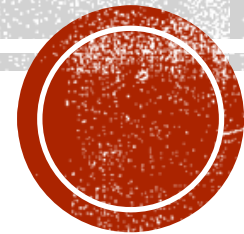




Indian Institute of Information Technology Allahabad

Data Structures

Hashing



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Today

- Hashing!
 - What operations are we trying to support?
 - Hash Functions
 - Dealing with collisions
 - What makes a good hash function?
 - *Universal* hash families are what we're looking for!

Hash Tables Overview

What operations does it support?

The Task

Again, we want to keep track of objects that have keys 5 (aka, **nodes with keys**)

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Sorted Arrays



$O(n)$ INSERT/DELETE: first, find the relevant element (via SEARCH) and move a bunch of elements in the array

$O(\log n)$ SEARCH: use binary search to see if an element is in A

Linked Lists



$O(1)$ INSERT: just insert the element at the head of the linked list

$O(n)$ SEARCH/DELETE: since the list is not necessarily sorted, you need to scan the list (delete by manipulating pointers)

Hash Table: Motivation

| OPERATION | SORTED ARRAY | UNSORTED LINKED LIST | HASH TABLES (HOPEFULLY) |
|-----------|--------------|----------------------|-------------------------|
| SEARCH | $O(\log(n))$ | $O(n)$ | $O(1)$ |
| DELETE | $O(n)$ | $O(n)$ | $O(1)$ |
| INSERT | $O(n)$ | $O(1)$ | $O(1)$ |

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What is a *naive* way to achieve these runtimes?

Attempt 1: Direct Addressing

Suppose you're storing
numbers from 1 - 1000:



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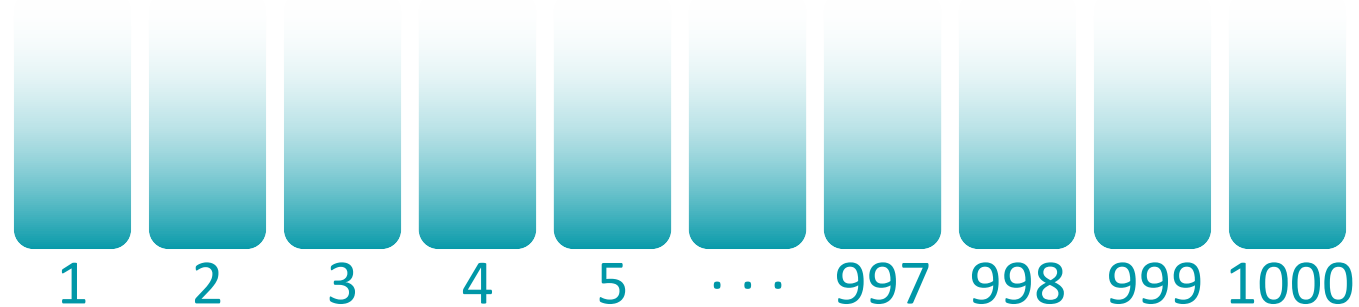
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(each address/bucket stores one type of item)

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$O(1)$ INSERT/DELETE/SEARCH: Just index into the bucket!

Attempt 1: Direct Addressing

Suppose you're storing
numbers from 1 - 1000:

2 4 5 998 999

Not bad!

But what's the issue with this
approach?

$O(1)$ INSERT/DELETE/SEARCH: Just index into the bucket!

Attempt 1: Direct Addressing

Suppose you're storing numbers from 1 - 10^{10} :

2 3 1000 1002 10^{10}

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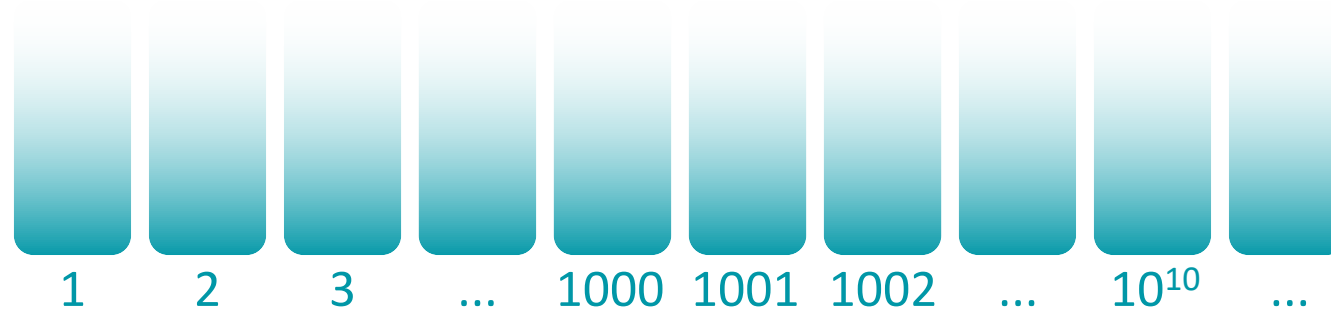
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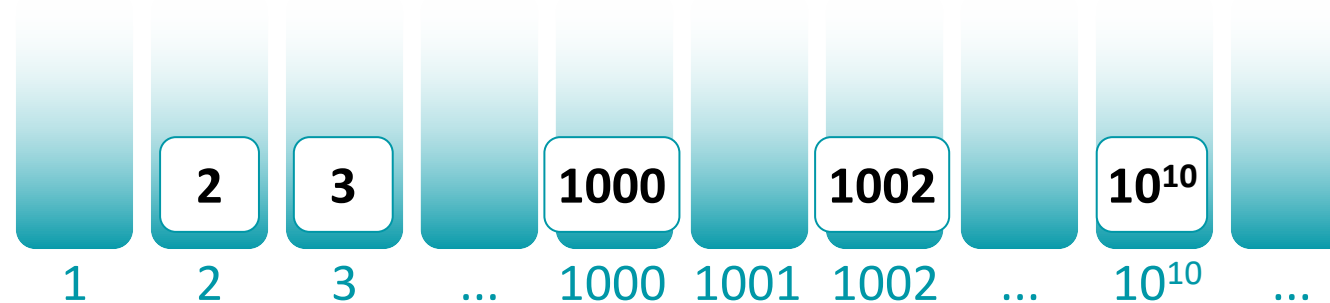
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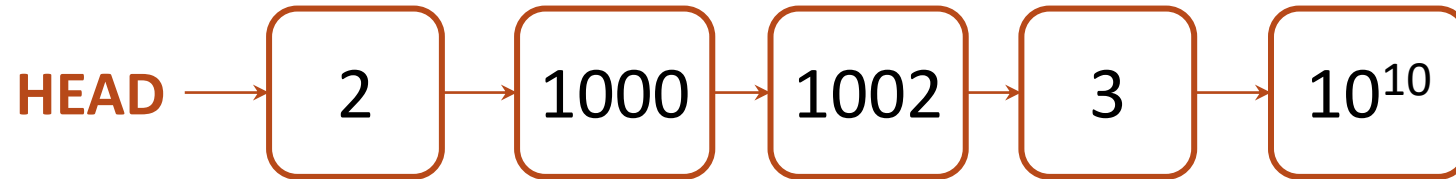
Reasonable Attempt (???): *Direct Addressing!*

**But the space requirement is
HUGE...**

$O(1)$ INSERT/DELETE/SEARCH: Just index into the bucket!

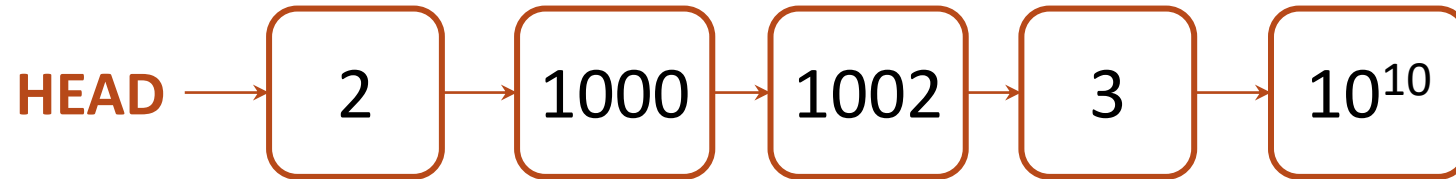
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- **Good news:** Space is now proportional to the number of objects you deal with
- **Bad news:** Searching for an object is now going to scale with the number of inputs you deal with... not close to our desired $O(1)$!
- The direct-addressing approach still has merit because of its fast object search/access

How to improve this?

