## Randomized Algorithms II

Summer 2017 • Lecture 5

#### A Few Notes

#### Homework 2

Due tomorrow night at 11:59 p.m. on Gradescope.

#### Homework 3

Released tomorrow night.

Due Friday 7/21 at 11:59 p.m. on Gradescope.

## **Outline for Today**

#### More randomized algorithms!

**Hashing Basics** 

Universal Hash Functions

What's the Source of the Randomness?

# Hashing Basics

## Randomized Algorithms

A randomized algorithm is an algorithm that incorporates randomness as part of its operation.

Often aim for properties like ...

Good average-case behavior

Getting exact answers with high probability

Getting answers that are close to the right answer

#### **Data Structures**

	Sorted linked lists	Sorted arrays	Balanced BSTs
Search	O(n) expected & worst- case	O(log n) expected & worst- case	O(log n) expected & worst- case O(n) worst-case for generic BSTs
Insert/ Delete	O(n) expected & worst- case without a pointer to the element	O(n) expected & worst- case	O(log n) expected & worst case

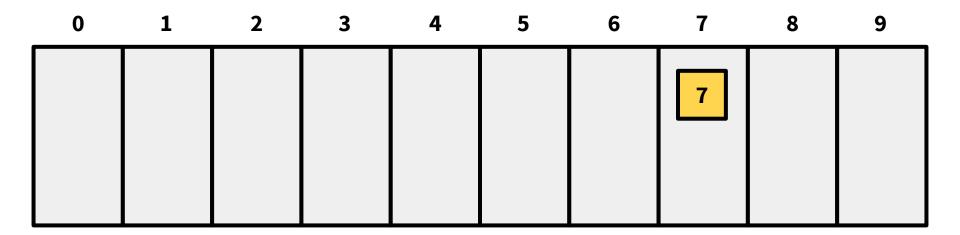
#### **Data Structures**

	Sorted linked lists	Sorted arrays	Balanced BSTs	Hash tables
Search	O(n) expected & worst- case	O(log n) expected & worst- case	O(log n) expected & worst- case O(n) worst-case for generic BSTs	O(1) expected O(n) worst-case
Insert/ Delete	O(n) expected & worst- case without a pointer to the element	O(n) expected & worst- case	O(log n) expected & worst case	O(1) expected O(n) worst-case without a pointer to the element

How might we get O(1)-time? Try direct addressing!

One type of item per address.

insert(7)

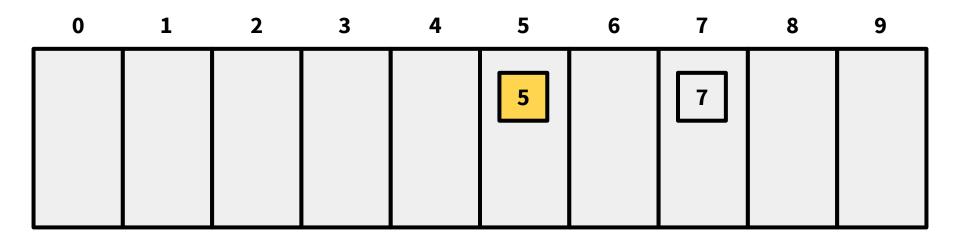


How might we get O(1)-time? Try direct addressing!

One type of item per address.

insert(7)

insert(5)



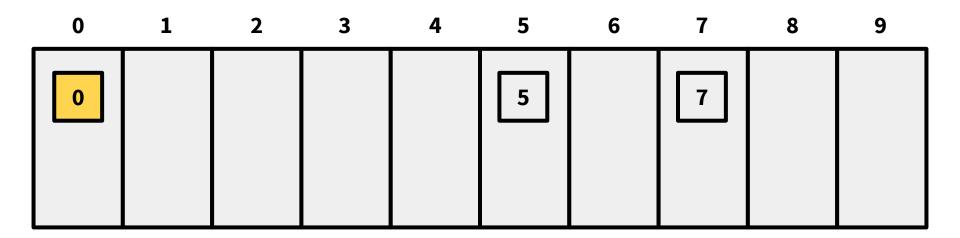
How might we get O(1)-time? Try direct addressing!

One type of item per address.

```
insert(7)
```

insert(5)

insert(0)



How might we get O(1)-time? Try direct addressing!

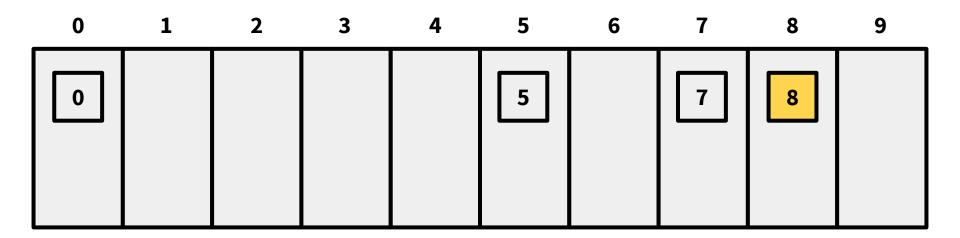
One type of item per address.

```
insert(7)
```

insert(5)

insert(0)

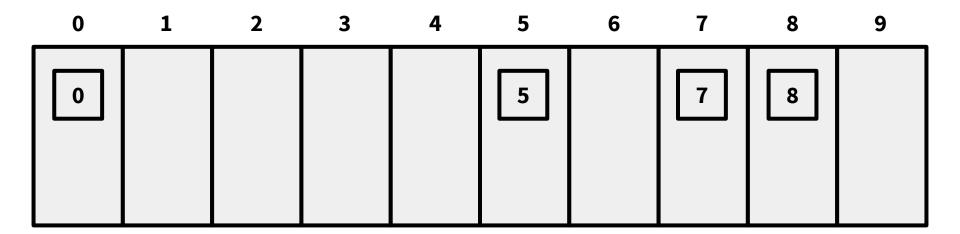
insert(8)



How might we get O(1)-time? Try direct addressing!

