x_i , after x_i has been inserted

$$1 + \frac{1}{n} \sum_{i=1}^{n} (1+t)$$

- How to compute t?
 - o Consider the *construction* process of the hash table



Closed Address - Analysis

- For successful search: (assuming that x_i is the ith element inserted into the table, i=1,2,...,n)
 - o For each *i*, the probability of that x_i is searched is 1/n.
 - \circ For a specific x_i , the number of elements examined in a successful search is t+1, where t is the number of elements inserted into the same list as x_i after x_i has been inserted. And for any j, the probability of that x_i is inserted into the same list of x_i is 1/m. So, the cost is:

$$\frac{1}{n} \sum_{i=1}^{n} \left(1 + \sum_{j=i+1}^{n} \frac{1}{m} \right)$$
 Expected number of elements in front of the searched one in the same linked list.

Expected number of



Closed Address: Analysis

- The average cost of a successful search:
 - o Define α =n/m as *load factor*,

The average cost of a successful search is:

$$\frac{1}{n} \sum_{i=1}^{n} \left(1 + \sum_{i=i+1}^{n} \frac{1}{m} \right) = 1 + \frac{1}{nm} \sum_{i=1}^{n} (n-i) = 1 + \frac{1}{nm} \sum_{i=1}^{n-1} i$$

$$= 1 + \frac{n-1}{2m} = 1 + \frac{\alpha}{2} - \frac{\alpha}{2n} = \Theta(1+\alpha)$$

Number of elements in front of the searched one in the same linked list

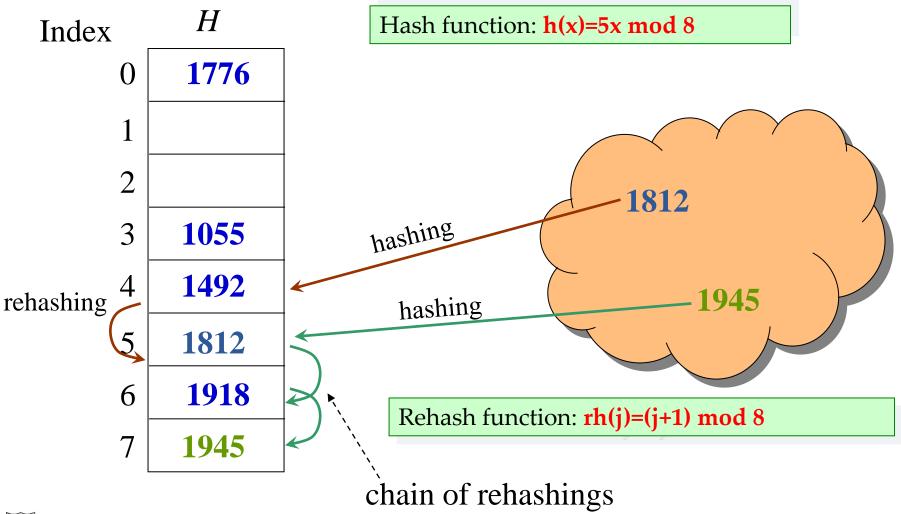


Collision Handling: Open Address

- All elements are stored in the hash table
 - o No linked list is used
 - o The load factor α cannot be larger than 1
- Collision is settled by "rehashing"
 - o A function is used to get a new hashing address for each collided address
 - The hash table slots are *probed* successively, until a valid location is found.
- The probe sequence can be seen as a permutation of (0,1,2,..., *m*-1)



Linear Probing: An Example





Commonly Used Probing

Linear probing:

Given an ordinary hash function h', which is called an auxiliary hash function, the hash function is: (clustering may occur)

$$h(k,i) = (h'(k)+i) \mod m \quad (i=0,1,...,m-1)$$

Quadratic Probing:

Given auxiliary function h' and nonzero auxiliary constant c_1 and c_2 , the hash function is: (secondary clustering may occur)

$$h(k,i) = (h'(k)+c_1i+c_2i^2) \mod m \quad (i=0,1,...,m-1)$$

Double hashing:

Given auxiliary functions h_1 and h_2 , the hash function is:

$$h(k,i) = (h_1(k) + ih_2(k)) \mod m \quad (i=0,1,...,m-1)$$



Equally Likely Permutations

Assumption

o Each key is equally likely to have any of the *m*! permutations of (1,2...,*m*) as its probe sequence

Note

Both linear and quadratic probing have only *m* distinct probe sequence, as determined by the first probe



Analysis for Open Address Hashing

- The average number of probes in an unsuccessful search is at most $1/(1-\alpha)$ $(\alpha = n/m < 1)$
 - o Assuming uniform hashing

The probability of the first probed position being occupied is $\frac{n}{m}$, and that of the $j^{th}(j > 1)$ position occupied is $\frac{n-j+1}{m-j+1}$. So the probability of the number of probes no less than i will be:

$$\frac{n}{m} \cdot \frac{n-1}{m-1} \cdot \frac{n-2}{m-2} \cdot \dots \cdot \frac{n-i+2}{m-i+2} \le (\frac{n}{m})^{i-1} = \alpha^{i-1}$$

The the average number of probe is: $\sum_{i=1}^{\infty} \alpha^{i-1} = \sum_{i=0}^{\infty} \alpha^i = \frac{1}{1-\alpha}$

Lectures on Algorithm Design and Analysis (LADA), Spring 2020

See [CLRS] p.1199, C.25

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Analysis for Open Address Hashing

 The average cost of probes in an successful search is at most $\frac{1}{\alpha} \ln \frac{1}{1-\alpha}$ ($\alpha = n/m < 1$)

o Assuming uniform hashing

To search for the $(i+1)^{th}$ inserted element in the table, the cost is the same as that for inserting it when there are just i elements in the table.

At that time, $\alpha = \frac{i}{m}$. So the cost is $\frac{1}{1 - \frac{i}{m}} = \frac{m}{m - i}$.

So the average cost for a successful search is:

$$\frac{1}{n} \sum_{i=0}^{n-1} \frac{m}{m-i} = \frac{m}{n} \sum_{i=0}^{n-1} \frac{1}{m-i} = \frac{1}{\alpha} \sum_{i=m-n+1}^{m} \frac{1}{i}$$

$$\leq \frac{1}{\alpha} \int_{-\infty}^{m} \frac{dx}{x} = \frac{1}{\alpha} \ln \frac{m}{m-n} = \frac{1}{\alpha} \ln$$

90% full: 2.559 $\leq \frac{1}{\alpha} \int_{m-n}^{m} \frac{dx}{x} = \frac{1}{\alpha} \ln \frac{m}{m-n} = \frac{1}{\alpha} \ln \frac{1}{1-\alpha}$

For your reference:

Half full: 1.387;

Hash Function

- A good hash function satisfies the assumption of simple uniform hashing
 - Heuristic hashing functions
 - The division method: $h(k)=k \mod m$
 - The multiplication method: $h(k) = \lfloor m(kA \mod 1) \rfloor$ (0<*A*<1)
 - o No single function can avoid the worst case $\Theta(n)$.
 - So "universal hashing" is proposed.
 - o Rich resource about hashing function
 - Gonnet and Baeza-Yates: Handbook of Algorithms and Data Structures, Addison-Wesley, 1991.