We like the **functionality of a direct-addressable** array for
constant time access
(super fast INSERT/DELETE/SEARCH)

But reserving an bucket/array slot for each possible key leads to
unreasonable space requirements

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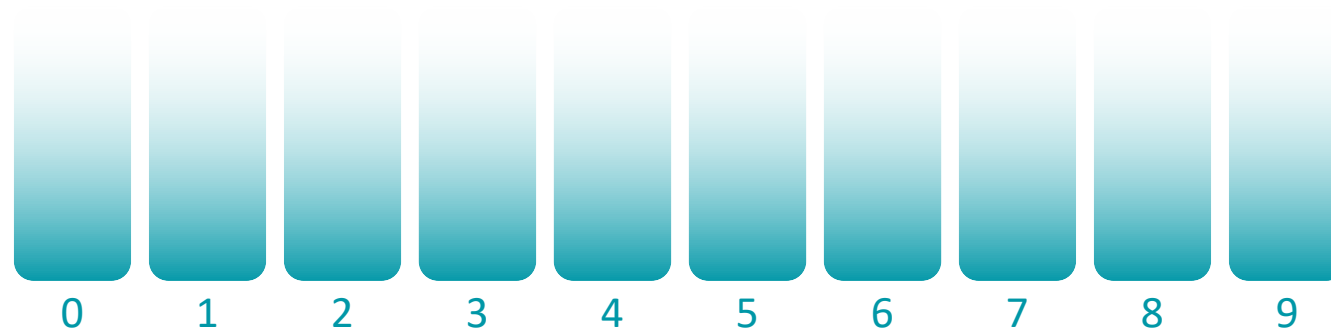
Let's try bucketing **by the least-
significant digit...**

Bucketing Attempt 1

Suppose you're storing numbers from 1 - 10^{10} :



Bucket by last digit?

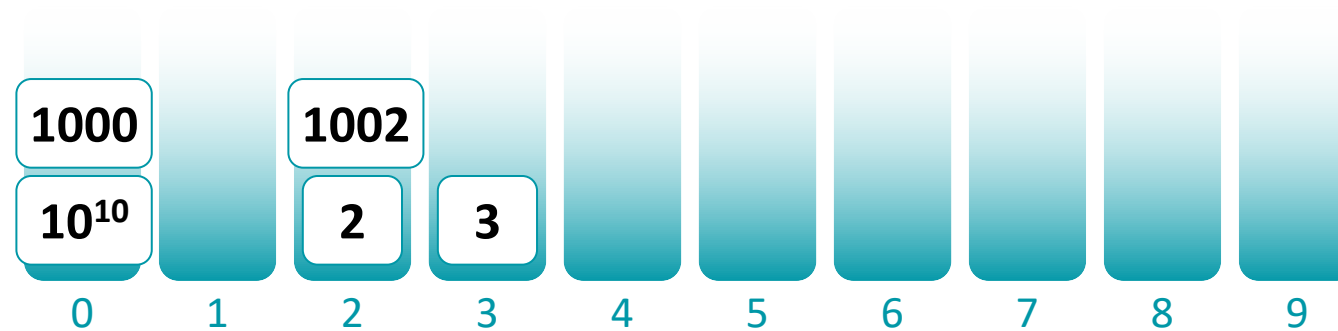


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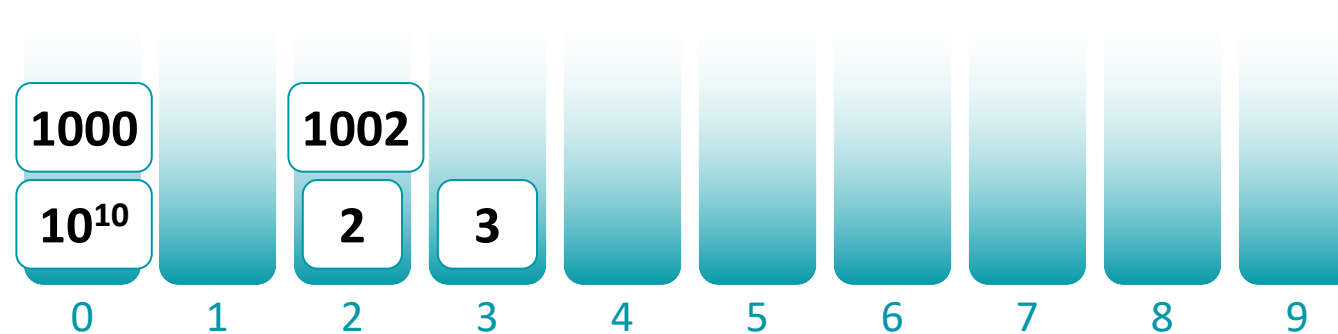


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$O(1)$ INSERT:

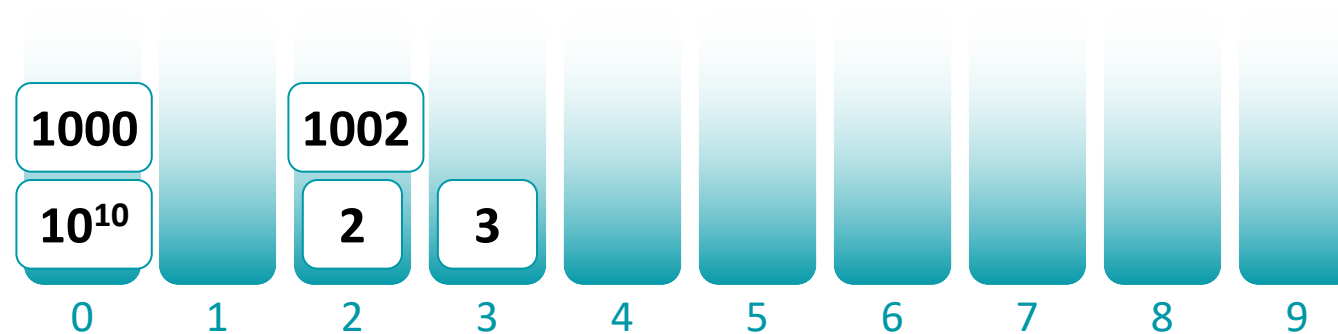
Just index into the bucket (& insert at front of a linked list)!

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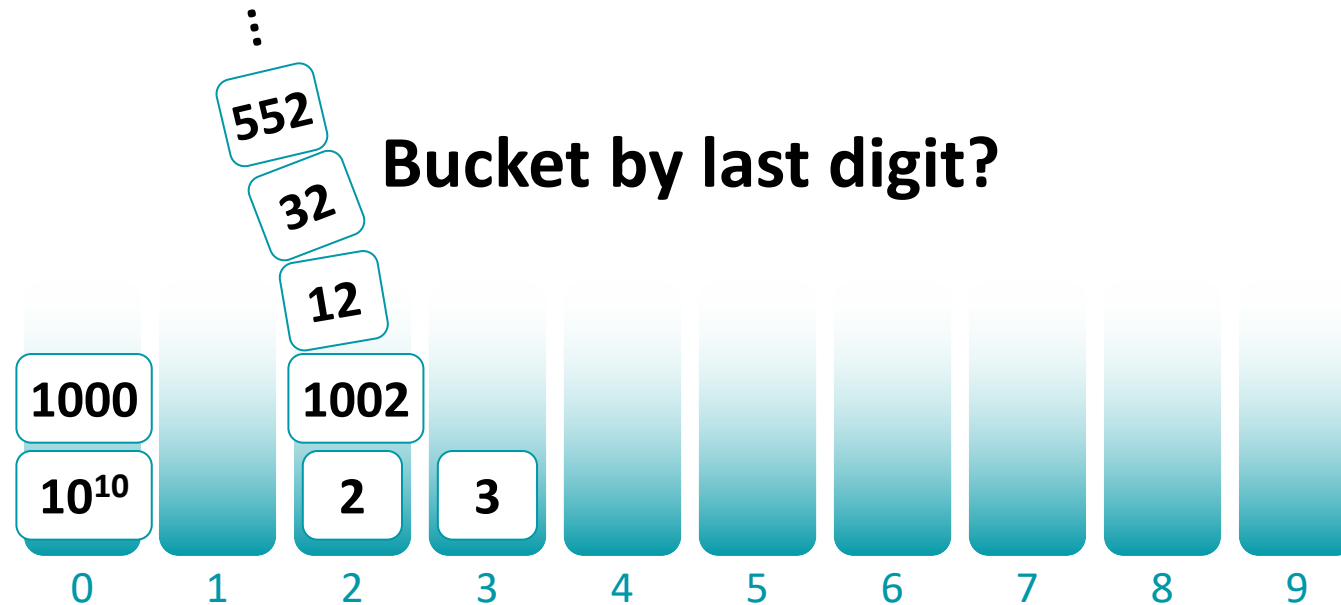


$O(\underline{??????})$ SEARCH/DELETE:

Go visit bucket & search through until you find it...

Bucketing Attempt 1

Under this scheme, a bad guy could give us inputs that yields quite ugly worst-case runtimes...

