# Lectures 5: Hashing

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# Hashing

- Useful tool for
  - Dictionary data structures
  - Cryptography
  - Complexity theory
  - Streaming algorithms
- Today
  - Universal hashing
  - Perfect hashing

## Maintaining a Dictionary

- Large universe of "keys" denoted by U
  - U could be all strings of ASCII characters of length at most 80
- Much smaller "dictionary", which is a subset S of U
  - S could be the set of all English words
- Want to support operations to maintain the dictionary
  - Insert(x): add the key x to S
  - Query(x): is the key x in S?
  - Delete(x): remove the key x from S

#### **Dictionary Settings**

- Static: don't support insert and delete operations, just want to optimize your data structure for fast query operations
  - For example, the English dictionary does not change (or only very gradually)
  - Could use a sorted array with binary search
- Insertion-only: just support insert and query operations
- Dynamic: support insert, delete, and query operations
  - Could use a balanced search tree (AVL trees) to get O(log |S|) time per iteration
- Hashing gives an alternative approach, often the fastest and most convenient way to solve these problems

## Formal Hashing Setup

- Universe U is very large
  - $\bullet\,$  E.g., set of ASCII strings of length 80 is  $128^{80}$
- Care about a small subset  $S \subset U$ . Let N = |S|.
  - S could be the names of all students in this class
- Our data structure is an array A of size M and a "hash function" h: U  $\rightarrow$  {0, 1, ..., M-1}.
  - Typically  $M \ll U$ , so can't just store each key x in A[x]
  - Insert(x) will try to place key x in A[h(x)]
- But what if h(x) = h(y) for  $x \neq y$ ? We let each entry of A be a linked list.
  - To insert an element x into A[h(x)], insert it at the top of the list
  - Hope linked lists are small

## Implementation Details

- Hashing easy to implement
- Query(x): compute i = h(x) and walk down the list A[i] until you find x or walk off the list
- Insert(x): place x at the top of the list A[i]
- Delete(x): remove x from the list A[i] by walking down the list

#### How to Choose the Hash Function h?

- Want it to be unlikely that h(x) = h(y) for different keys x and y
- Want our array size M to be of size O(N), where N is number of keys
- Want to quickly compute h(x) given x
  - For now, we will treat this computation as O(1) time
- How long do Query(x) and Delete(x) take?
  - O(length of list A[h(x)]) time
- How long does Insert(x) take?
  - O(1) time no matter what
- So how long can these lists A[h(x)] be?

#### Bad Sets Exist for any Hash Function

- Claim: For any hash function h: U ->  $\{0, 1, 2, ..., M-1\}$ , if  $|U| \ge (N-1)M + 1$ , there is a set S of N elements of U that all hash to the same location
- Proof: If every location had at most N-1 elements of U hashing to it, then we would have  $|U| \le (N-1)M$
- So there's no good hash function h that works for every S. Thoughts?
- Universal Hashing: Let's randomly choose h!
  - We show for *any* sequence of insert, query, and delete operations, the expected number of operations, over a random choice of h, will be small

#### Universal Hashing

• Definition: A randomized algorithm H for constructing a hash function h: U -> {0, 1, 2, ..., M-1} is universal if for all  $x \neq y$ ,  $\Pr_{h \leftarrow H}[h(x) = h(y)] \leq \frac{1}{M}$ 

$$\Pr_{h \leftarrow H}[h(x) = h(y)] \le \frac{1}{M}$$

- We also say a set H of hash functions is a universal hash function family if choosing  $h\in H$  uniformly at random is universal
- Note the condition holds for every x ≠ y, and the randomness is only over the choice of h from H
- Equivalently, for every  $x \neq y$ , we have:  $\frac{|h \in H \mid h(x) = h(y)|}{|H|} \leq \frac{1}{M}$

## Universal Hashing Examples

**Example 1:** The following three hash families with hash functions mapping the set  $\{a, b\}$  to  $\{0, 1\}$  are universal, because at most 1/M of the hash functions in them cause a and b to collide, were  $M = |\{0, 1\}|$ .

