



Design and Analysis  
of Algorithms I

# Data Structures

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Hash Tables: Some  
Implementation Details

# Hash Table: Supported Operations

Purpose : maintain a (possibly evolving) set of stuff.  
(transactions, people + associated data, IP addresses, etc.)

Insert : add new record

Using a “key”

Delete : delete existing record

Lookup : check for a particular record  
( a “dictionary” )

AMAZING  
GUARANTEE

All operations in  
 $O(1)$  time ! \*

\* 1. properly implemented    2. non-pathological data

# High-Level Idea

Setup : universe  $U$  [e.g., all IP addresses, all names, all chessboard configurations, etc. ]  
[ generally, REALLY BIG ]

Goal : want to maintain evolving set  $S \subseteq U$   
[ generally, of reasonable size ]

Solution : 1.) pick  $n = \#$  of “buckets” with  $n \sim \text{size of } S$ . Assume  $n=2^*S$   
(for simplicity assume  $|S|$  doesn't vary much)

2.) choose a hash function  $h : U \rightarrow \{0, 1, 2, \dots, n-1\}$

3.) use array  $A$  of length  $n$ , store  $x$  in  $A[h(x)]$

can resize size of hash table if  $|S|$  is dynamic in practice

Hash function tells  
us where to store the given key  
 $x$  belonging to universe  $U$

## Naïve Solutions

1. Array-based  
solution

[ indexed by  $u$  ]

-  $O(1)$  operations  
but  $\theta(|U|)$  space

2. List-based  
solution

something like linked list

-  $\theta(|S|)$  space but  
 $\theta(|S|)$  Lookup

Consider  $n$  people with random birthdays (i.e., with each day of the year equally likely). How large does  $n$  need to be before there is at least a 50% chance that two people have the same birthday?

- ☒ 23 50 %
- ☐ 57 99 %
- ☐ 184 99.99....%
- ☐ 367 100%

BIRTHDAY  
"PARADOX"

Implication:  
If you had 10000 indexes in your array,  
just 100 keys are enough to have probability of collision=50%

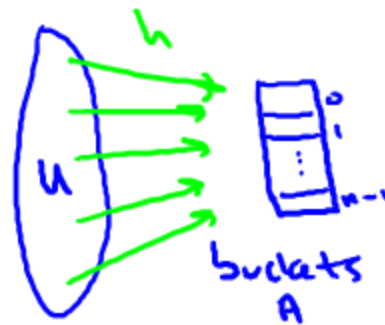
# Resolving Collisions

Collision: distinct  $x, y \in U$  such that  $h(x) = h(y)$

Solution #1 : (separate) chaining

- keep linked list in each bucket
- given a key/object  $x$ , perform Insert/Delete/Lookup in the list in  $A[h(x)]$

Linked list for  $x$  → Bucket for  $x$



Solution #2 : open addressing. (only one object per bucket)

- Hash function now specifies probe sequence  $h_1(x), h_2(x), \dots$   
(keep trying till find open slot)

Use 2 hash functions

- Examples : linear probing (look consecutively), double hashing

# What Makes a Good Hash Function?

Note : in hash table with chaining, Insert is  $\theta(1)$   
 $\theta(\text{list length})$  for Insert/Delete.

Insert new object  $x$  at  
front of list in  $A[h(x)]$

could be anywhere from  $m/n$  to  $m$  for  $m$  objects

Equal-length lists

Point : performance depends on the choice of hash function!  
(analogous situation with open addressing)

All  
objects in  
same  
bucket

## Properties of a “Good” Hash function

1. Should lead to good performance  $\Rightarrow$  i.e., should “spread data out” (gold standard – completely random hashing)
2. Should be easy to store/ very fast to evaluate.

Completely random hashing is as bad as a vanilla linked list to store/evaluate.  
We would have to remember entire hash map, and loop up/evaluating hash function would take  $O(n)$  time

# Bad Hash Functions

Example : keys = phone numbers (10-digits).

$$|u| = 10^{10}$$

-Terrible hash function :  $h(x) = 1^{\text{st}} 3 \text{ digits of } x$   
(i.e., area code)

$$\text{choose } n = 10^3$$

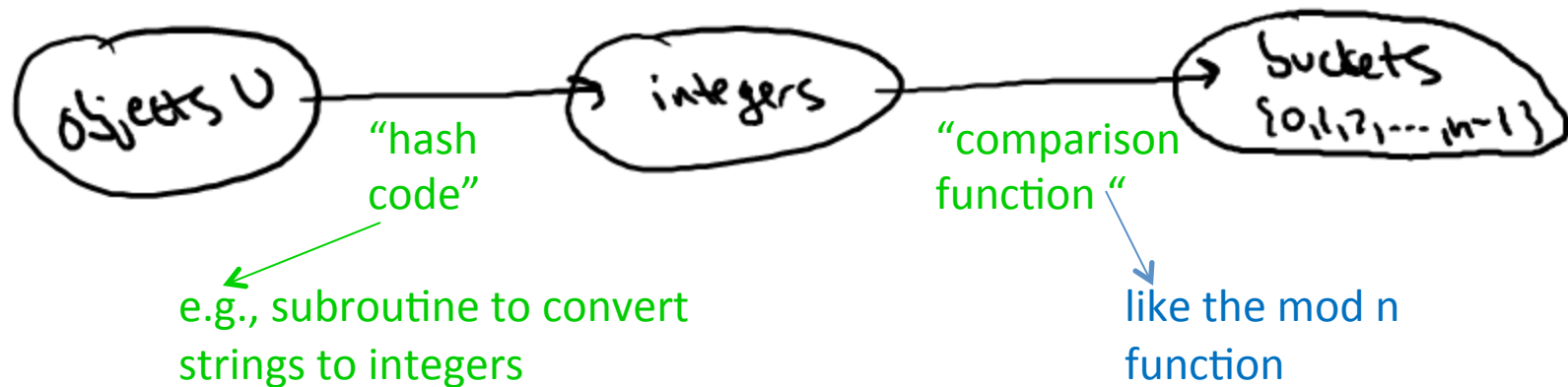
- mediocre hash function :  $h(x) = \text{last 3 digits of } x$   
[still vulnerable to patterns in last 3 digits]

Example : keys = memory locations. (will be multiples of a power of 2)

-Bad hash function :  $h(x) = x \bmod 1000$  (again  $n = 10^3$ )

=> All odd buckets guaranteed to be empty.

# Quick-and-Dirty Hash Functions



## How to choose $n = \#$ of buckets

1. Choose  $n$  to be a prime ( within constant factor of  $\#$  of objects in table)
2. Not too close to a power of 2
3. Not too close to a power of 10