One type of item per address.

```
insert(7) search(7)
insert(5) search(2)
insert(0)
insert(8)
```



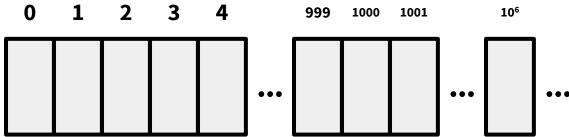
How might we get O(1)-time? Try direct addressing!

What's the issue with this approach?

How might we get O(1)-time? Try direct addressing!

What's the issue with this approach? 🧐

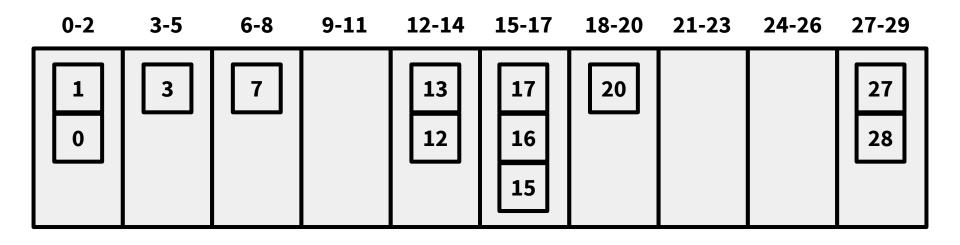
Similar to counting\_sort and bucket\_sort (for  $k \le num\_buckets$ ), if the set of items being inserted/deleted (e.g.  $\{0, 1, 2, ..., 999, 1000, ..., 10^6, ...\}$ ) is large, then the sheer space required to maintain this data structure becomes an issue.



How might we get O(1)-time? Try direct addressing!

Can we fix this issue by assigning multiple types of item per address, like case (2) of bucket\_sort?

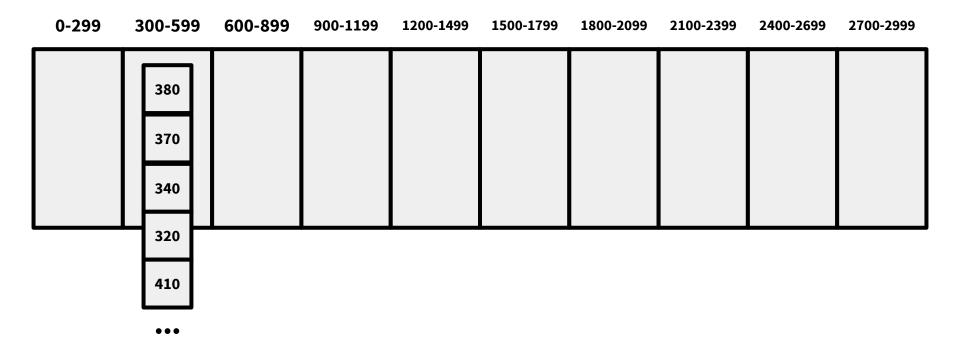
Sometimes, this binning approach is useful. search(12) still runs pretty fast.



How might we get O(1)-time? Try direct addressing!

Can we fix this issue by assigning multiple types of item per address, like case (2) of bucket\_sort?

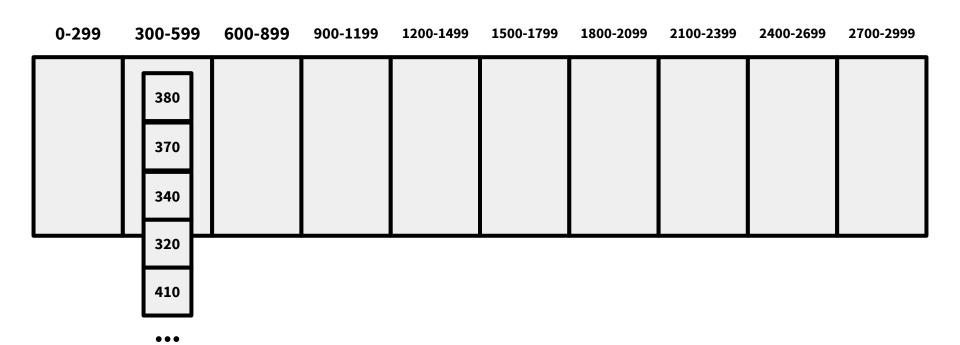
Other times, it causes an issue. search (432) is slow.



This is an example of a hash table.

Albeit one with a basic bucketing scheme.

Can we do better?



There exists a universe U of keys, size |U|.

|U| is really big.

What is |U| if U is the set of ASCII strings of length 16?



There exists a universe U of keys, size |U|.

|U| is really big.

What is |U| if U is the set of ASCII strings of length 16?  $(U) = 128^{16}$ .

There exists a universe U of keys, size |U|.

|U| is really big.

What is |U| if U is the set of ASCII strings of length 16? (9)  $|U| = 128^{16}$ .



We hash the keys to n buckets.

|U| >>> n; i.e. |U| is a lot bigger than n.

We don't know which of the |U| possible keys we'll need to store; Unless otherwise stated, let's assume ≤ n.

There exists a universe U of keys, size |U|.

|U| is really big.

