



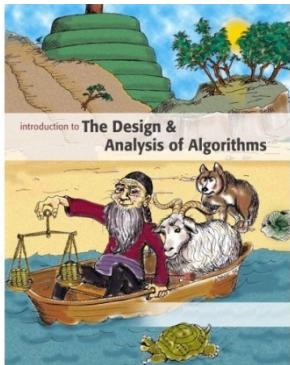
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Introduction to

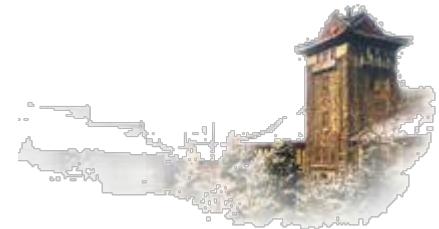
Algorithm Design and Analysis

[9] Hashing



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In the Last Class...

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



Hashing

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



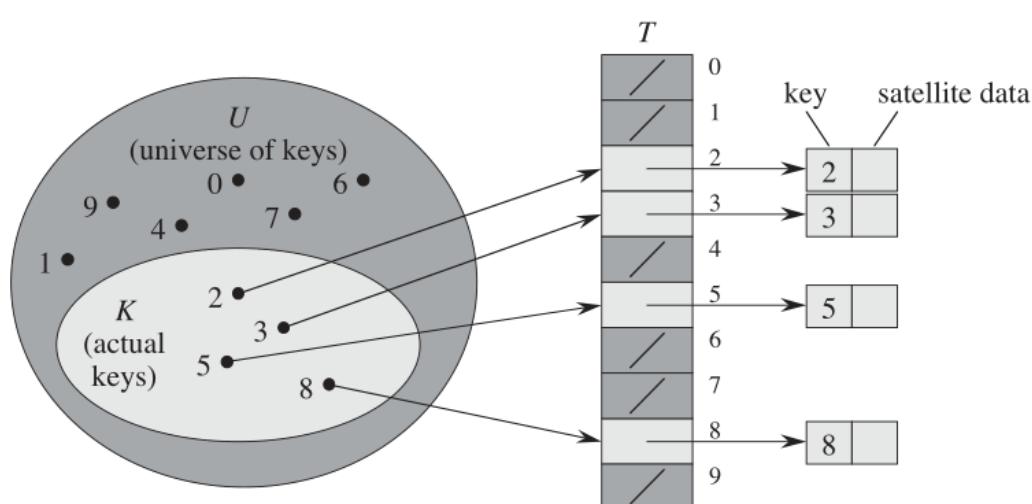
Cost for Searching

- Brute force
 - $O(n)$
- Balanced BST
 - $O(\log n)$
- Hashing – almost constant time
 - $O(1+\alpha)$
- “Mission impossible”
 - $O(1)$



Searching - a Brute Force Approach

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

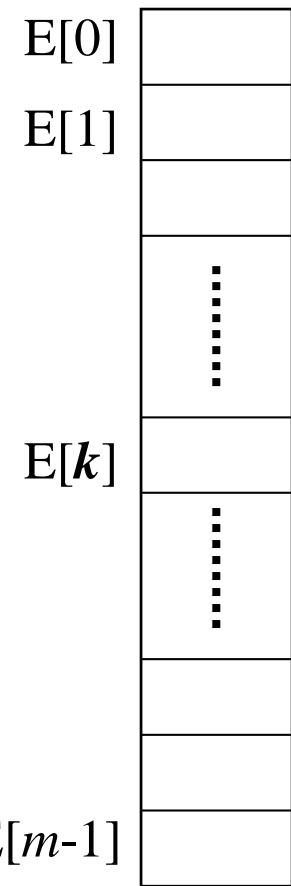


Direct-address Table

```
DIRECT-ADDRESS-SEARCH ( $T, k$ )
    return  $T[k]$ 
DIRECT-ADDRESS-INSERT ( $T, x$ )
     $T[key[x]] := x$ 
DIRECT-ADDRESS-DELETE ( $T, x$ )
     $T[key[x]] := \text{NIL}$ 
```

Hashing: the Idea

Hash Table (in feasible size)