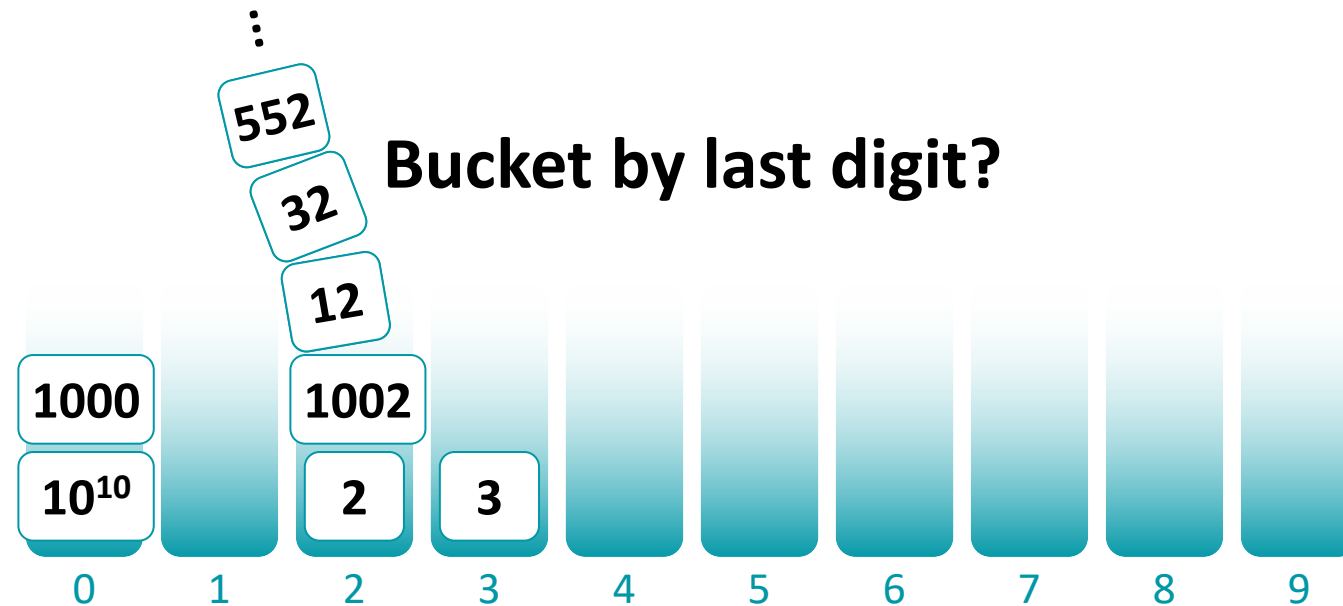


$O(n)$ SEARCH/DELETE:

Go visit bucket & search through until you find it...

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Maybe another bucketing scheme?

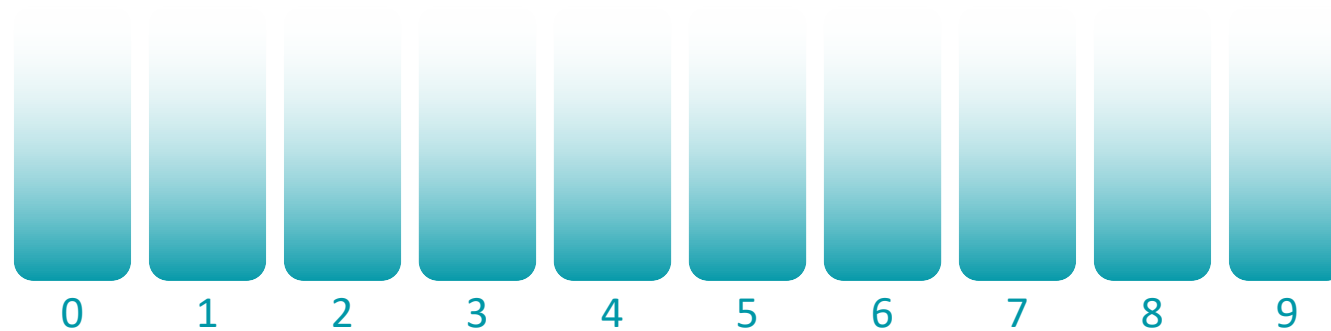
find it...

Bucketing Attempt 2

Suppose you're storing numbers from 1 - 10^{10} :



Bucket by last digit of (number * 7) mod 3

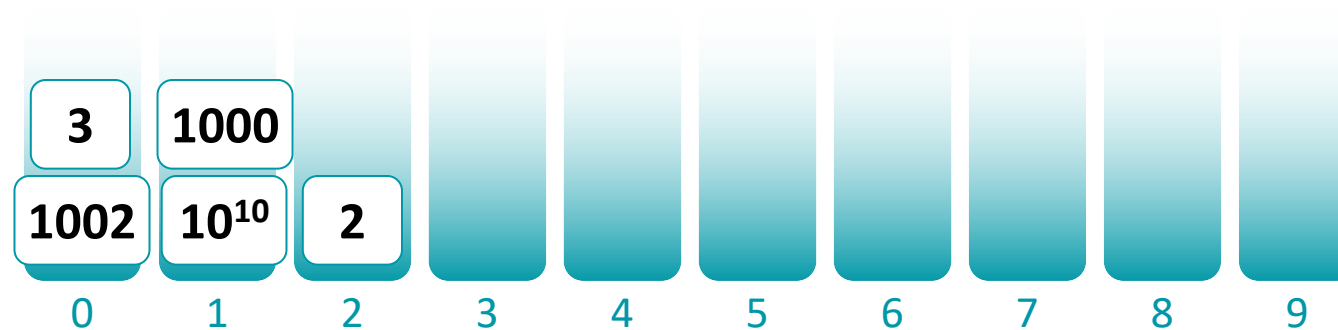


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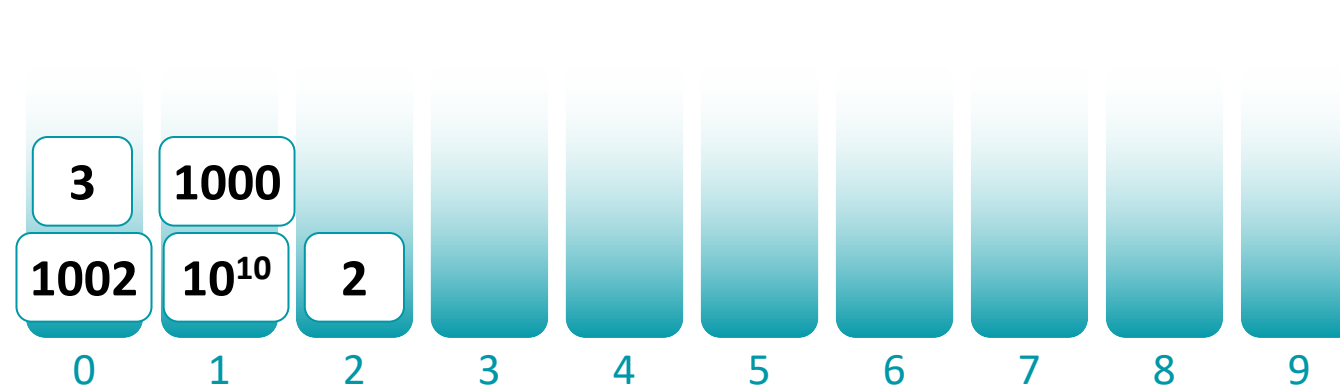


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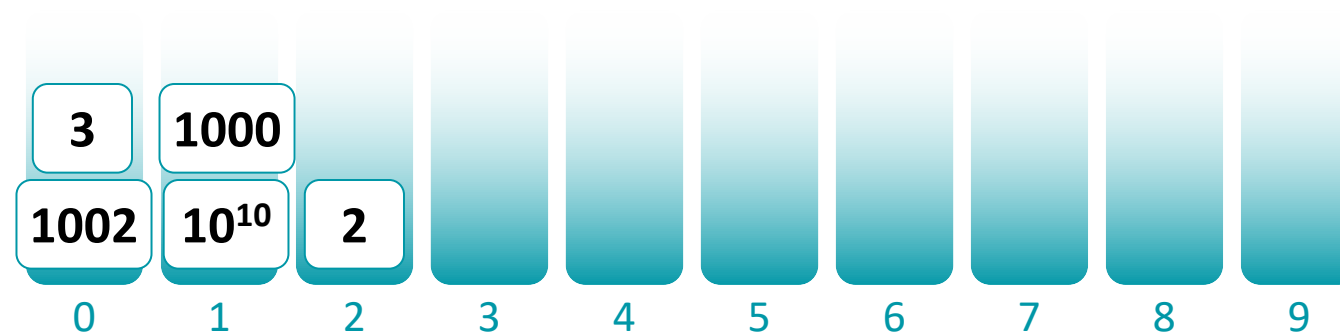
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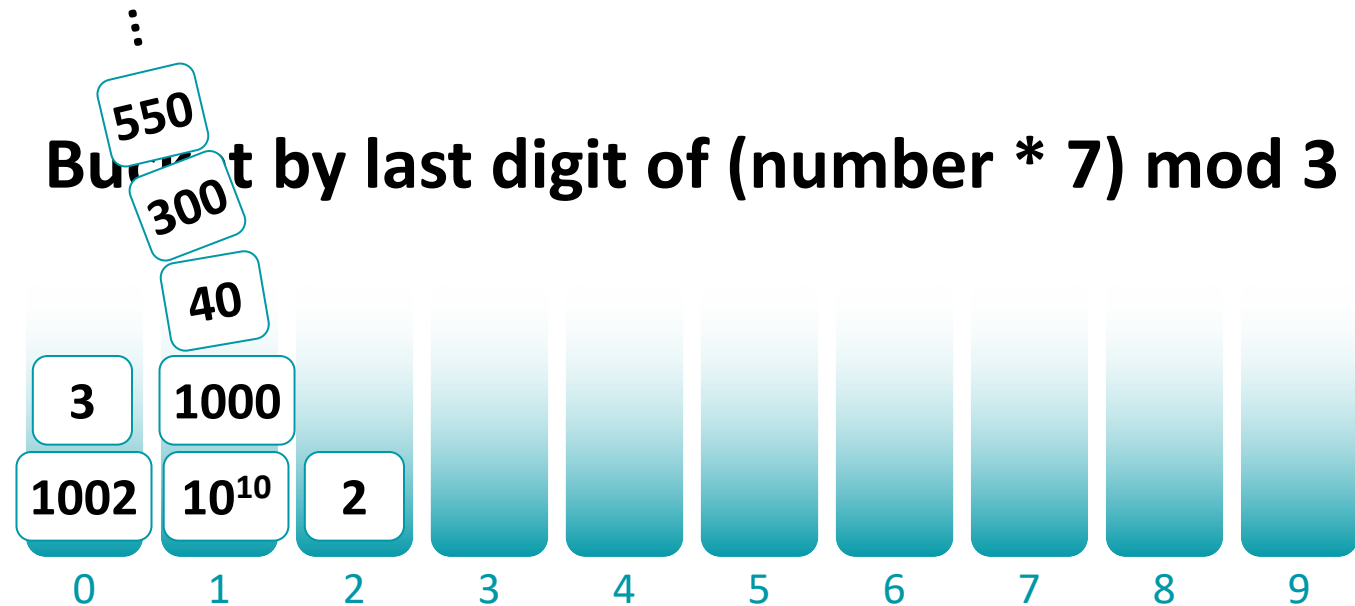


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$O(n)$ SEARCH/DELETE:

Go visit bucket & search through until you find it...