What is |U| if U is the set of ASCII strings of length 16? (9)  $|U| = 128^{16}$ .

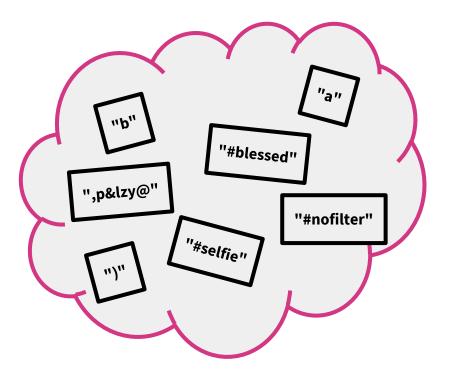
We hash the keys to n buckets.

|U| >>> n; i.e. |U| is a lot bigger than n.

We don't know which of the |U| possible keys we'll need to store; Unless otherwise stated, let's assume ≤ n.

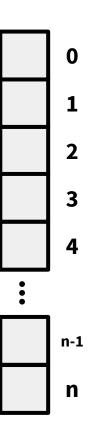
There's a hash function h:  $U \rightarrow \{1, ..., n\}$  that maps keys to buckets.

### An Example

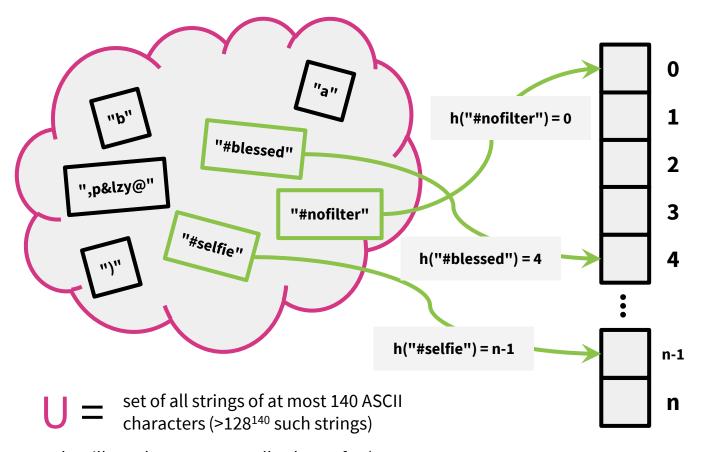


set of all strings of at most 140 ASCII characters (>128<sup>140</sup> such strings)

And we'll need to store a small subset of U (say, the ones that might be trending hashtags on Twitter); we're assuming the number of hashtags ≤ n, the number of buckets.



#### An Example



And we'll need to store a small subset of U (say, the ones that might be trending hashtags on Twitter); we're assuming the number of hashtags ≤ n, the number of buckets.

List of n buckets.

Each bucket stores an unsorted linked list.

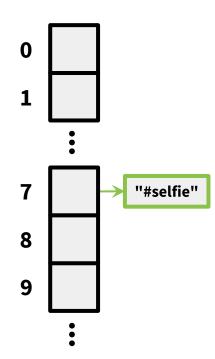
insert in O(1) since it's unsorted; search in O(n).

h:  $U \rightarrow \{1, ..., n\}$  can be any function

For concreteness, suppose it's length.

Suppose we insert a bunch of keys and then search.

insert("#selfie")



List of n buckets.

Each bucket stores an unsorted linked list.

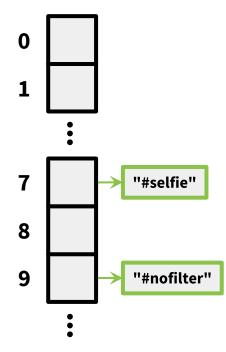
insert in O(1) since it's unsorted; search in O(n).

h:  $U \rightarrow \{1, ..., n\}$  can be any function

For concreteness, suppose it's length.

Suppose we insert a bunch of keys and then search.

```
insert("#selfie")
insert("#nofilter")
```



List of n buckets.

Each bucket stores an unsorted linked list.

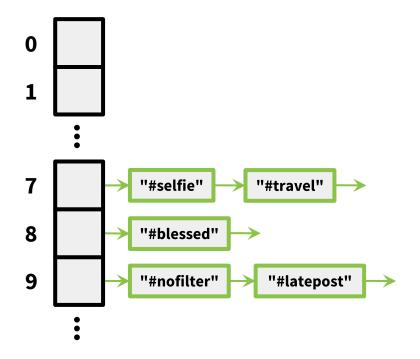
insert in O(1) since it's unsorted; search in O(n).

h:  $U \rightarrow \{1, ..., n\}$  can be any function

For concreteness, suppose it's length.

Suppose we insert a bunch of keys and then search.

```
insert("#selfie")
insert("#nofilter")
insert("#blessed")
insert("#travel")
insert("#latepost")
search("#travel")
Scans through
all elements in
bucket
h("#travel")
```



Is choosing h:  $U \rightarrow \{1, ..., n\}$  to be length a good idea? (9)



Is choosing h:  $U \rightarrow \{1, ..., n\}$  to be length a good idea? (9)

Not really. In fact, it's quite terrible since there are a lot of hashtags that share the same lengths.