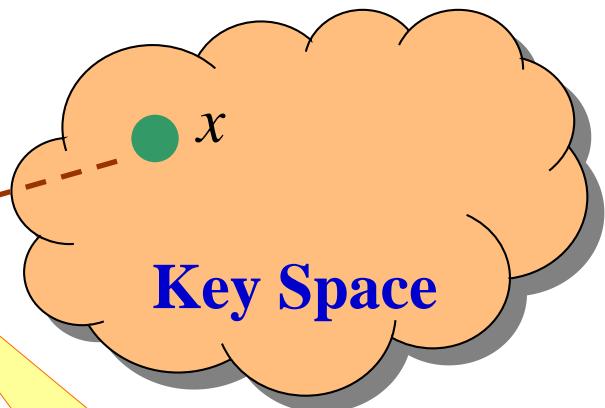
- ***Index distribution***
- ***Collision handling***

Hash
Function

$$H(x)=k$$

A calculated
array index for
the key

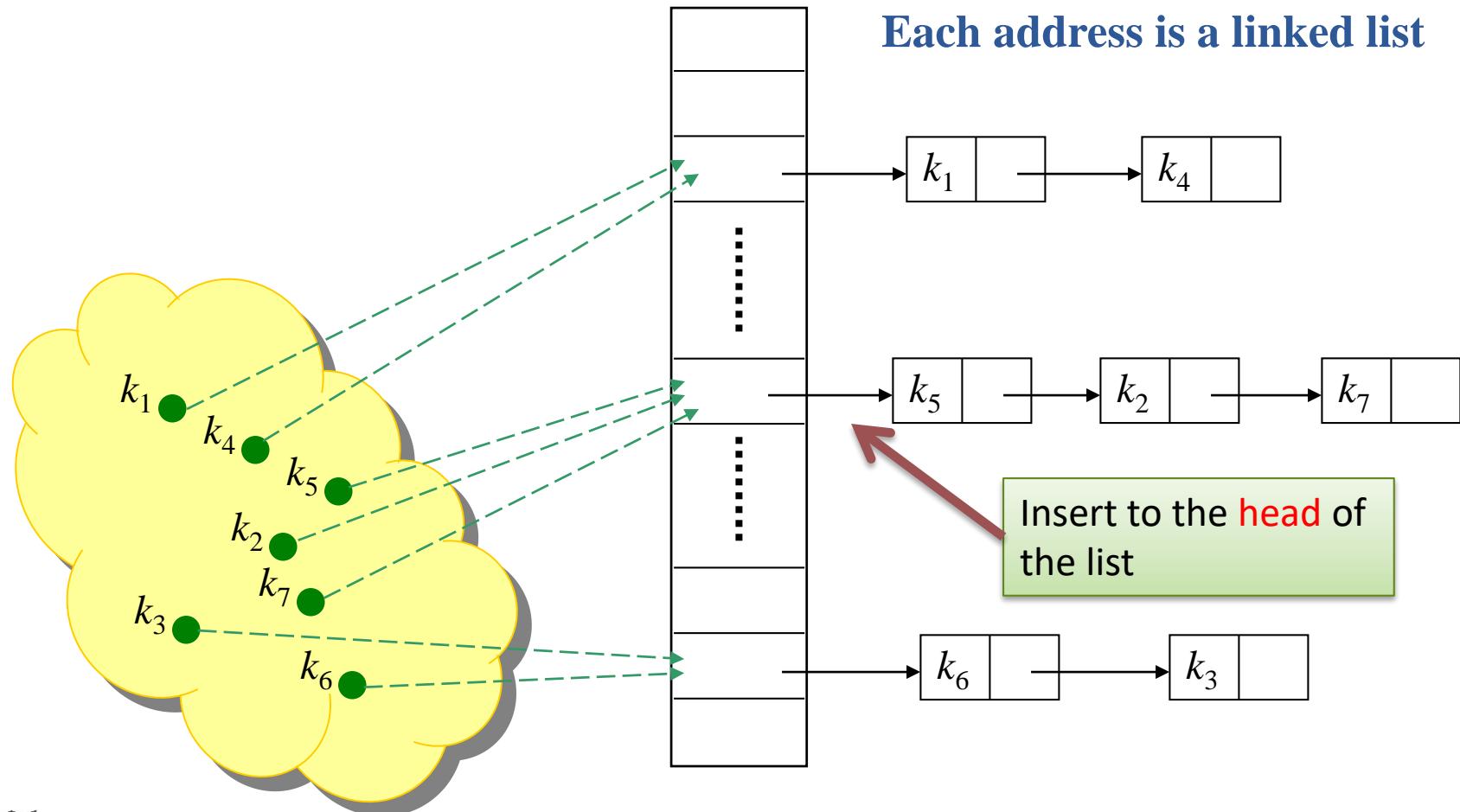
quite large, but only a
small part is used in an
application