Bucketing Attempt 2

Under this scheme, a bad guy could give us inputs that yields quite ugly worst-case runtimes...

⋮

Seems like a bad guy could still thwart us.
There are other bucketing schemes we could use,
so to reason about them more formally,
let's talk about **HASH FUNCTIONS**.

Go visit bucket & search through until you
find it...

Hash Functions

What are “good” hash functions?

Some Terminology

There exists a universe **U** of keys, with size M .

Generally, M is *really big*. Examples:

- U = the set of all ASCII strings of length 20. $M = 26^{20}$
- U = the set of all IPv4 addresses. $M = 2^{32}$
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A hash function $h: \mathbf{U} \rightarrow \{1, \dots, n\}$
maps elements of **U** to buckets $1, \dots, n$

Some Terminology

There exists a universe U of keys, with size M .

NOTE:

For this lecture, I'm assuming that the
of elements I receive = # of buckets (both are n).

This doesn't have to be the case, but we usually aim for

#buckets = $O(\# \text{ elements that show up})$
(otherwise, we're using "too much" space)

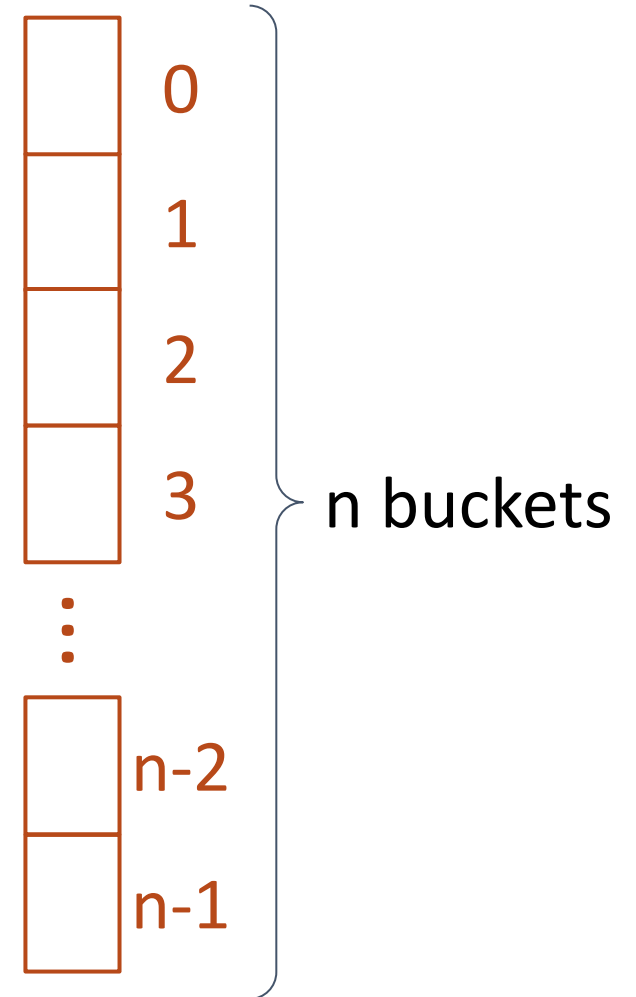
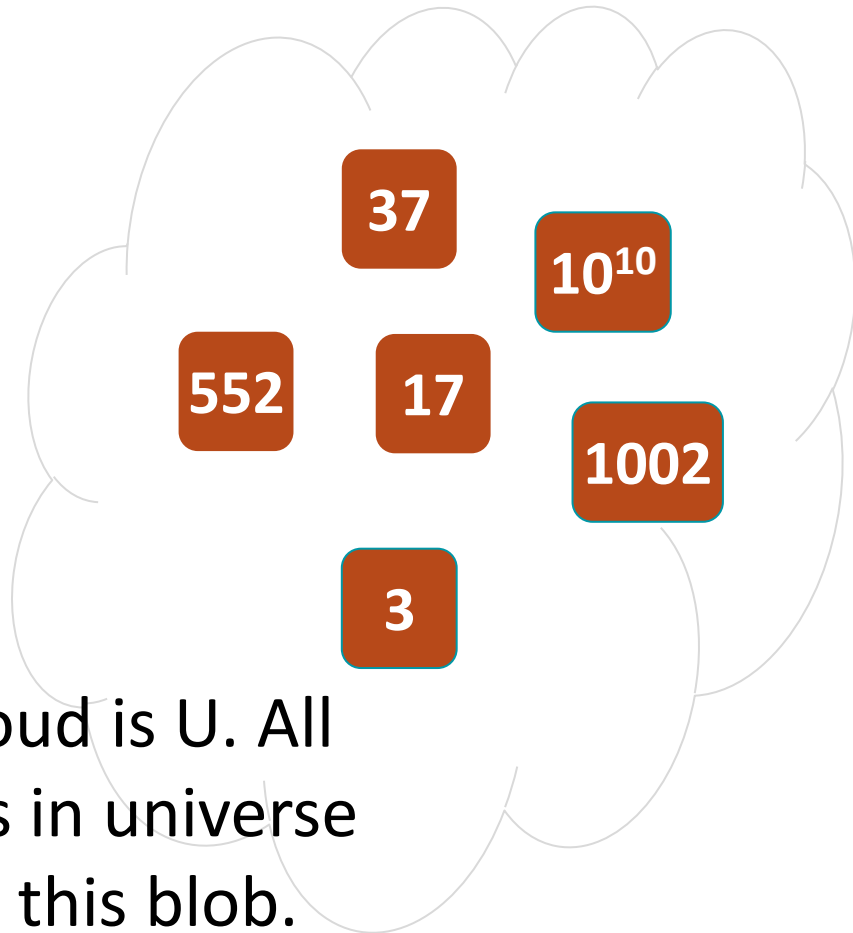
n