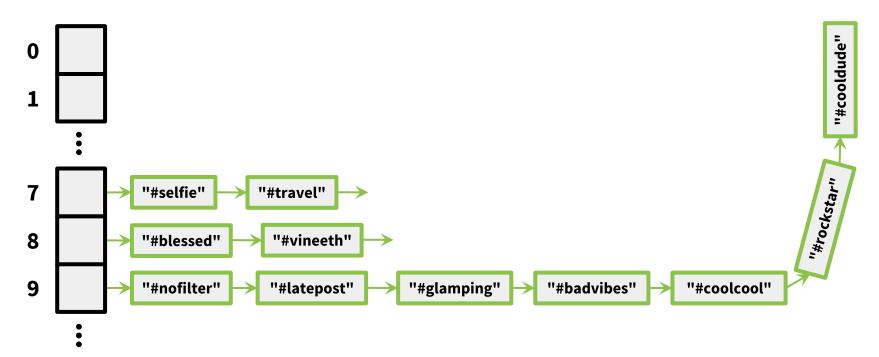
Is choosing h:  $U \rightarrow \{1, ..., n\}$  to be length a good idea? (9)



Not really. In fact, it's quite terrible since there are a lot of hashtags that share the same lengths.

So how do we choose a better h?

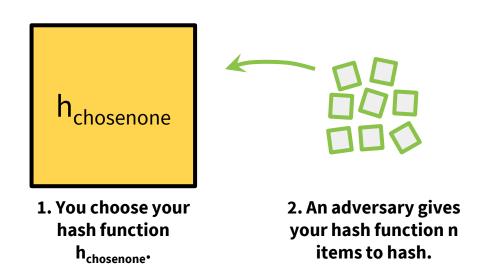
The items need to be spread out in the buckets.



(1) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have size O(1) after hashing any n items?

(1) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have size O(1) after hashing any n items? after an adversary chooses n items to hash?

(1) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have size O(1) after hashing any n items? after an adversary chooses n items to hash?



Is it possible to construct h<sub>chosenone</sub> such that you're guaranteed that all buckets will have size O(1)? This would be ideal.

(1) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have size O(1) after hashing any n items? after an adversary chooses n items to hash?

You probably couldn't think of how. Why not? It's impossible!

(1) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have size O(1) after hashing any n items? after an adversary chooses n items to hash?

You probably couldn't think of how. Why not? It's impossible!

 $h_{\text{chosenone}}$  is defined from a domain of  $\left|U\right|$  items to a range of n buckets.

(1) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have size O(1) after hashing any n items? after an adversary chooses n items to hash?

You probably couldn't think of how. Why not? It's impossible!

 $h_{chosenone}$  is defined from a domain of |U| items to a range of n buckets. By the pigeonhole principle, at least one of the buckets receives at least |U|/n items.

(1) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have size O(1) after hashing any n items? after an adversary chooses n items to hash?

You probably couldn't think of how. Why not? It's impossible!

 $h_{chosenone}$  is defined from a domain of |U| items to a range of n buckets. By the pigeonhole principle, at least one of the buckets receives at least |U|/n items. Recall that |U| >> n, so |U|/n > n; therefore at least one of the buckets receives at least n items.

(1) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have size O(1) after hashing any n items? after an adversary chooses n items to hash?

You probably couldn't think of how. Why not? It's impossible!

 $h_{chosenone}$  is defined from a domain of |U| items to a range of n buckets. By the pigeonhole principle, at least one of the buckets receives at least |U|/n items. Recall that |U| >> n, so |U|/n > n; therefore at least one of the buckets receives at least n items.

Let's call the set of items that get hashed to this bucket  $U_{bigbucket}(h_{chosenone})$  where  $U_{bigbucket} \subset U$ . The adversary could choose to hash n items from  $U_{bigbucket}$ . This is a valid set of n items, and results in one bucket with all n items, by construction. Therefore, **(1)** is impossible.

(1) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have size O(1) after hashing any n items? after an adversary chooses n items to hash?

You probably couldn't think of how. Why not? It's impossible!

 $h_{\text{chosenone}}$  is defined from a domain of  $\left|U\right|$  items to a range of n buckets. By the pigeonhole principle, at least one of the buckets receives at least |U|/n items. Recall that |U| >> n, so |U|/n > n; therefore at least one of the buckets receives at least n items.

Notation indicating  $U_{bigbucket}$  is a function of  $h_{chosenone}$ 



Let's call the set of items that get hashed to this bucket U<sub>bigbucket</sub>(h<sub>chosenone</sub>) where  $U_{\text{bigbucket}} \subset U$ . The adversary could choose to hash n items from U<sub>bigbucket</sub>. This is a valid set of n items, and results in one bucket with all n items, by construction. Therefore, (1) is impossible.

(1) Can we design a single h<sub>chosenone</sub>: U → {1, ..., n} such that all buckets will have size O(1) after hashing any n items?

Impossible

(2) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have **expected** size O(1) after hashing any n items?

(1) Can we design a single h<sub>chosenone</sub>: U → {1, ..., n} such that all buckets will have size O(1) after hashing any n items?

Impossible

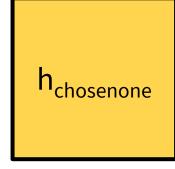
(2) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have **expected** size O(1) after hashing any n items? after an adversary chooses n items to hash?

(1) Can we design a single h<sub>chosenone</sub>: U → {1, ..., n} such that all buckets will have size O(1) after hashing any n items?



(2) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have **expected** size O(1) after hashing any n items? after an adversary chooses n items to hash?

1. You choose your hash function h<sub>chosenone</sub>.





2. An adversary gives your hash function n items to hash.

Is it possible to construct h<sub>chosenone</sub> such that you're guaranteed that all buckets will have expected size O(1)? This would be good.

(1) Can we design a single h<sub>chosenone</sub>: U → {1, ..., n} such that all buckets will have size O(1) after hashing any n items?



(2) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have **expected** size O(1) after hashing any n items? after an adversary chooses n items to hash?