Value of a
specific key



Collision Handling: Closed Address



Closed Address - Analysis

- **Assumption - simple uniform hashing**
 - For $j=0,1,2,\dots,m-1$, the average length of the list at $E[j]$ is n/m .
- **The average cost for an unsuccessful search**
 - Any key that is not in the table is equally likely to hash to any of the m address.
 - 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 i^{th} element inserted into the table, $i=1,2,\dots,n$)
 - For each i , the probability of that x_i is searched is $1/n$.
 - 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

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

- How to compute t ?

- Consider the *construction* process of the hash table



Closed Address - Analysis

- For successful search: (assuming that x_i is the i^{th} element inserted into the table, $i=1,2,\dots,n$)
 - For each i , the probability of that x_i is searched is $1/n$.
 - 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_j is inserted into the same list of x_i is $1/m$. So, the cost is:

Cost for
computing
hashing

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

Expected number of
elements in front of the
searched one in the same
linked list.



Closed Address: Analysis

- The average cost of a successful search:

- Define $\alpha = n/m$ as *load factor*,

The average cost of a successful search is :

$$\begin{aligned} \frac{1}{n} \sum_{i=1}^n \left(1 + \sum_{j=i+1}^n \frac{1}{m} \right) &= 1 + \frac{1}{nm} \sum_{i=1}^n (n-i) = 1 + \frac{1}{nm} \sum_{i=1}^{n-1} i \\ &= 1 + \frac{n-1}{2m} = 1 + \frac{\alpha}{2} - \frac{\alpha}{2n} = \Theta(1 + \alpha) \end{aligned}$$