	a	b
$h_1$	0	0
$h_2$	0	1

	a	b
$h_1$	0	1
$h_2$	1	0

	a	b
$h_1$	0	0
$h_2$	1	0
$h_3$	0	1

## Examples that are Not Universal

$\neg$	a	b
$h_1$	0	0
$h_3$	1	1

	a	$\boldsymbol{b}$	$\boldsymbol{c}$
$h_1$	0	0	1
$h_2$	1	1	0
$h_3$	1	0	1

• Note that a and b collide with probability more than 1/M = 1/2

# Universal Hashing Example

• The following hash function is universal with  $M = |\{0,1,2\}|$ 

	a	b	c	
$h_0$	0	0	0	← Note!
$h_1$	0	1	2	
$h_2$	1	2	0	
$h_3$	2	0	1	

#### Using Universal Hashing

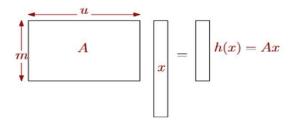
- Theorem: If H is universal, then for any set  $S \subseteq U$  with |S| = N, for any  $x \in S$ , if we choose h at random from H, the **expected** number of collisions between x and other elements in S is at most N/M.
- Proof: For  $y \in S$  with  $y \neq x$ , let  $C_{xy} = 1$  if h(x) = h(y), otherwise  $C_{xy} = 0$  Let  $C_x = \sum_{y \neq x} C_{xy}$  be the total number of collisions with x  $E\big[C_{xy}\big] = \Pr[h(x) = h(y)] \leq \frac{1}{M}$  By linearity of expectation,  $E[C_x] = \sum_{y \neq x} E[C_{xy}] < \frac{N}{M}$

#### Using Universal Hashing

- Corollary: If H is universal, for any **sequence** of L insert, query, and delete operations in which there are at most M keys in the data structure at any time, the expected cost of the L operations for a random  $h \in H$  is O(L)
  - Assumes the time to compute h is O(1)
- Proof: For any operation in the sequence, its expected cost is O(1) by the last theorem, so the expected total cost is O(L) by linearity of expectation

#### But how to Construct a Universal Hash Family?

- Suppose  $|U| = 2^u$  and  $M = 2^m$
- Let A be a random m x u binary matrix, and  $h(x) = Ax \mod 2$



• Claim: for  $x \neq y$ ,  $\Pr_{x}[h(x) = h(y)] = \frac{1}{M} = \frac{1}{2^{m}}$ 

## But how to Construct a Universal Hash Family?

- Claim: For  $x \neq y$ ,  $\Pr_{x}[h(x) = h(y)] = \frac{1}{M} = \frac{1}{2^{m}}$
- Proof:  $A \cdot x \bmod 2 = \sum_i A_i x_i \bmod 2$ , where  $A_i$  is the i-th column of A If h(x) = h(y), then Ax=Ay mod 2, so  $A(x-y) = 0 \bmod 2$  If  $x \neq y$ , there exists an  $i^*$  for which  $x_{i^*} \neq y_{i^*}$  Fix  $A_j$  for all  $j \neq i^*$ , which fixes  $b = \sum_{j \neq i^*} A_j (x_j y_j) \bmod 2$   $A(x-y) = 0 \bmod 2$  if and only if  $A_{i^*} = b$   $\Pr_{A_{i^*}}[A_{i^*} = b] = \frac{1}{2^m} = \frac{1}{M}$

So  $h(x) = Ax \mod 2$  is universal

#### k-wise Independent Families

• Definition: A hash function family H is k-universal if for every set of k distinct keys  $x_1, \dots, x_k$  and every set of k values  $v_1, \dots, v_k \in \{0, 1, \dots, M-1\}$ ,

$$Pr[h(x_1) = v_1 \text{ AND } h(x_2) = v_2 \text{ AND } ... \text{ AND } h(x_k) = v_k] = \frac{1}{M^k}$$

- If H is 2-universal, then it is universal. Why?
- $h(x) = Ax \mod 2$  for a random binary A is not 2-universal. Why?
- Exercise: Show Ax + b mod 2 is 2-universal, where A in  $\{0,1\}^{m\ x\ n}$  and  $b\in\{0,1\}^m$  are chosen independently and uniformly at random