Can you think of such an h<sub>chosenone</sub>?

Probably not. This is the same question as **(1)**! Since the adversary is choosing the n items, there's no randomness anywhere in the process. As a result, the **expected** size of a bucket is trivially just the size.

(1) Can we design a single  $h_{chosenone}$ :  $U \rightarrow \{1, ..., n\}$  such that all buckets will have size O(1) after hashing any n items?

Impossible!

(2) Can we design a single h<sub>chosenone</sub>: U → {1, ..., n} such that all buckets will have **expected** size O(1) after hashing any n items?

Impossible as stated. But if we introduce randomness...

In order for all buckets to have **expected** size O(1) after hashing any n items, we need to introduce randomness.

(1) Can we design a single h<sub>chosenone</sub>: U → {1, ..., n} such that all buckets will have size O(1) after hashing any n items?

Impossible

(2) Can we design a single h<sub>chosenone</sub>: U → {1, ..., n} such that all buckets will have **expected** size O(1) after hashing any n items?

Impossible as stated. But if we introduce randomness...

In order for all buckets to have **expected** size O(1) after hashing any n items, we need to introduce randomness.

Where? Well there's only one option ... in our choice of hash function.

We will randomly choose h from a large set of hash functions! (There won't be an h to rule them all).

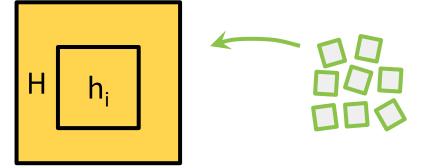


(3) Can we design a set  $H = \{h_1, ..., h_k\}$  where  $h: U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, all buckets will have **expected** size O(1) after hashing any n items?

(3) Can we design a set  $H = \{h_1, ..., h_k\}$  where  $h: U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, all buckets will have **expected** size O(1) after hashing any h items? after an adversary chooses h items to hash?

(3) Can we design a set  $H = \{h_1, ..., h_k\}$  where  $h: U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, all buckets will have **expected** size O(1) after hashing any h items? after an adversary chooses h items to hash?

1. You choose your set of hash functions H.

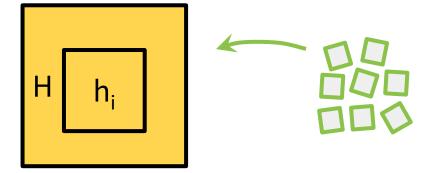


2. An adversary gives your hash function n items to hash.

3. You randomly pick a hash function h<sub>i</sub> from H to hash the n items.

(3) Can we design a set  $H = \{h_1, ..., h_k\}$  where  $h: U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, all buckets will have **expected** size O(1) after hashing any h items? after an adversary chooses h items to hash?

1. You choose your set of hash functions H.



2. An adversary gives your hash function n items to hash.

3. You randomly pick a hash function h<sub>i</sub> from H to hash the n items.

Is it possible to construct H such that you're guaranteed that all buckets will have expected size O(1)? This would be good.

(3) Can we design a set  $H = \{h_1, ..., h_k\}$  where  $h: U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, all buckets will have **expected** size O(1) after hashing any h items? after an adversary chooses h items to hash?

Yes! But it's not very useful.

(3) Can we design a set  $H = \{h_1, ..., h_k\}$  where  $h: U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, all buckets will have **expected** size O(1) after hashing any h items? after an adversary chooses h items to hash?

Yes! But it's not very useful.

Let H be the set of n hash functions where h<sub>i</sub> hashes all keys in the entire universe to bucket i.

(3) Can we design a set  $H = \{h_1, ..., h_k\}$  where  $h: U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, all buckets will have **expected** size O(1) after hashing any h items? after an adversary chooses h items to hash?

Yes! But it's not very useful.

Let H be the set of n hash functions where h<sub>i</sub> hashes all keys in the entire universe to bucket i. With probability 1/n, a bucket b will have all the keys that the adversary chose get hashed to it. Otherwise, the bucket will be empty.

(3) Can we design a set  $H = \{h_1, ..., h_k\}$  where  $h: U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, all buckets will have **expected** size O(1) after hashing any h items? after an adversary chooses h items to hash?

Yes! But it's not very useful.

Let H be the set of n hash functions where h<sub>i</sub> hashes all keys in the entire universe to bucket i. With probability 1/n, a bucket b will have all the keys that the adversary chose get hashed to it. Otherwise, the bucket will be empty.

E[size\_of(b)] = P(all keys hashed to it) 
$$\cdot$$
 n + P(0 keys hashed to it)  $\cdot$  0  
=  $(1/n) \cdot$  n  
= 1

But P(lots of keys get hashed to one bucket) = 1.

(3) Can we design a set  $H = \{h_1, ..., h_k\}$  where  $h: U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, all buckets will have **expected** size O(1) after hashing any h items? after an adversary chooses h items to hash?

Yes! But it's not very useful.

Let H be the set of n hash functions where h<sub>i</sub> hashes all keys in the entire universe to bucket i. With probability 1/n, a bucket b will have all the keys that the adversary chose get hashed to it. Otherwise, the bucket will be empty.

E[size\_of(b)] = P(all keys hashed to it) 
$$\cdot$$
 n + P(0 keys hashed to it)  $\cdot$  0  
=  $(1/n) \cdot$  n  
= 1

But P(lots of keys get hashed to one bucket) = 1.

This is not good. Maybe we should be using a different metric.