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

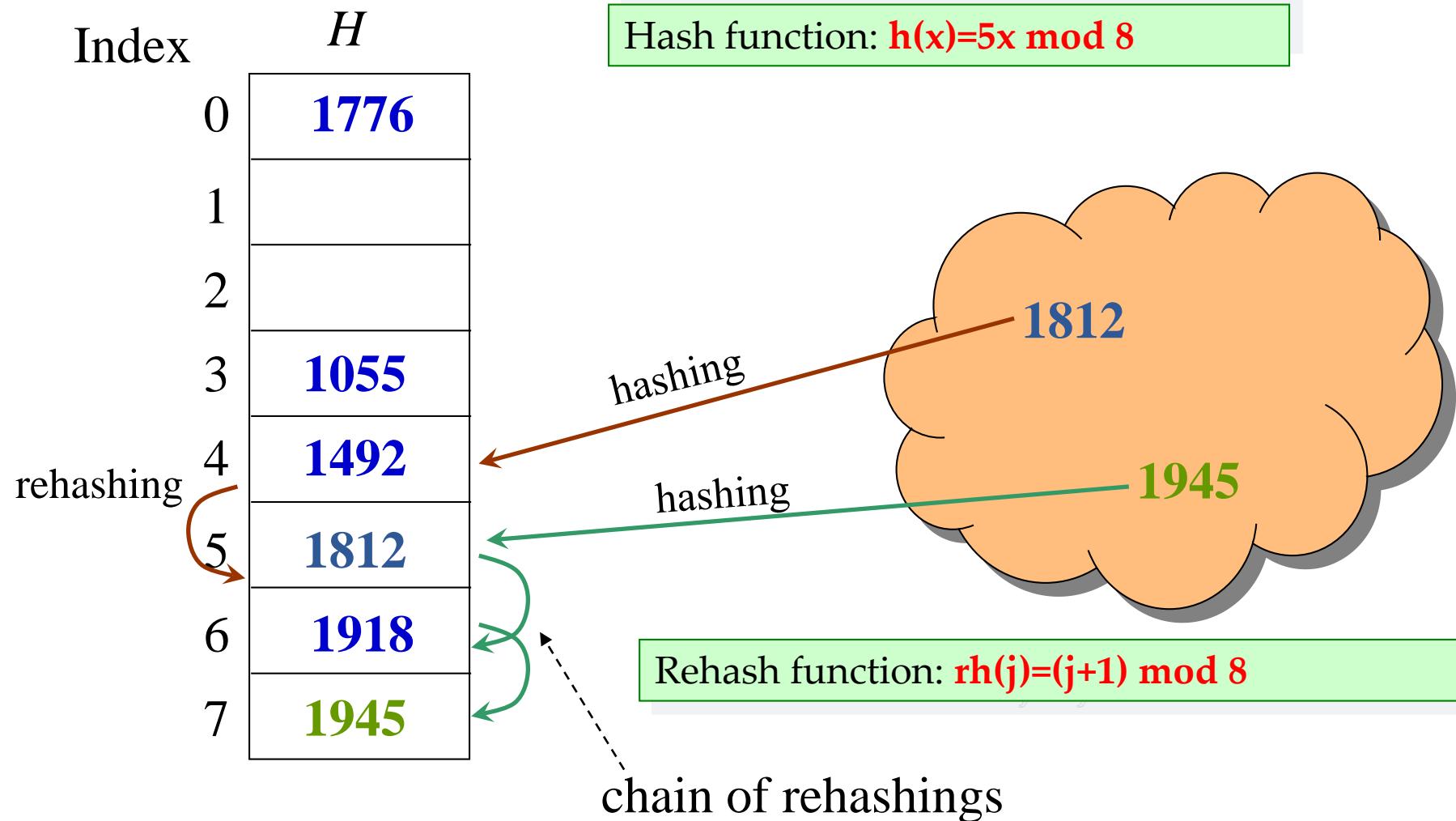


Collision Handling: Open Address

- All elements are stored in the hash table
 - No linked list is used
 - The load factor α cannot be larger than 1
- Collision is settled by “rehashing”
 - 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 $(0,1,2,\dots, m-1)$



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)**

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

Quadratic Probing:

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

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

Double hashing:

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

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



Equally Likely Permutations

- **Assumption**
 - Each key is equally likely to have any of the $m!$ permutations of $(1, 2, \dots, m)$ 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 $1/(1-\alpha)$ ($\alpha=n/m<1$)
 - 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:

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

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

See [CLRS] p.1199, C.25



Analysis for Open Address Hashing

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

$$\begin{aligned}\frac{1}{n} \sum_{i=0}^{n-1} \frac{m}{m-i} &= \frac{m}{n} \sum_{i=0}^{n-1} \frac{1}{m-i} = \frac{1}{\alpha} \sum_{i=m-n+1}^m \frac{1}{i} \\ &\leq \frac{1}{\alpha} \int_{m-n}^m \frac{dx}{x} = \frac{1}{\alpha} \ln \frac{m}{m-n} = \frac{1}{\alpha} \ln \frac{1}{1-\alpha}\end{aligned}$$

For your reference:

Half full: 1.387;

90% full: 2.559



Hash Function

- A good hash function satisfies the assumption of simple uniform hashing
 - Heuristic hashing functions
 - The division method: $h(k) = k \bmod m$
 - The multiplication method: $h(k) = \lfloor m(kA \bmod 1) \rfloor$ ($0 < A < 1$)
 - No single function can avoid the worst case $\Theta(n)$.
 - So “universal hashing” is proposed.
 - 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 $\Theta(1+\alpha)$
 - 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
 - `return`
- Insertion with expansion: cost $size$



Worst-case Analysis

- For n execution of insertion operations
 - A bad analysis: the worst case for one insertion is the case when expansion is required, up to n
Of course NOT!
 - So, the worst case cost is in $O(n^2)$.
- Note the expansion is required during the i th operation only if $i=2^k$, and the cost of the i th operation

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

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



Amortized Analysis – How?