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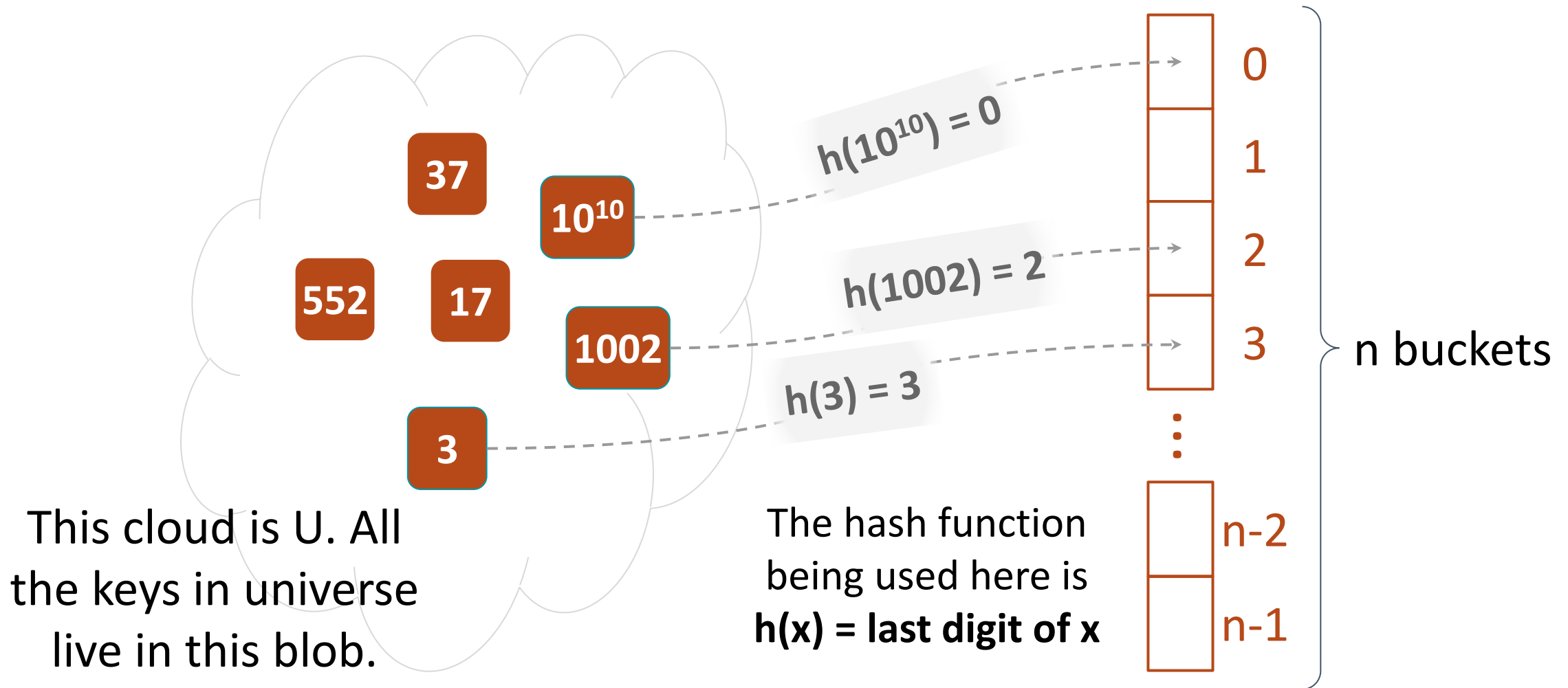
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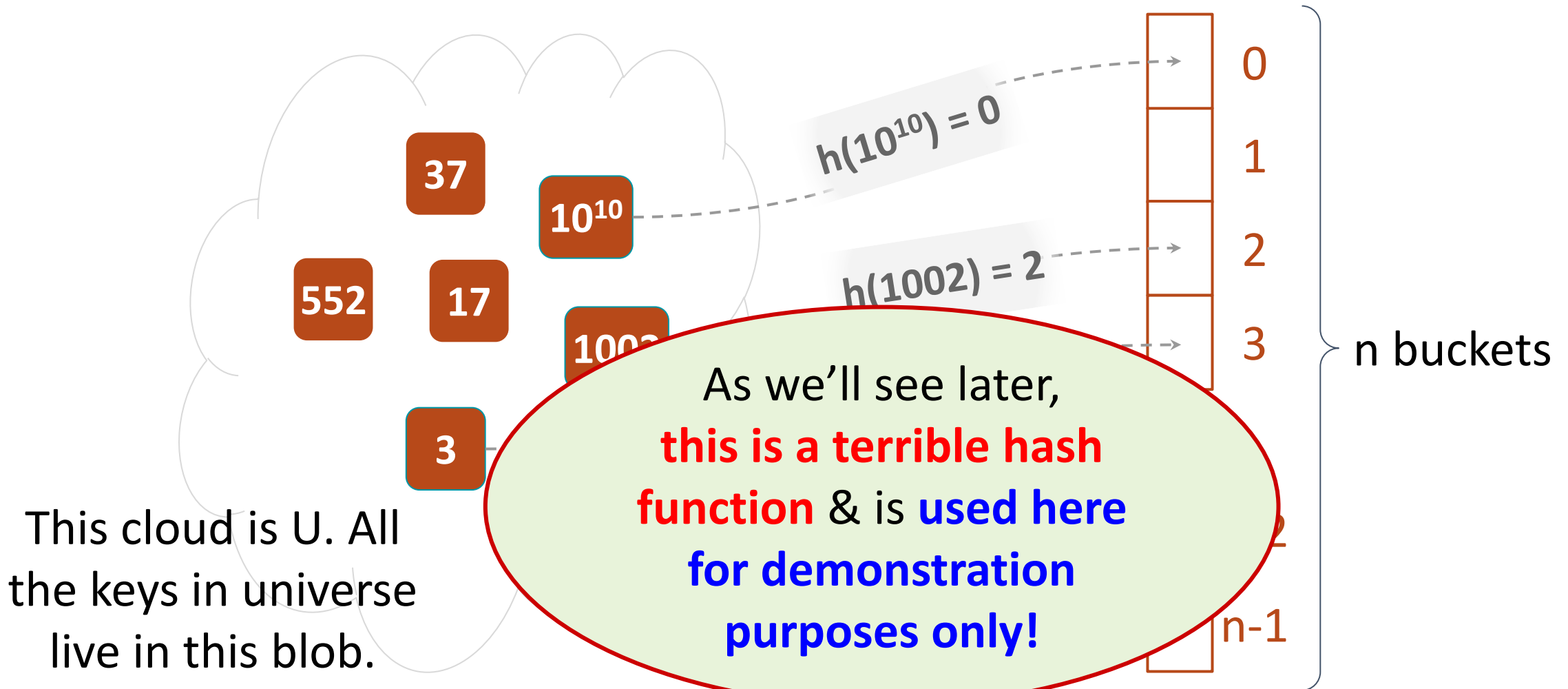
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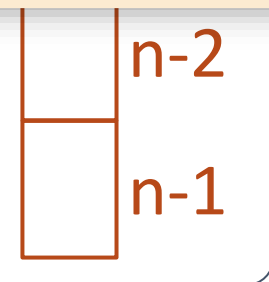
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- A hash function tells you where to start looking for an object.
- For example, if a particular hash function h has $h(1002) = 2$, then we say “1002 hashes to 2”, and we go to bucket 2 to search for 1002, or insert 1002, or delete 1002.

This cloud is U . All the keys in universe live in this blob.

The hash function being used here is $h(x) = \text{last digit of } x$

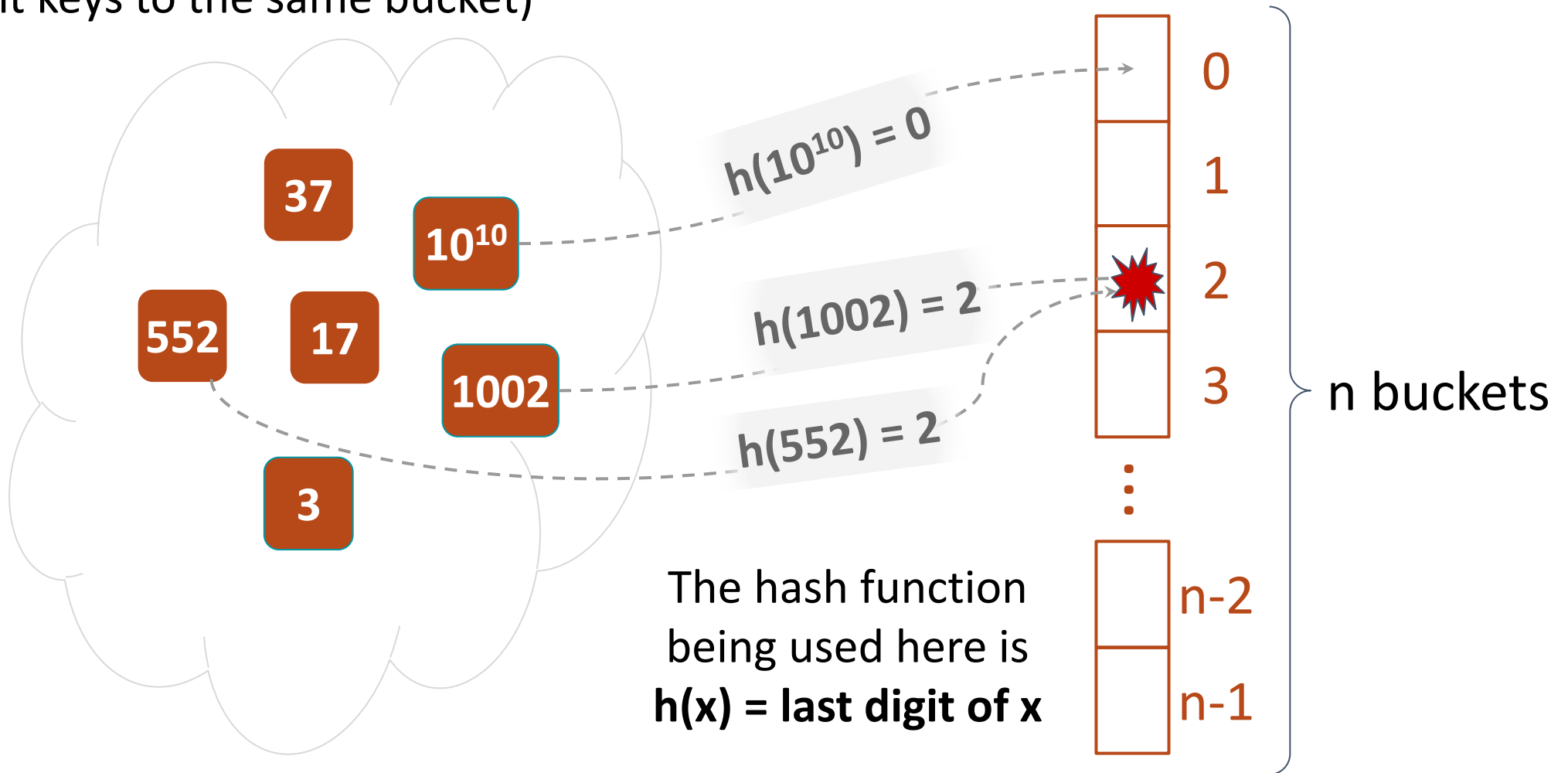


Collisions

Collisions are inevitable!

(when a hash function would map 2 different keys to the same bucket)

This is because of the *Pigeonhole Principle*. Since the size of universe $U > \#$ of buckets, every hash function (no matter how clever), suffers from at least one collision.