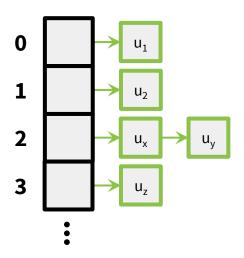
- (3) Can we design a set  $H = \{h_1, ..., h_k\}$  where  $h: U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, all buckets will have **expected** size O(1) after hashing any n items? after an adversary chooses n items to hash?
- **(4)** Can we design a set  $H = \{h_1, ..., h_k\}$  where h:  $U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, after an adversary chooses n items  $\{u_1, ..., u_n\}$  to hash, the **expected** number of items in  $u_x$ 's bucket is O(1)?

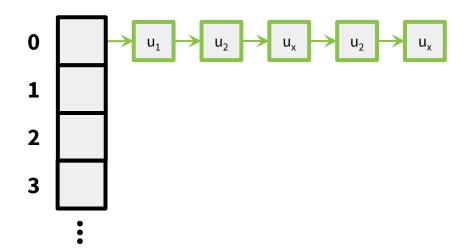
- (3) Can we design a set  $H = \{h_1, ..., h_k\}$  where  $h: U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, all buckets will have **expected** size O(1) after hashing any n items? after an adversary chooses n items to hash?
- **(4)** Can we design a set  $H = \{h_1, ..., h_k\}$  where  $h: U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, after an adversary chooses n items  $\{u_1, ..., u_n\}$  to hash, the **expected** number of items in  $u_x$ 's bucket is O(1)?

As an analogy for the difference between (3) and (4), consider the "small classes illusion." Suppose a university offers 10 classes, 9 of which have 1 person in them and the last of which has 500 persons in them. Using reasoning from (3), the university might tout average class sizes of ~50, when in reality, it should report much class sizes experienced by the average student, as in (4).

**(4)** Can we design a set  $H = \{h_1, ..., h_k\}$  where  $h: U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, after an adversary chooses n items  $\{u_1, ..., u_n\}$  to hash, the **expected** number of items in  $u_x$ 's bucket is O(1)?

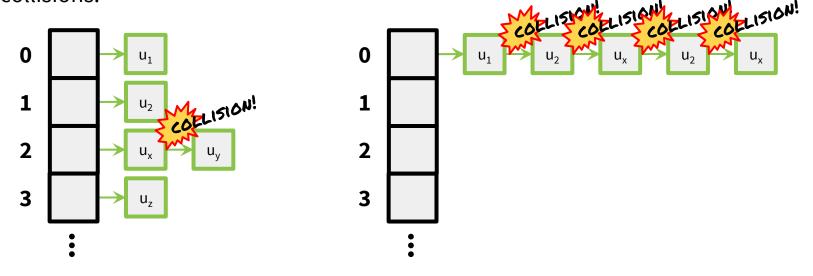
We can think of this statement in terms of minimizing the expected number of collisions.





**(4)** Can we design a set  $H = \{h_1, ..., h_k\}$  where h:  $U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, after an adversary chooses n items  $\{u_1, ..., u_n\}$  to hash, the **expected** number of items in  $u_x$ 's bucket is O(1)?

We can think of this statement in terms of minimizing the expected number of collisions.



**(4)** Can we design a set  $H = \{h_1, ..., h_k\}$  where h:  $U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, after an adversary chooses n items  $\{u_1, ..., u_n\}$  to hash, the **expected** number of items in  $u_x$ 's bucket is O(1)?

Yes! This time it's possible.

	$h_1$	h <sub>2</sub>	h <sub>3</sub>	h <sub>4</sub>	h <sub>5</sub>	h <sub>6</sub>	h <sub>7</sub>	h <sub>8</sub>
"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 buckets i.e. h<sub>8</sub> hashes "b" to bucket 1.

**(4)** Can we design a set  $H = \{h_1, ..., h_k\}$  where h:  $U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, after an adversary chooses n items  $\{u_1, ..., u_n\}$  to hash, the **expected** number of items in  $u_x$ 's bucket is O(1)?

Yes! This time it's possible.

Let H be the exhaustive set of all hash functions that map elements in the universe U to buckets 1 to n, which has size  $|H| = n^{|U|}$ .

	$h_1$	h <sub>2</sub>	h <sub>3</sub>	$h_4$	h <sub>5</sub>	h <sub>6</sub>	h <sub>7</sub>	h <sub>8</sub>
"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 buckets i.e. h<sub>8</sub> hashes "b" to bucket 1.

**(4)** Can we design a set  $H = \{h_1, ..., h_k\}$  where h:  $U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, after an adversary chooses n items  $\{u_1, ..., u_n\}$  to hash, the **expected** number of items in  $u_x$ 's bucket is O(1)?

Yes! This time it's possible.

Let H be the exhaustive set of all hash functions that map elements in the universe U to buckets 1 to n, which has size  $|H| = n^{|U|}$ .

e.g. Suppose U = {"a", "b", "c"} and n = 2 (there are 2 buckets). H would be a set of 8 hash functions. One h would map "a", "b", and "c" all to bucket 0. Another h would map "a" and "b" to bucket 0 and "c" to bucket 1. etc. etc.

,	h <sub>1</sub>	h <sub>2</sub>	h <sub>3</sub>	h <sub>4</sub>	h <sub>5</sub>	h <sub>6</sub>	h <sub>7</sub>	h <sub>8</sub>	
"a"	0	0	0	0	1	1	1	1	
"b"	0	0	1	1	0	0	1	1	N. N.
"c"	0	1	0	1	0	1	0	1	

The 0's and 1's represent the buckets i.e. h<sub>8</sub> hashes "b" to bucket 1.