- **Amortized equation:**
$$\text{amortized cost} = \text{actual cost} + \text{accounting cost}$$
- **Design goals for accounting cost**
 - 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



Array Doubling

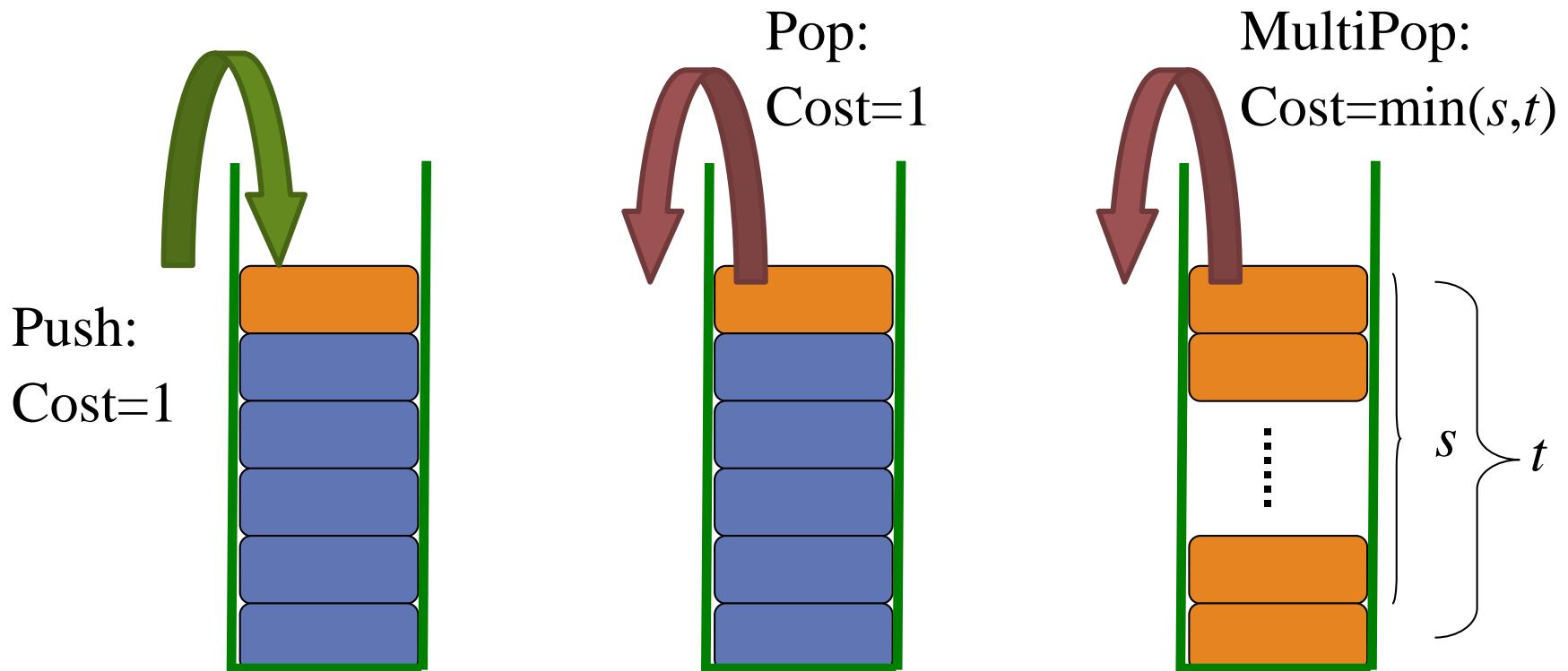
- Why non-negative accounting cost?
 - For any possible sequence of operations?

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

k is the number of elements upon doubling



Multi-pop Stack



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



Multi-pop Stack

- Why non-negative accounting cost?
 - 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

0	00000000	0
1	00000001	1
2	00000010	3
3	00000011	4
4	00000100	7
5	00000101	8
6	00000110	10
7	00000111	11
8	00001000	15
9	00001001	16
10	00001010	18
11	00001011	19
12	00001100	22
13	00001101	23
14	00001110	25
15	00001111	26
16	00010000	31

Cost measure: bit flip

amortized cost:

set 1: 2

set 0: 0



Binary Counter

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

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



Thank you!

Q & A

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