# Randomized Algorithms II

Inge Li Gørtz

Thank you to Kevin Wayne and Philip Bille for inspiration to slides

# Randomized algorithms

- Last weeks
	- Contention resolution
	- Global minimum cut
	- Expectation of random variables
		- Guessing cards
	- Quicksort
	- Selection
- Today
	- Hash functions and hash tables







# **Hashing**

## **Dictionaries**

- Dictionary problem. Maintain a dynamic set of S ⊆ U subject to the following operations:
	- Lookup(x): return true if  $x \in S$  and false otherwise
	- Insert(x): Set  $S = S \cup \{x\}$
	- Delete(x): Set  $S = S \setminus \{x\}$
- Universe size. Typically  $|U| = 2^64$  and  $|S| << |U|$ .
- Satellite information. Information associated with each element.
- Goal. A compact data structure with fast operations.
- Applications. Many! A key component in other data structures and algorithms.

# Chained Hashing

- Chained hashing [Dumey 1956].
	- $\cdot$  n =  $|S|$ .
	- Hash function. Pick some crazy, chaotic, random function h that maps U to  $\{0, ..., m-1\}$ , where  $m = \Theta(n)$ .
	- Initialise an array A[0, …, m-1].
	- A[i] stores a linked list containing the keys in S whose hash value is i.



# Uniform random hash functions

- $\cdot$  *E.g.*  $h(x) = x \mod 11$ . Not crazy, chaotic, random.
- Suppose  $|U| \ge n^2$ : For any hash function h there will be a set S of n elements that all map to the same position!
	- => we end up with a single linked list.
- Solution: randomization.
	- For every element  $u \in U$ : select h(u) uniformly at random in  $\{0, \ldots, m-1\}$  independently from all other choices.
- Claim. The probability that  $h(u) = h(v)$  for two elements  $u \neq v$  is  $1/m$ .
- Proof.
	- m<sup>2</sup> possible choices for the pair of values (h(u), h(v)). All equally likely.
	- Exactly m of these gives a collision.

## Chained Hashing with Random Hash Function

• Expected length of the linked list for h(x)?

• Indikator random variable:

• Random variable  $L_{\rm x}$  = length of linked list for x.  $L_{\rm x}$ 

$$
L_x = |\{y \in S \mid h(y) = h(x)\}|
$$

1 if  $h(x) = h(y)$ 

$$
I_y = \begin{cases} 1 & \text{if } h(x) = h(y) \\ 0 & \text{otherwise} \end{cases} \qquad \qquad L_x = \sum_{y \in S} I_y \qquad \qquad E[I_y] = \Pr[h(y) = h(x)] = \frac{1}{m} \text{ for } x \neq y.
$$

• The expected length of the linked list for x:

0 otherwise

$$
E[L_x] = E\left[\sum_{y \in S} I_y\right] = \sum_{y \in S} E[I_y] = 1 + \sum_{y \in S \setminus \{x\}} \frac{1}{m} = 1 + (n-1) \cdot \frac{1}{m} = \Theta(1).
$$

# Chained Hashing with Random Hash Function

- Constant time and O(n) space for the hash table.
- But:
	- Need O(U) space for the hash function.
	- Need a lot of random bits to generate the hash function.
	- Need a lot of time to generate the hash function.
- Do we need a truly random hash function?
- When did we use the fact that h was random in our analysis?

## Chained Hashing with Random Hash Function

• Expected length of the linked list for h(x)?

• Indikator random variable:

• Random variable  $L_{\rm x}$  = length of linked list for x.  $L_{\rm x}$ 

$$
L_x = |\{ y \in S \mid h(y) = h(x) \}|
$$

1 if  $h(x) = h(y)$  $L_x = \sum$ *Iy*

$$
I_{y} = \begin{cases} 1 & \text{if } h(x) = h(y) \\ 0 & \text{otherwise} \end{cases} \qquad \qquad L_{x} = \sum_{y \in S} I_{y} \qquad \qquad E[I_{y}] = \begin{vmatrix} Pr[h(y) = h(x)] = \frac{1}{m} & \text{for } x \neq y. \end{vmatrix}
$$

• The expected length of the linked list for x:

0 otherwise

$$
E[L_x] = E\left[\sum_{y \in S} I_y\right] = \sum_{y \in S} E[I_y] = 1 + \sum_{y \in S \setminus \{x\}} \frac{1}{m} = 1 + (n-1) \cdot \frac{1}{m} = \Theta(1).
$$

*y*∈*S*

#### Universal hash functions

- Universal hashing [Carter and Wegman 1979].
	- $\cdot$  Let  $H$  be a family of functions mapping  $U$  to the set  $\{0, \ldots, m-1\}.$
	- $\cdot$   $H$  is universal if for any  $x, y \in U$ , where  $x \neq y$ , and  $h$  chosen uniformly at random in  $H$ ,

 $Pr[h(x) = h(y)] \le 1/m$ .

 $\cdot$  Require that any  $h \in H$  can be represented compactly and that we can compute the value  $h(u)$ efficiently for any  $u\in U$  .