**(4)** Can we design a set  $H = \{h_1, ..., h_k\}$  where h:  $U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, after an adversary chooses n items  $\{u_1, ..., u_n\}$  to hash, the **expected** number of items in  $u_x$ 's bucket is O(1)?

E[number of items in  $u_x$ 's bucket] =  $\sum_{y=1}^{y=1} P[h(u_x) = h(u_y)]$ 

E[number of items in 
$$u_x$$
's bucket] =  $\sum_{y=1}^{\infty} P[h(u_x) = h(u_y)]$ 

$$= 1 + \sum_{y \neq x} P[h(u_x) = h(u_y)]$$

E[number of items in 
$$u_x$$
's bucket] =  $\sum_{y=1}^{y=1} P[h(u_x) = h(u_y)]$ 

$$= 1 + \sum_{y \neq x} P[h(u_x) = h(u_y)]$$
You will prove this is the case for the exhaustive set H in your Homework.

E[number of items in 
$$u_x$$
's bucket] =  $\sum_{y=1}^{\infty} P[h(u_x) = h(u_y)]$ 

$$= 1 + \sum_{y \neq x} P[h(u_x) = h(u_y)]$$

$$= 1 + \sum_{y \neq x} 1/n$$
You will prove this is the case for the exhaustive set H in your Homework.

$$= 1 + (n-1/n)$$

E[number of items in 
$$u_x$$
's bucket] =  $\sum_{y=1}^{\infty} P[h(u_x) = h(u_y)]$ 

$$= 1 + \sum_{y \neq x} P[h(u_x) = h(u_y)]$$
You will prove this is the case for the exhaustive set H in your Homework.

$$= 1 + (n-1/n)$$

#### The Good News

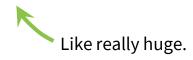
**(4)** Can we design a set  $H = \{h_1, ..., h_k\}$  where h:  $U \rightarrow \{1, ..., n\}$ , such that if we chose a random h in H, after an adversary chooses n items  $\{u_1, ..., u_n\}$  to hash, the **expected** number of items in  $u_x$ 's bucket is O(1)?

Yes! This is great news! It means that we can choose H to be the exhaustive set of all hash functions, and the insert, delete, search operations on any n elements will have an expected runtime of O(1) per operation.

#### The Bad News

The exhaustive set of all hash functions is HUGEEE!!!

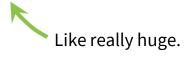
How many bits would it take to write down the name of one of the  $n^{|U|}$  hash functions in this H? (



#### The Bad News

#### The exhaustive set of all hash functions is HUGEEE!!!

How many bits would it take to write down the name of one of the  $n^{|U|}$  hash functions in this H? (P) log  $n^{|U|} = |U|$  log n.



To see why, consider how much memory it would take to write down the name of one of the 8 hash functions from earlier. You could assign  $h_1$  the id 000,  $h_2$  the id 001, etc. So 8 hash functions requires log 8 = 3 bits to write down.

|U| log n bits is enough to do direct addressing!

# H Is Too Big

How can we fix this issue of the size of H?

## 3 Min Break

## **Universal Hash Functions**

## H Is Too Big

#### How can we fix this issue of the size of H?

Pick from a smaller set H, that still guarantees (4).

Recall the bound that allowed us to achieve this guarantee:

E[number of items in 
$$u_x$$
's bucket] =  $\sum_{y=1}^{\infty} P[h(u_x) = h(u_y)]$   
=  $1 + \sum_{y \neq x} P[h(u_x) = h(u_y)]$   
This step!  
=  $1 + \sum_{y \neq x} 1/n$   
=  $1 + (n-1/n)$   
 $\leq 2$ 

## **Universal Hash Family**

This bound is so important, there's a special name for sets H that satisfy it.

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

# **Universal Hash Family**

This bound is so important, there's a special name for sets H that satisfy it.

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

A **universal hash family** describes a set of hash functions that satisfy the bound:  $P_{h \in H}[h(u_x) = h(u_y)] \le 1/n$ .

The exhaustive set of hash functions is an example of a universal hash family but, as discussed previously, it's too big to be practical.

# A Smaller Universal Hash Family

Identifying new smaller universal hash families is an active field of research in computer science.

One of the more well-studied universal hash families:

```
To hash an integer x in \{0, ..., |U|-1\} to a bucket \{1, ..., n\}:
```

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

for some prime 
$$p \ge |U|$$
 and  $a \in \{1, ..., p - 1\}$  and  $b \in \{0, ..., p - 1\}$ 

To select an  $h_{a,b}$  from this family:



1. Determine |U|.



2. Find the smallest prime p ≥ |U|.



3. Let a be a random number in {1, ..., p - 1}.

b

3. Let **b** be a random number in {0, ..., p - 1}.

### How Small Is This H?

There are p-1 choices for a and p choices for b, so  $|H| = p(p-1) = O(p^2) = O(|U|^2)$ .

That's much better than  $n^{|U|}$ .

The space need to store h is  $\log |U|^2 = O(\log |U|) \ll O(|U|\log n)$ .



1. Determine |U|.



2. Find the smallest prime p ≥ |U|.



3. Let a be a random number in {1, ..., p - 1}.

b

3. Let **b** be a random number in {0, ..., p - 1}.

# **Another Universal Hash Family**

Another of the more well-studied universal hash families (using matrix multiplication!):

To hash a u-bit string x (i.e. bit string of length u) to a bucket  $\{1, ..., n\}$  (i.e. bit string of length b = log(n)):

$$h_A(x) = Ax$$

for some  $b \times u$  matrix A of 0's and 1's, using binary (modulo 2) arithmetic.

