### **Design and Analysis of Algorithms**

CSE 5311

Lecture 8 Hashing II

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## A weakness of hashing

**Problem:** For any hash function h, a set of keys exists that can cause the average access time of a hash table to skyrocket.

• An adversary can pick all keys from  $\{k \in U : h(k) = i\}$  for some slot i.

**IDEA:** Choose the hash function at random, independently of the keys.

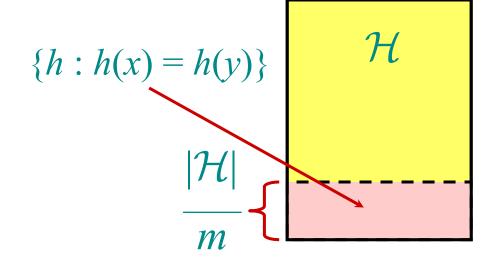
• Even if an adversary can see your code, he or she cannot find a bad set of keys, since he or she doesn't know exactly which hash function will be chosen.



## Universal hashing

**Definition.** Let U be a universe of keys, and let  $\mathcal{H}$  be a finite collection of hash functions, each mapping U to  $\{0, 1, ..., m-1\}$ . We say  $\mathcal{H}$  is *universal* if for all  $x, y \in U$ , where  $x \neq y$ , we have  $|\{h \in \mathcal{H} : h(x) = h(y)\}| = |\mathcal{H}|/m$ .

That is, the chance of a collision between x and y is 1/m if we choose h randomly from  $\mathcal{H}$ .





## Universality is good

**Theorem.** Let h be a hash function chosen (uniformly) at random from a universal set  $\mathcal{H}$  of hash functions. Suppose h is used to hash n arbitrary keys into the m slots of a table T. Then, for a given key x, we have

E[# collisions with x] < n/m.

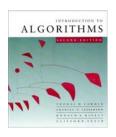


### Proof of theorem

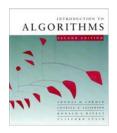
**Proof.** Let  $C_x$  be the random variable denoting the total number of collisions of keys in T with x, and let

 $c_{xy} = \begin{cases} 1 & \text{if } h(x) = h(y), \\ 0 & \text{otherwise.} \end{cases}$ 

Note: 
$$E[c_{xy}] = 1/m$$
 and  $C_x = \sum_{y \in T - \{x\}} c_{xy}$ .



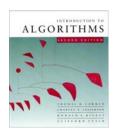
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- Linearity of expectation.
  - $E[c_{xv}] = 1/m$ .



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$$=\frac{n-1}{m}$$
.

- Linearity of expectation.
  - $E[c_{xy}] = 1/m$ .

Algebra.



# Constructing a set of universal hash functions

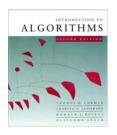
Let m be prime. Decompose key k into r+1digits, each with value in the set  $\{0, 1, ..., m-1\}$ . That is, let  $k = \langle k_0, k_1, ..., k_r \rangle$ , where  $0 \le k_i < m$ .

#### **Randomized strategy:**

Pick  $a = \langle a_0, a_1, ..., a_r \rangle$  where each  $a_i$  is chosen randomly from  $\{0, 1, ..., m-1\}$ .

Define 
$$h_a(k) = \sum_{i=0}^r a_i k_i \mod m$$
. Dot product, modulo m

How big is  $\mathcal{H} = \{h_a\}$ ?  $|\mathcal{H}| = m^{r+1}$ . — REMEMBER THIS!



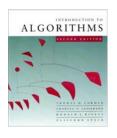
## Universality of dot-product hash functions

**Theorem.** The set  $\mathcal{H} = \{h_a\}$  is universal.

*Proof.* Suppose that  $x = \langle x_0, x_1, ..., x_r \rangle$  and y = $\langle y_0, y_1, ..., y_r \rangle$  be distinct keys. Thus, they differ in at least one digit position, wlog position 0. For how many  $h_a \in \mathcal{H}$  do x and y collide?

We must have  $h_a(x) = h_a(y)$ , which implies that

$$\sum_{i=0}^{r} a_i x_i \equiv \sum_{i=0}^{r} a_i y_i \pmod{m}.$$



Equivalently, we have

$$\sum_{i=0}^{r} a_i (x_i - y_i) \equiv 0 \pmod{m}$$

or

$$a_0(x_0 - y_0) + \sum_{i=1}^r a_i(x_i - y_i) \equiv 0 \pmod{m}$$
,

which implies that

$$a_0(x_0 - y_0) \equiv -\sum_{i=1}^r a_i(x_i - y_i) \pmod{m}$$
.