Array Doubling

- Cost for search in a hash table is $\Theta(1+\alpha)$
 - o If we can keep α constant, the cost will be $\Theta(1)$
- What if the hash table is more and more loaded?
 - Space allocation techniques such as array doubling may be needed
- The problem of "unusually expensive" individual operation



Looking at the Memory Allocation

- hashingInsert(HASHTABLE *H*, ITEM *x*)
- integer *size*=0, *num*=0;
- if *size*=0 then allocate a block of size 1; *size*=1;
- if *num=size* then
- allocate a block of size 2size;
- move all item into new table;
- size=2size;
- insert *x* into the table;
- *num=num+1*;

→ Elementary insertion: cost 1

Insertion with

expansion: cost size

return



Worst-case Analysis

- For n execution of insertion operations
 - o A bad analysis: the worst case for the insertion is the case when expansion is required, up to *n*
 - o So, the worst case cost is in $O(n^2)$.
- Note the expansion is required during the ith operation only if i= 2^k , and the cost of the ith operation

$$c_i = \begin{cases} i & \text{if } i-1 \text{ is exactly the power of 2} \\ 1 & otherwise \end{cases}$$

So the total cost is: $\sum_{i=1}^{n} c_i \leq n + \sum_{j=0}^{\lfloor \log n \rfloor} 2^j < n + 2n = 3n$



Amortized Analysis – Why?

- Unusually expensive operations
 - o E.g., Insert-with-array-doubling
- Relation between expensive and usual operations
 - o Each piece of the doubling cost corresponds to some previous insert

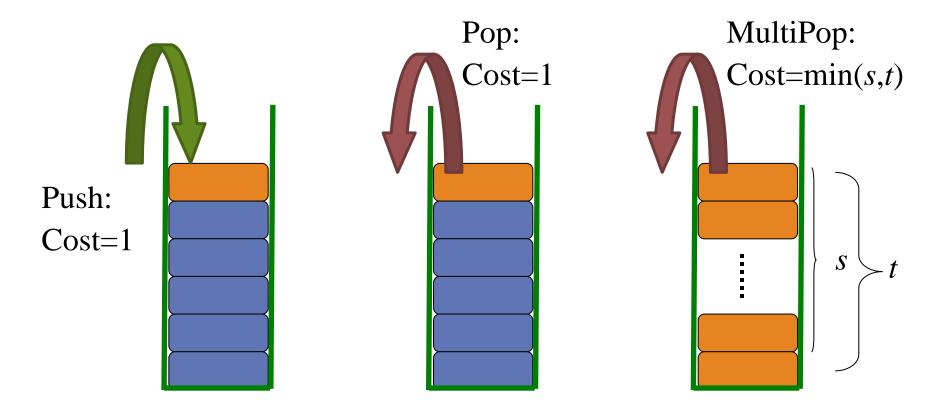


Amortized Analysis – How?

- Amortized equation: amortized cost = actual cost + accounting cost
- Design goals for accounting cost
 - o In any legal sequence of operations, the sum of the accounting costs is nonnegative
 - The amortized cost of each operation is fairly regular, in spite of the wide fluctuate possible for the actual cost of individual operations



Multi-pop Stack



Amortized cost: push:2; pop, multipop: 0



Multi-pop Stack

- Why non-negative accounting cost?
 - o For any possible sequence of operations?

	Amortized	Actual	Accounting
Push	2	1	1
Multi-pop	0	k	-k

k is the number of elements upon multi-pop

Binary Counter

```
0000000
\mathbf{0}
                     0
   0000001
   0000010
                     3
                             Cost measure: bit flip
3
   00000011
4
   00000100
   00000101
   00000110
                                amortized cost:
                      10
   00000111
                      11
                                set 1: 2
                     15
   00001000
8
                                set 0: 0
9
   00001001
                      16
                     18
10
   00001010
11
   00001011
                     19
   00001100
12
                     22
13
   00001101
                     23
   00001110
14
15
   00001111
                     26
   00010000
16
```



Binary Counter

- Why non-negative accounting cost?
 - o For any possible sequence of operations?

	Amortized	Actual	Accounting
Set 1	2	1	1
Set 0	0	1	-1



Array Doubling

- Why non-negative accounting cost?
 - o For any possible sequence of operations?

	Amortized	Actual	Accounting
Insert (normal)	3	1	2
Insert (doubling)	3	k+1	-k+2

k is the number of elements upon doubling

Thank you!

Q & A

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