To select an h<sub>A</sub> from this family:



1. Determine |U|.



2. u = log(|U|).



3. b = log(n).

A

3. Let A be a b × u random matrix of 0's and 1's.

### How Small Is This H?

How many possible binary matrices of size  $b \times u$  for A?

$$2^{ub} = O(|U|^{\log(n)}).$$

That's much better than  $n^{|U|}$ , but larger than the other universal hash family  $O(|U|^2)$ .





2. u = log(|U|).



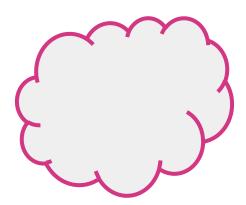
3. b = log(n).

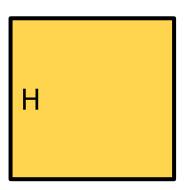


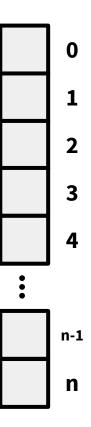
3. Let A be a b × u random matrix of 0's and 1's.

### **Hash Tables**

Let's say you wanted to implement a hash table ... 1. You choose your set of hash functions H, likely a universal hash family like H = mod p mod n.

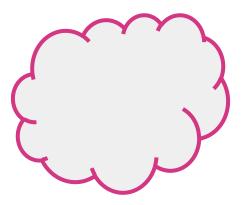






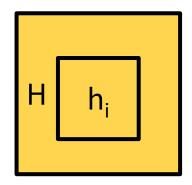
## **Hash Tables**

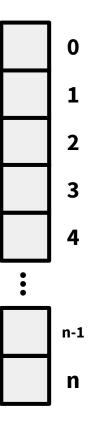
Let's say you wanted to implement a hash table ...



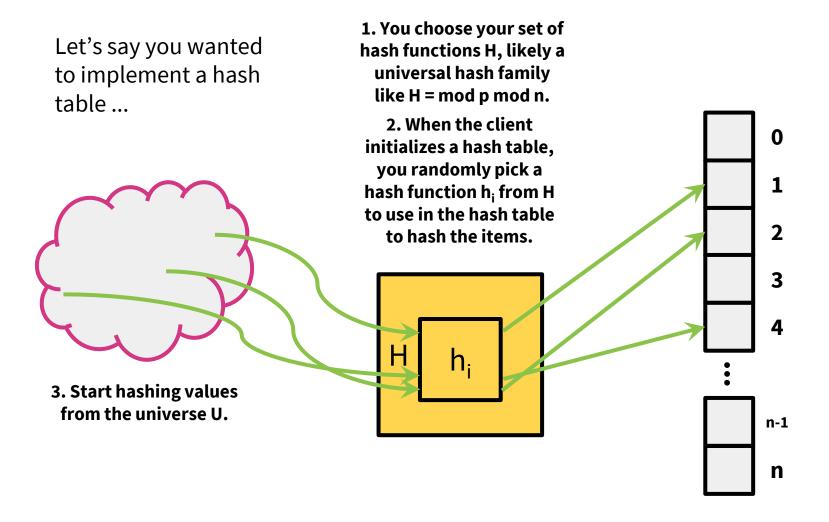
 You choose your set of hash functions H, likely a universal hash family like H = mod p mod n.

2. When the client initializes a hash table, you randomly pick a hash function h<sub>i</sub> from H to use in the hash table to hash the items.





## **Hash Tables**



As was the case in quicksort, we want the average-case runtime for a specific input to be low.

This is why it was important for us to select our pivot randomly as opposed to select, say, the first element in the sublist in quicksort.

As was the case in quicksort, we want the average-case runtime for a specific input to be low.

This is why it was important for us to select our pivot randomly as opposed to select, say, the first element in the sublist in quicksort.

If we selected the pivot as the first element, then the expected runtime over all of the inputs would be low, but for certain inputs (namely, the one in reverse order), runtime would be guaranteed to be high.

As was the case in quicksort, we want the average-case runtime for a specific input to be low.

This is why it was important for us to select our pivot randomly as opposed to select, say, the first element in the sublist in quicksort.

If we selected the pivot as the first element, then the expected runtime over all of the inputs would be low, but for certain inputs (namely, the one in reverse order), runtime would be guaranteed to be high.

Instead, since our algorithm supplied the randomness, for any specific input (even the one in reverse order), the expected runtime is low.

As was the case in quicksort, we want the average-case runtime for a specific input to be low.

This is why it was important for us to select our pivot randomly as opposed to select, say, the first element in the sublist in quicksort.

If we selected the pivot as the first element, then the expected runtime over all of the inputs would be low, but for certain inputs (namely, the one in reverse order), runtime would be guaranteed to be high.

Instead, since our algorithm supplied the randomness, for any specific input (even the one in reverse order), the expected runtime is low.

Note this does not say anything the worst-case runtime, which you can think of as the case in which the adversary controls the randomness we're using. This remains the same even though we introduced randomness into our algorithm.

As was the case in quicksort, we want the average-case runtime for a specific input to be low.

This is why it was important for us to select our pivot randomly as opposed to select, say, the first element in the sublist in quicksort.

If we selected the pivot as the first element, then the expected runtime over all of the inputs would be low, but for certain inputs (namely, the one in reverse order), runtime would be guaranteed to be high.

Instead, since our algorithm supplied the randomness, for any specific input (even the one in reverse order), the expected runtime is low.

Note this does not say anything the worst-case runtime, which you can think of as the case in which the adversary controls the randomness we're using. This remains the same even though we introduced randomness into our algorithm.

Same thing here with hash tables.