Collision Resolution: Chaining

To resolve collisions, one common method is to use **chaining!**

We're just giving a formal name to our bucketing example from earlier:
represent each bucket's contents as a *linked list* !

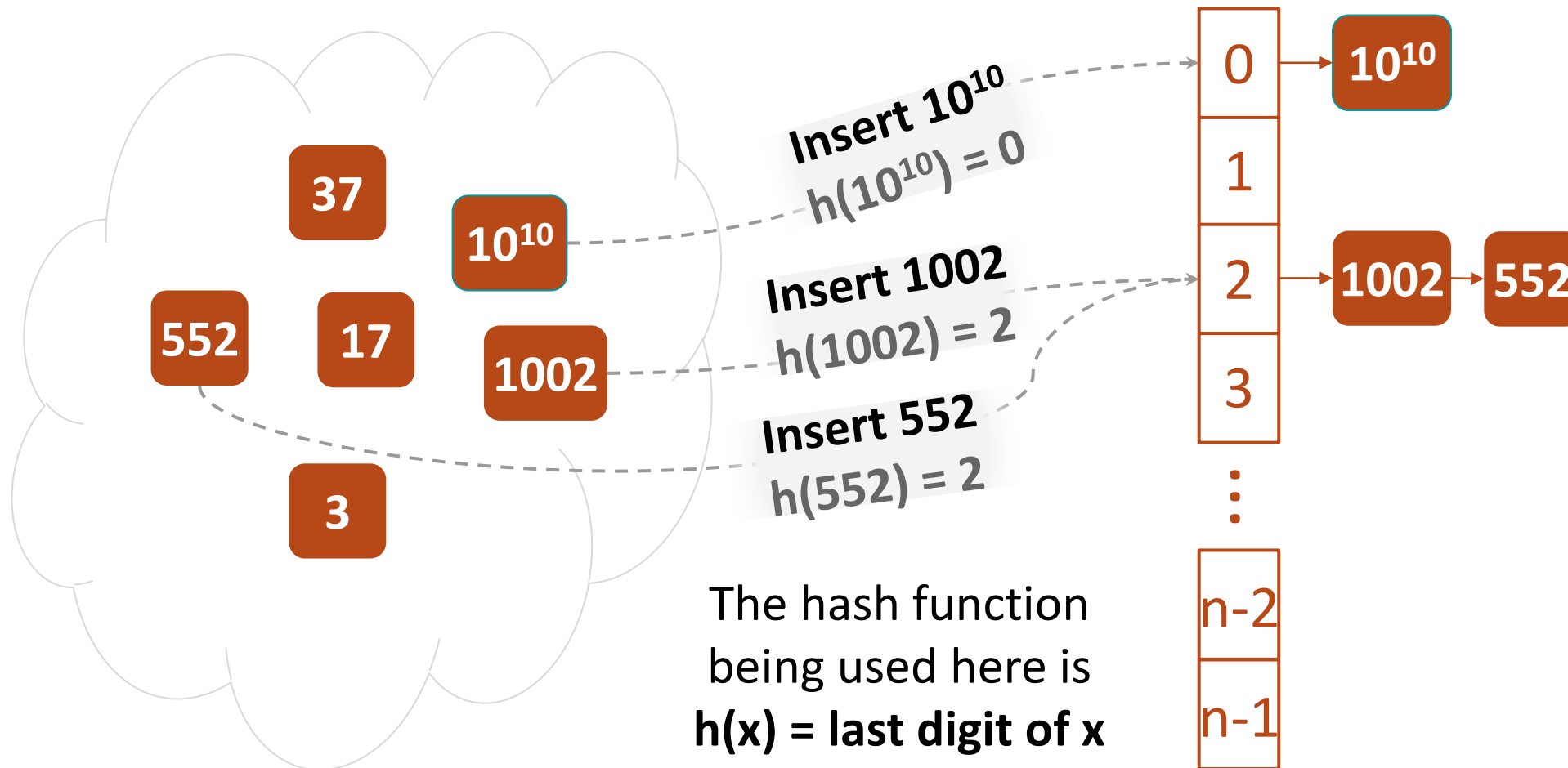
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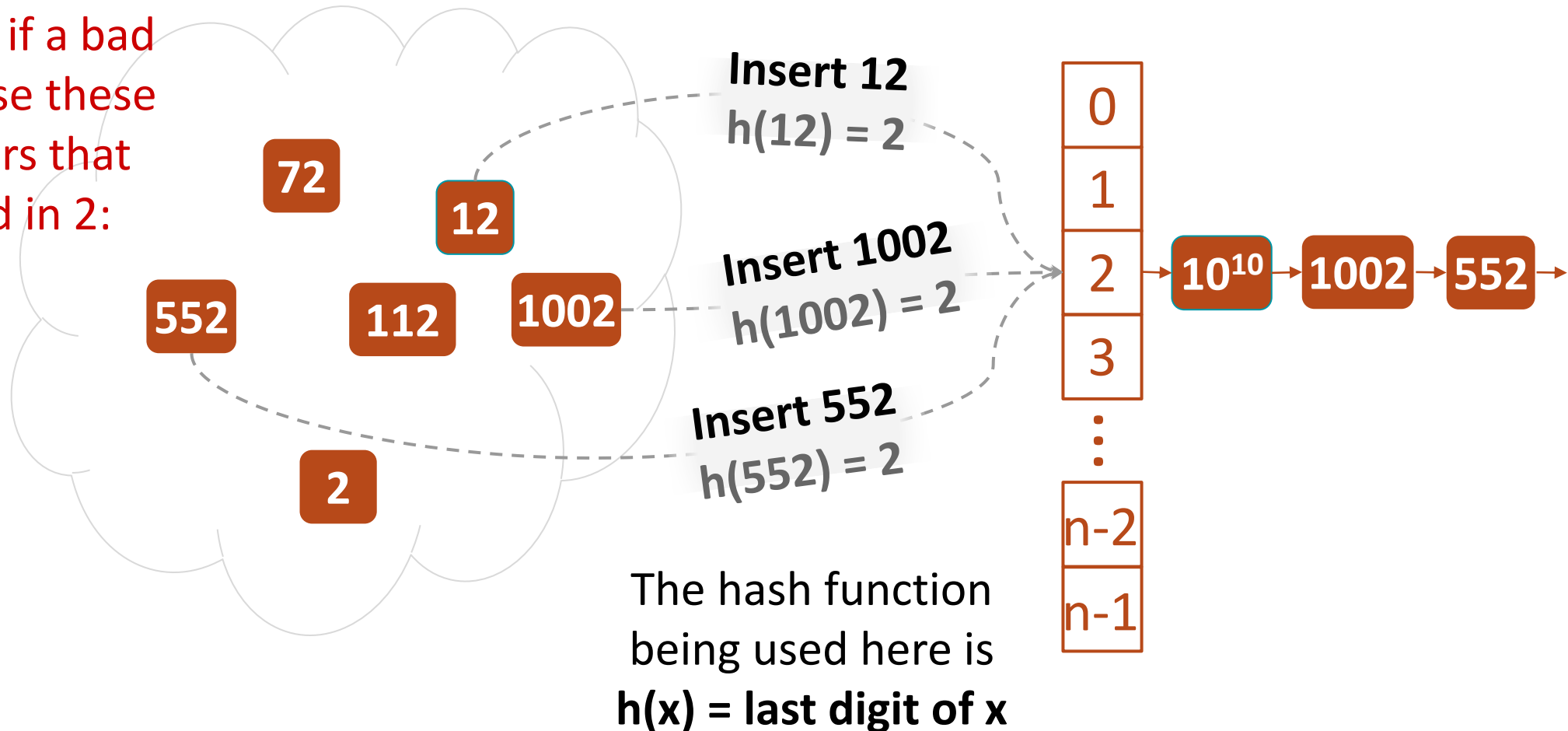
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Collision Resolution: Chaining

But if the items are all clumped together in a single bucket, SEARCH/DELETE may be very slow because of the linked list traversal...

Imagine if a bad guy chose these numbers that all end in 2:



Hash Table Goals

Remember worst-case analysis:

OUR GOAL: Design a function $h: U \rightarrow \{1, \dots, n\}$ so that no matter what n items of U a bad guy chooses & the operations they choose to perform, the buckets will be balanced.

(Here, balanced means $O(1)$ entries per bucket)

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- They get hashed into n buckets
- At least 1 bucket has at least M/n items hashed to it (Pigeonhole)
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The n items the bad guy chooses are items that all land in this very full bucket. That bucket has size $\Omega(n)$.

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Maybe there's a way to weaken the adversary...

LET'S BRING IN SOME

RANDOMNESS!

- The univer
- They get h
- At least 1
- M is wayy

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Hash Functions and Randomness

What it means to weaken the adversary & ways to do it

Intuition

So, our strategy is to define a set of hash functions, and then we randomly choose a hash function h from this set to use!

You can think of it like a game:

1. You announce your set of hash functions, H .
2. The adversary chooses n items for your hash function to hash.
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What would make a “good” set of hash functions H ?

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What would make a “good” set of hash functions H ?