- Positional number systems. For integers x and b, the base-b representation of x is x written in base b.
- Example.
	- $\cdot$  (10)<sub>10</sub> = (1010)<sub>2</sub> (1⋅2<sup>3</sup> + 0⋅2<sup>2</sup> + 1⋅2<sup>1</sup> + 0⋅2<sup>0</sup>)
	- $\cdot$  (107)<sub>10</sub> = (212)<sub>7</sub> (2⋅7<sup>2</sup> + 1⋅7<sup>1</sup> + 2⋅7<sup>0</sup>)

• Hash function. Given a prime p and  $a = (a_1a_2...a_r)_p$ , define

$$
h_a((x_1x_2...x_r)_p) = a_1x_1 + a_2x_2 + ... + a_rx_r \mod p
$$

- Example.
	- $\cdot$  p = 7
	- $a = (107)_{10} = (212)_{7}$
	- $x = (214)_{10} = (424)_{7}$
	- $\cdot h_a(x) = 2 \cdot 4 + 1 \cdot 2 + 2 \cdot 4 \mod 7 = 18 \mod 7 = 4$
- Universal family.
	- $\cdot H = \{h_a | (a_1 a_2 ... a_r)_p \in \{0, ..., p 1\}^r\}$
	- $\cdot$  Choose random hash function from H  $\sim$  choose random a.
	- H is universal (analysis next).
	- $\cdot$  O(1) time evaluation.
	- $\cdot$  O(1) space.
	- Fast construction.



## Uniform Hashing

 $\cdot$  Lemma 1. For any prime  $p$ , any integer  $z\neq 0 \mod p$ , and any two integers  $\alpha,\beta$ :

 $\alpha z = \beta z \mod p \implies \alpha = \beta \mod p.$ 

- Proof.
	- $\cdot$  Show  $(\alpha \beta)$  is divisible by  $p$ :
		- $\alpha z = \beta z \mod p \implies (\alpha \beta)z = 0 \mod p.$
		- By assumption  $z$  not divisible by  $p$ .
		- $\cdot$  Since  $p$  is prime  $\alpha \beta$  must be divisible by  $p$ .
	- Thus  $\alpha = \beta \mod p$  as claimed.

 $a = (a_1 a_2 ... a_r)_p$ , show that if  $x \neq y$  then  $Pr[h_a(x) = h_a(y)] \leq 1/p$ .

\n- Recall: 
$$
x = (x_1x_2...x_r)_p
$$
 and  $y = (y_1y_2...y_r)_p$ :
\n- $x \neq y \Leftrightarrow (x_1x_2...x_r)_p \neq (y_1y_2...y_r)_p \Rightarrow x_j \neq y_j$  for some  $j$ .
\n

 $\cdot$  Lemma 2. Let  $j$  be such that  $x_j \neq y_j$ . Assume the coordinates  $a_i$  have been chosen for all  $i \neq j.$  The probability of choosing  $a_j$  such that  $h_a(x) = h_a(y)$  is  $1/p$ .

$$
h_a(x) = h_a(y) \Leftrightarrow \sum_{i=1}^r a_i x_i \mod p = \sum_{i=1}^r a_i y_i \mod p \Leftrightarrow a_j \begin{cases} x_j - y_j \end{cases} = \sum_{i \neq j} a_i (x_i - y_i) \mod p
$$
  
There is exactly one value  $0 \le a_j < p$  that satisfies  $a_j z = c \mod p$ .  
  

$$
a_j \begin{cases} x_j - y_j \end{cases} = \sum_{i \neq j} a_i (x_i - y_i) \mod p
$$
 fixed value since

- $\cdot$  Assume there was two such values  $a_j$  and  $a'_j$ .
	- Then  $a_j z = a'_j z \mod p$ .
	- Lemma 1  $\Rightarrow a_j = a'_j \mod p$ . Since  $a_j < p$  and  $a'_j < p$  we have  $a_j = a'_j$ .
- Probability of choosing  $a_j$  such that  $h_a(x) = h_a(y)$  is  $1/p$ .

all  $a_i$  fixed for  $i\neq j$ .  $= 0$ 

- $\cdot$  Lemma 2. Let  $j$  be such that  $x_j \neq y_j$ . Assume the coordinates a<sub>i</sub> have been chosen for all  $i \neq j$ . The probability of choosing  $a_j$  such that  $h_a(x) = h_a(y)$  is  $1/p$ .
- Theorem. For random  $a = (a_1 a_2 ... a_r)_p$ , if  $x \neq y$  then  $Pr[h_a(x) = h_a(y)] = 1/p$ .
- Proof.
	- $\cdot$  *E* : the event that  $h_a(x) = h_a(y)$ .
	- $\cdot$   $F_b$  : the event that the values  $a_i$  for  $i\neq j$  gets the sequence of values  $b.$
	- $\cdot$  Lemma 2 shows that  $\Pr[E \,|\, F_b] = 1/p$  for all  $b.$
	- Thus

$$
Pr[E] = \sum_{b} Pr[E \mid F_b] \cdot Pr[F_b] = \sum_{b} \frac{1}{p} \cdot Pr[F_b] = \frac{1}{p} \sum_{b} \cdot Pr[F_b] = \frac{1}{p}
$$

# **Dictionaries**

- Theorem. We can solve the dictionary problem (without special assumptions) in:
	- O(n) space.
	- O(1) expected time per operation (lookup, insert, delete).



- Other universal families.
	- $\cdot$  For prime  $p > 0$ .

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

$$
H = \{h_{a,b} \mid a \in \{1, ..., p-1\}, b \in \{0, ..., p-1\}\}.
$$

 $\cdot$  Hash function from  $k$ -bit numbers to  $l$ -bit numbers.

$$
h_a(x) = (ax \mod 2^k) \gg (k-l)
$$
  

$$
H = \{h_a \mid a \text{ is an odd integer in } \{1, \dots, 2^k - 1\}\}
$$