## More Efficient Universal Hashing

- Given a key x, suppose x =  $[x_1, ..., x_k]$  where each  $x_i \in \{0, 1, ..., M-1\}$
- Suppose M is prime
- Choose random  $r_1, \dots, r_k \in \{0,1,\dots,M-1\}$  and define  $h(x) = r_1x_1 + r_2x_2 + \dots + r_kx_k \bmod M$
- Uses less randomness than matrix product
- Claim: the family of such hash functions is universal, that is,  $\Pr_h[h(x)=h(y)] \leq \tfrac{1}{M} \text{ for all } x \text{ and } y$

#### More Efficient Universal Hashing

- Claim: the family of such hash functions is universal, that is,  $\Pr_h[h(x)=h(y)] \leq \frac{1}{M} \text{ for all } x \neq y$
- Proof: Since  $x \neq y$ , there is an  $i^*$  for which  $x_{i^*} \neq y_{i^*}$  Let  $h'(x) = \sum_{j \neq i^*} r_j x_j$ , and  $h(x) = h'(x) + r_{i^*} x_{i^*} \mod M$  If h(x) = h(y), then  $h'(x) + r_{i^*} x_{i^*} = h'(y) + r_{i^*} y_{i^*} \mod M$  So  $r_{i^*}(x_{i^*} y_{i^*}) = h'(y) h'(x) \mod M$ , or  $r_{i^*} = \frac{h'(y) h'(x)}{x_{i^*} y_{i^*}} \mod M$  This happens with probability exactly 1/M

#### Perfect Hashing

- If we fix the dictionary S of size N, can we find a hash function h so that all query(x) operations take constant time?
- Claim: If H is universal and M = N^2, then  $\Pr_{h \leftarrow H}[\text{ no collisions in S}] \geq \frac{1}{2}$
- Proof: How many pairs (x,y) in S are there?

Answer: N(N-1)/2

For each pair, the probability of a collision is at most 1/M

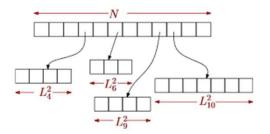
 $Pr[exists a collision] \le (N(N-1)/2)/M \le \frac{1}{2}$ 

Just try a random h and check if there are any collisions

Problem: our hash table has  $M=N^2$  space! How can we get O(N) space?

# Perfect Hashing in O(N) Space – 2 Level Scheme

- Choose a hash function  $h \colon U \to \{1, 2, ... \,, N\}$  from a universal family
- Let  $L_i$  be the number of items x in S for which h(x) = i
- Choose N "second-level" hash functions  $h_1, h_2, ..., h_N$ , where  $h_i \colon U \to \{1, ..., L_i^2\}$



By previous analysis, can choose hash functions  $h_1,h_2,\dots,h_N$  so that there are no collisions, so O(1) time

Hash table size is  $\sum_{i=1,\dots,n} L_i^2$  How big is that??

#### Perfect Hashing in O(N) Space – 2 Level Scheme

• Theorem: If we pick h from a universal family H, then

$$\Pr_{h \leftarrow H} \left[ \sum_{i=1,...,N} L_i^2 > 4N \right] \le \frac{1}{2}$$

• Proof: It suffices to show  $E[\sum_i L_i^2] < 2N$  and apply Markov's inequality Let  $C_{x,y} = 1$  if h(x) = h(y). By counting collisions on both sides,  $\sum_i L_i^2 = \sum_{x,y} C_{x,y}$  If x = y, then  $C_{x,y} = 1$ . If  $x \neq y$ , then  $E[C_{x,y}] = Pr[C_{x,y} = 1] \leq \frac{1}{N}$   $E[\sum_i L_i^2] = \sum_{x,y} E[C_{x,y}] = N + \sum_{x \neq y} E[C_{x,y}] = N + N(N-1)/(2N) < 2N$ 

So just choose a random h in H, and check if  $\sum_{i=1,\dots,n}L_i^2\leq 4N,$  and if so, then choose  $h_1,\dots,h_N$