What we want

Design a set $H = \{h_1, h_2, h_3, \dots, h_k\}$ where $h_i : U \rightarrow \{1, \dots, n\}$, such that if we chose a random h in H and after an adversary chooses n items $\{u_1, u_2, \dots, u_n\}$ to hash,

for any item u_i ,
the **expected** # of items in u_i 's bucket is **$O(1)$**

Let's see an example of a set of hash functions H that achieves this goal!

H = Exhaustive Set of All Hash Functions

WHAT WE WANT:

Design a set $\mathbf{H} = \{h_1, h_2, h_3, \dots, h_k\}$ where $h_i : U \rightarrow \{1, \dots, n\}$, such that if we chose a uniformly random \mathbf{h} in \mathbf{H} and after an adversary chooses n items $\{u_1, u_2, \dots, u_n\}$ to hash,

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\mathbf{H} = the exhaustive set of all hash functions that map elements in the universe U to buckets 1 to n .

\mathbf{H} contains a total of n^M hash functions.

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Here is an example

where

$U = \{“a”, “b”, “c”\}$

so $M = 3$. Also, we

have $n = 2$.

| | h_1 | h_2 | h_3 | h_4 | h_5 | h_6 | h_7 | h_8 |
|-----|-------|-------|-------|-------|-------|-------|-------|-------|
| “a” | 0 | 0 | 0 | 0 | 1 | 1 | 1 | 1 |
| “b” | 0 | 0 | 1 | 1 | 0 | 0 | 1 | 1 |
| “c” | 0 | 1 | 0 | 1 | 0 | 1 | 0 | 1 |

The 0's and 1's represent the binary buckets i.e. h_8 will hash “b” to bucket 1.

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$\mathbb{E}[\# \text{ of items in } u_i \text{ 's bucket}] =$

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This probability is taken over the random choice of hash function!

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How do we know that $P[h(u_i) = h(u_j)] = 1/n$?

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H = Exhaustive Set of All Hash Functions

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H contains a total of n^M hash functions.

$$\mathbb{E}[\# \text{ of items in } u_i \text{ 's bucket}] = \sum_{j=1}^n P[h(u_i) = h(u_j)]$$

This probability is taken over the random choice of hash function!

$$= P[h(u_i) = h(u_i)] + \sum_{j \neq i} P[h(u_i) = h(u_j)]$$

$$= 1 + \sum_{j \neq i} P[h(u_i) = h(u_j)]$$

How do we know that $P[h(u_i) = h(u_j)] = 1/n$?

$$= 1 + \sum_{j \neq i} \frac{1}{n}$$

$$= 1 + \frac{n-1}{n} \leq 2$$

O(1)

This is what we wanted!

H = Exhaustive Set of All Hash Functions

H = the exhaustive set of all hash functions that map elements in the universe U to buckets 1 to m

Good News:

H achieves our goal!

If we choose a *uniformly random hash function* from Exhaustive Set of All Hash Functions, then INSERT/DELETE/SEARCH on any n elements will have **expected runtime of $O(1)$.**

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Bad News:

How many bits does it take to store a uniformly random hash function?

A lot!

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We'd use a lookup table: one entry per element of U , each storing which bucket to hash that element to.

$(M \text{ elements}) * (\log(n) \text{ bits to write down a bucket \#}) = M \log n \text{ bits}$
This is HUGE... (& enough to do direct addressing!)

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How do we fix this size issue?

Universal Hash Families

“Good” sets of hash functions that aren’t as large!

What we wanted

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The fact that
 $P[h(u_i)=h(u_j)] = 1/n$
did all the work
here

$$\begin{aligned}&= 1 + \sum_{j \neq i} P[h(u_i) = h(u_j)] \\ &= 1 + \sum_{j \neq i} \frac{1}{n} \\ &= 1 + \frac{n-1}{n} \leq 2\end{aligned}$$

$O(1)$
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What we wanted

H = the exhaustive set of all hash functions that map elements in the universe U to buckets 1 to n .

H contains a total of n^M hash functions.

The exhaustive set of all hash functions achieved our goal but was way too big, so let's pick **h** from a *smaller* hash family where

$$P[h(u_i) = h(u_j)] \leq 1/n$$

T
P[h(
dic

$$= 1 + \frac{1}{n} \leq 2$$

Universal Hash Family

A **hash family** is a fancy name for a set of hash functions.

A hash family \mathbf{H} is a **universal hash family** if, when h is chosen uniformly at random from \mathbf{H} ,

for all $u_i, u_j \in U$ with $u_i \neq u_j$,

$$P_{h \in H} [h(u_i) = h(u_j)] \leq \frac{1}{n}$$

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$$P_{h \in H} [h(u_i) = h(u_j)] \leq \frac{1}{n}$$

Then if we randomly choose h from a universal hash family \mathbf{H} , we'll be guaranteed that:

$$E[\# \text{ of items in } u_i\text{'s bucket}] \leq 2 = O(1)$$

Flashback of the Math

A hash family \mathbf{H} is a **universal hash family** if, when h is chosen uniformly at random from \mathbf{H} ,

$$\text{for all } u_i, u_j \in U \text{ with } u_i \neq u_j, \\ P_{h \in H} [h(u_i) = h(u_j)] \leq \frac{1}{n}$$

$$\begin{aligned} \mathbb{E}[\# \text{ of items in } u_i \text{ 's bucket}] &= \sum_{j=1}^n P[h(u_i) = h(u_j)] \\ &= P[h(u_i) = h(u_i)] + \sum_{j \neq i} P[h(u_i) = h(u_j)] \end{aligned}$$

This inequality is now what a universal hash family guarantees!

$$\begin{aligned} &= 1 + \sum_{j \neq i} P[h(u_i) = h(u_j)] \\ &\leq 1 + \sum_{j \neq i} \frac{1}{n} \end{aligned}$$

$$= 1 + \frac{n-1}{n} \leq 2$$

O(1)

This is what we wanted!

A Small Universal Hash Family?

Are there smaller ones universal hash families?

A Non-Example

$H = \{h_0, h_1\}$ where

$h_0 = \text{MOST_SIGNIFICANT_DIGIT}$

$h_1 = \text{LEAST_SIGNIFICANT_DIGIT}$

Why is this not a universal hash family?

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Why is this not a universal hash family?

$$P_{h \in H} [h(153) = h(173)] = 1 > \frac{1}{n}$$

There's a $\frac{1}{2}$ probability of choosing h_0 , and $h_0(153) = h_0(173) = \text{bucket 1}$

There's a $\frac{1}{2}$ probability of choosing h_1 , and $h_1(153) = h_1(173) = \text{bucket 3}$

Probability that a randomly chosen h from H collides 153 & 173 is 1!

An Example

Here is one of the more well-studied universal hash families:

Pick a prime $p \geq M$

Define $h_{a,b}(x) = ((ax + b) \bmod p) \bmod n$

$$H = \{ h_{a,b} : a \in \{1, \dots, p - 1\}, b \in \{0, \dots, p - 1\} \}$$

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Example: Suppose $n = 3$, and $p = 5$. Here's $h_{2,4}$:

$$h_{2,4}(1) = ((2 * 1 + 4) \bmod 5) \bmod 3 = (6 \bmod 5) \bmod 3 = 1 \bmod 3 = 1$$

$$h_{2,4}(4) = ((2 * 4 + 4) \bmod 5) \bmod 3 = (12 \bmod 5) \bmod 3 = 2 \bmod 3 = 2$$

$$h_{2,4}(3) = ((2 * 3 + 4) \bmod 5) \bmod 3 = (6 \bmod 5) \bmod 3 = 1 \bmod 3 = 1$$

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Define $h_{a,b}(x) = ((ax + b) \bmod p) \bmod n$

$$H = \{ h_{a,b} : a \in \{1, \dots, p - 1\}, b \in \{0, \dots, p - 1\} \}$$

To draw a hash function h from H :

Pick a random a
in $\{1, \dots, p - 1\}$.

&

Pick a random b
in $\{0, \dots, p - 1\}$.

An Example

Here is one of the more well-studied universal hash families:

To store $h_{a,b}$, you just need to store two numbers: **a** and **b**!

Since **a** and **b** are at most $p-1$, we need **$\sim 2 \cdot \log(p)$ bits**.

p is a prime that's close-ish to M , so this means the space needed =

$O(\log M)$

This is so much better than $O(M \log n)$!

$\{1, \dots, p-1\}$.

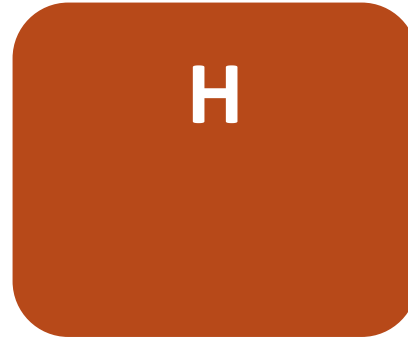
$\{0, \dots, p-1\}$.

Hash Tables

Putting everything together, what's the scheme?

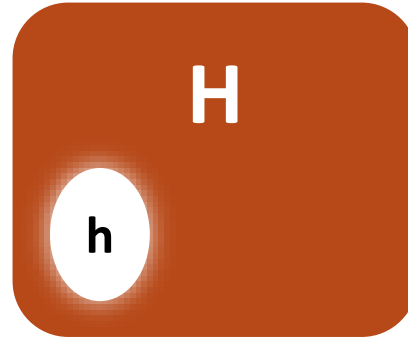
The Whole Scheme

You choose your set of hash functions **H**, a universal hash family like $H = \text{mod } p \text{ mod } n$.