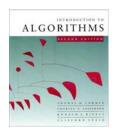
## **Fact from number theory**

**Theorem.** Let m be prime. For any  $z \in \mathbb{Z}_m$  such that  $z \neq 0$ , there exists a unique  $z^{-1} \in \mathbb{Z}_m$  such that

$$z \cdot z^{-1} \equiv 1 \pmod{m}$$
.

Example: m = 7.

$$z$$
 1 2 3 4 5 6  $z^{-1}$  1 4 5 2 3 6



## **Back to the proof**

We have

$$a_0(x_0 - y_0) \equiv -\sum_{i=1}^r a_i(x_i - y_i) \pmod{m},$$

and since  $x_0 \neq y_0$ , an inverse  $(x_0 - y_0)^{-1}$  must exist, which implies that

$$a_0 \equiv \left(-\sum_{i=1}^r a_i(x_i - y_i)\right) \cdot (x_0 - y_0)^{-1} \pmod{m}.$$

Thus, for any choices of  $a_1, a_2, ..., a_r$ , exactly one choice of  $a_0$  causes x and y to collide.



## Proof (completed)

- **Q.** How many  $h_a$ 's cause x and y to collide?
- A. There are m choices for each of  $a_1, a_2, ..., a_r$ , but once these are chosen, exactly one choice for  $a_0$  causes x and y to collide, namely

$$a_0 = \left( \left( -\sum_{i=1}^r a_i (x_i - y_i) \right) \cdot (x_0 - y_0)^{-1} \right) \mod m.$$

Thus, the number of h 's that cause x and y to collide is  $m^r \cdot 1 = |\mathcal{H}|/m$ .

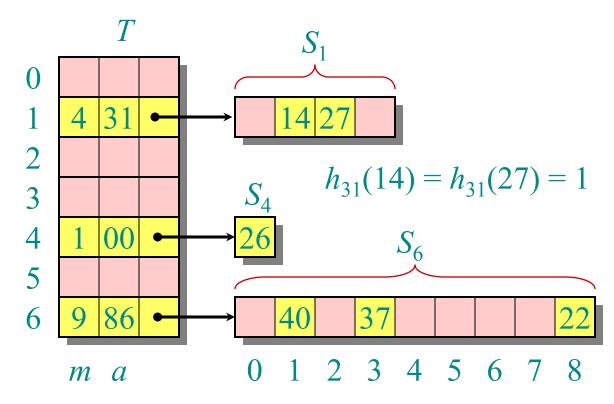


## Perfect hashing

Given a set of n keys, construct a static hash table of size m = O(n) such that SEARCH takes  $\Theta(1)$  time in the *worst case*.

IDEA: Two-level scheme with universal hashing at both levels.

No collisions at level 2!





Collisions at level 2

Theorem. Let  $\mathcal{H}$  be a class of universal hash

functions for a table of size  $m = n^2$ . Then, if we use a random  $h \in \mathcal{H}$  to hash n keys into the table, the expected number of collisions is at most 1/2.

**Proof.** By the definition of universality, the probability that 2 given keys in the table collide under h is  $1/m = 1/n^2$ . Since there are  $\binom{n}{2}$  pairs of keys that can possibly collide, the expected number of collisions is

$$\binom{n}{2} \cdot \frac{1}{n^2} = \frac{n(n-1)}{2} \cdot \frac{1}{n^2} < \frac{1}{2} . \square$$



## No collisions at level 2

Corollary. The probability of no collisions is at least 1/2.

*Proof. Markov's inequality* says that for any nonnegative random variable *X*, we have

$$\Pr\{X \ge t\} \le E[X]/t.$$

Applying this inequality with t = 1, we find that the probability of 1 or more collisions is at most 1/2.

Thus, just by testing random hash functions in  $\mathcal{H}$ , we'll quickly find one that works.



## Analysis of storage

For the level-1 hash table T, choose m = n, and let  $n_i$  be random variable for the number of keys that hash to slot i in T. By using  $n_i^2$  slots for the level-2 hash table  $S_i$ , the expected total storage required for the two-level scheme is therefore

$$E\left[\sum_{i=0}^{m-1}\Theta(n_i^2)\right] = \Theta(n),$$

since the analysis is identical to the analysis from recitation of the expected running time of bucket sort. (For a probability bound, apply Markov.)