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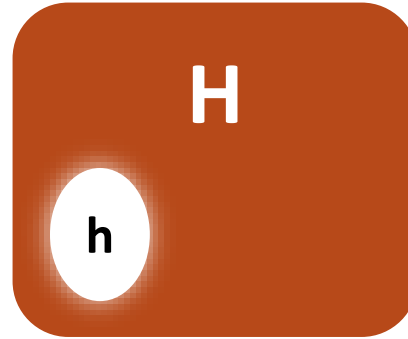
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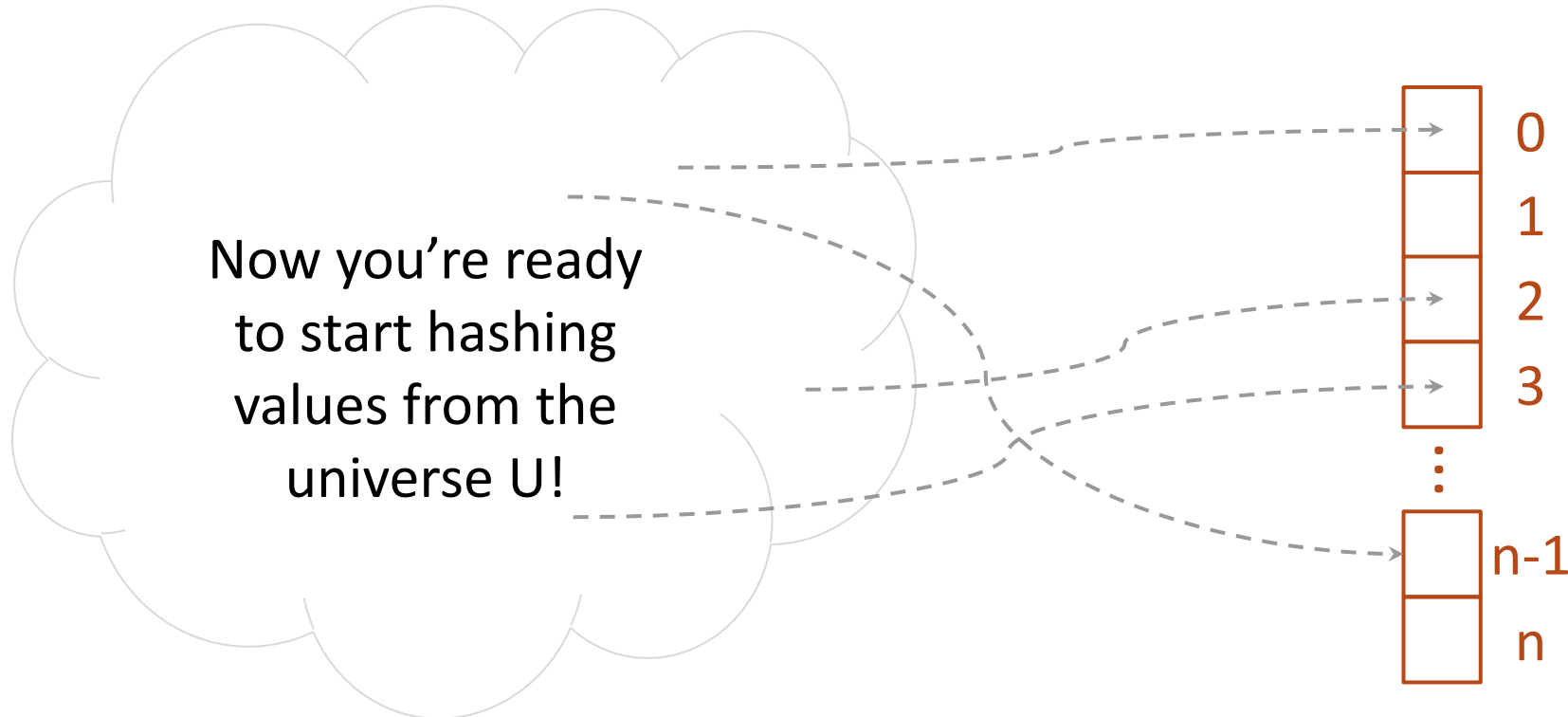
When the client initializes a hash table, randomly pick a hash function \mathbf{h} from \mathbf{H} to use in the hash table to hash the items.

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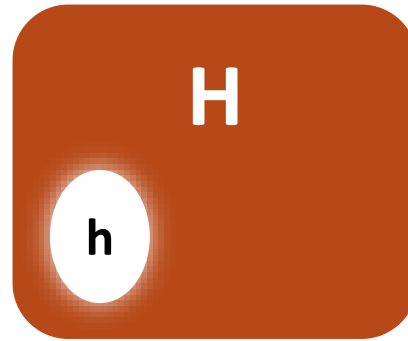


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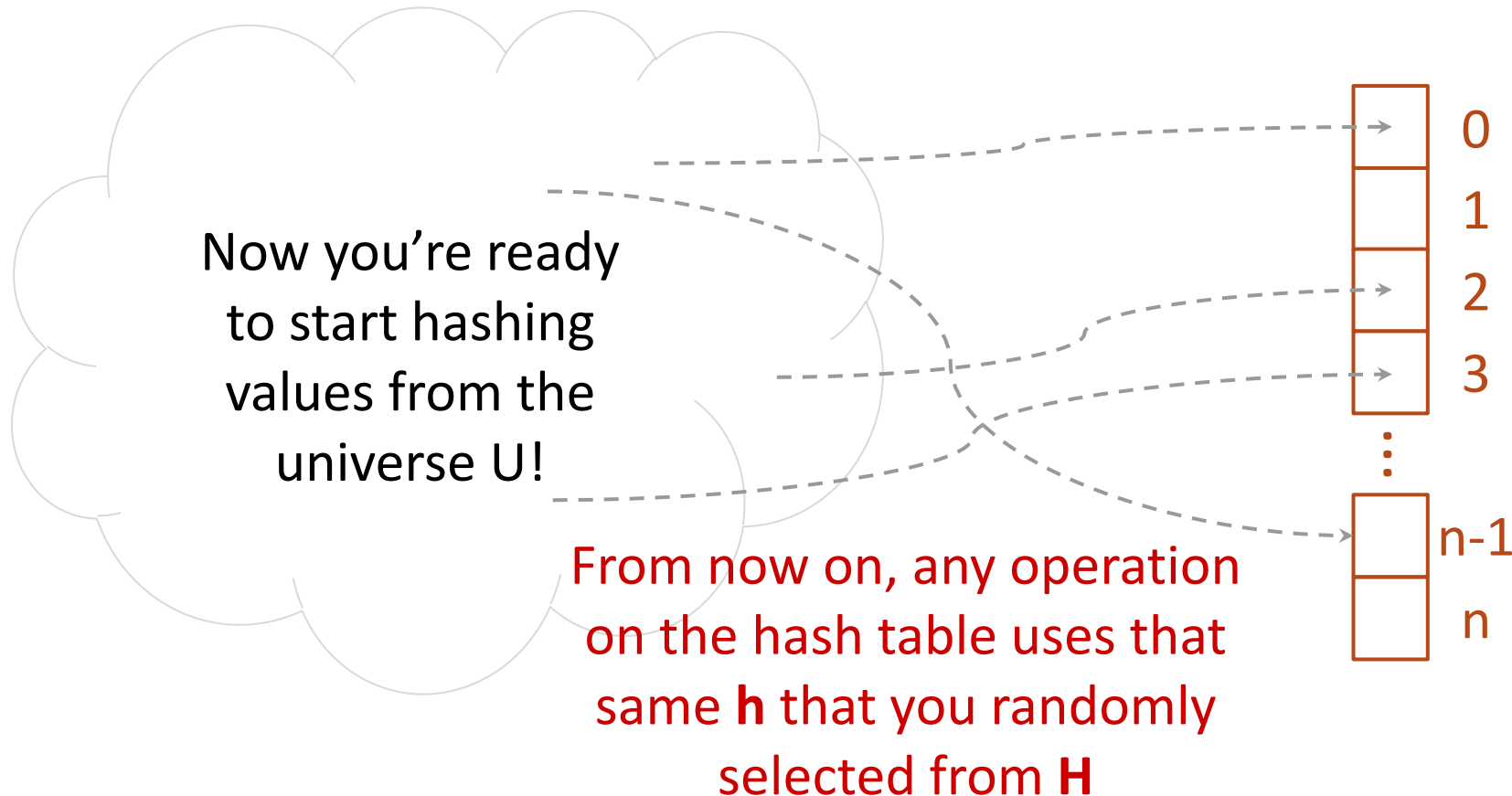


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We can now expect that these buckets will be pretty balanced

Hash Table: Motivation

| OPERATION | SORTED ARRAY | UNSORTED LINKED LIST | HASH TABLES (HOPEFULLY) |
|-----------|--------------|----------------------|-------------------------|
| SEARCH | $O(\log(n))$ | $O(n)$ | $O(1)$ |
| DELETE | $O(n)$ | $O(n)$ | $O(1)$ |
| INSERT | $O(n)$ | $O(1)$ | $O(1)$ |

*** Assuming we implement it cleverly with a “good” hash function**

Acknowledgement

- Stanford University

Thank You