

Note for Hashing

- **Hash function** => The function is chosen at random.

Given a typically large universe U of keys, and a positive integer m . A random hash function $h : U \rightarrow [m]$ is a randomly chosen function from $U \rightarrow [m]$.

My Understanding - 1

A random hash function is firstly a function that is selected from a set of hash functions randomly and it can map the keys from U to a range of numbers $0, \dots, m - 1$.

Equivalently, => For each x , the value at x is chosen at random.

It is a function h such that for each $x \in U$, $h(x) \in [m]$ is a random variable.

My Understanding - 2

A random hash function is let each key in U be the variable, and the result of hashing every time is random. For example, h_i means i^{th} hashing. $h_1(x) = a, h_2(x) = b$. a, b are random variables.

Chinese Version

1. 随机哈希函数首先是一个从一个含有多个hash functions的集合里随机挑选出来的方程，使得 $U \rightarrow [m]$.
2. 同样可以理解为一个哈希方程是让 U 里的每一个值作为哈希方程的自变量，每次对该自变量映射的结果都是随机的。

宏观上来看，每一个值在经过随机哈希后，输出的值是随机的。

Cryptographic hash functions such as MD5, SHA-1, and SHA-256 are not *random* hash functions.

- **Three things we care**

1. Space (seed size) needed to represent h . => the size of S_h , cannot be too big
2. Time needed to calculate $h(x)$ given $x \in U$. => The inner part of a lot of algorithms is hashing.
3. Properties of the random variable.

- **Hash function types**

Truly random

A hash function $h : U \rightarrow [m]$ is truly random if the variables $h(x)$ for $x \in U$ are **independent** and **uniform**.

一个哈希函数想要 truly random，就得满足对于 $x \in U, h(x)$ 的结果每次都是 m 种可能，每次 hashing 的结果互不影响（独立），且概率都一样，都是 $\frac{1}{m}$ （统一）。

一共有 $|U|$ 个输入，对于每一个输入，需要对应 m 个输出，此时一个输入需要 $\log_2 m$ 字节在计算机里，则一共需要 $|U| \log_2 m$ 个空间。

Universal

A random hash function $h : U \rightarrow [m]$ is **universal** if, for all $x \neq y \in U : \Pr[h(x) = h(y)] \leq \frac{1}{m}$.
 \Rightarrow Hash to the same value.

C-approximately universal

A random hash function $h : U \rightarrow [m]$ is **c-approximately universal** if, for all $x \neq y \in U : \Pr[h(x) = h(y)] \leq \frac{c}{m}$.

Strongly universal

A random hash function $h : U \rightarrow [m]$ is **strongly universal** (a.k.a. 2-independent) if,

1. Each key is hashed *uniformly* into $[m]$. \Rightarrow i.e., $\forall x \in U, q \in [m] : \Pr[h(x) = q] = \frac{1}{m}$.
2. Any two distinct keys hash *independently*.

Equivalently, if for all $x \neq y \in U$, and $q, r \in [m] : \Pr[h(x) = q \wedge h(y) = r] = \frac{1}{m^2}$.

C-approximately strongly universal

A random hash function $h : U \rightarrow [m]$ is c-approximately strongly universal if,

1. Each key is hashed c-approximately uniformly into $[m]$. \Rightarrow i.e.,
 $\forall x \in U, q \in [m] : \Pr[h(x) = q] \leq \frac{c}{m}$
2. Any two distinct keys hash independently.

- **Unordered sets / Hashing with chaining**

Maintain a set S of at most n keys from some unordered universe U , under three operations.

INSERT(x , S) Insert key x into S .

$\text{DELETE}(x, S)$ Delete key x from S .

$\text{MEMBER}(x, S)$ Return $x \in S$.

We could use *some form of balanced tree to store S , but they usually take $O(\log n)$ time operation*, and we want each operation to *run in expected constant time*. =>

The worst case for both INSERT and DELETE is rotating $\log_2 n$ times. And the worst case of MEMBER operation is finding the leaf node. That's the reason why these three operations are all run in $O(\log n)$, while hashing can help us run these three operations in constant time. =>
Hashing with Chaining

- **Hashing with Chaining => Universal Hashing**

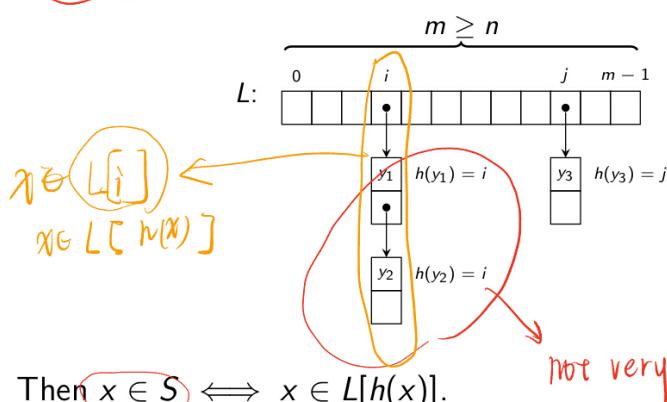
Hashing with chaining

Idea: Pick $m \geq n$ and a *universal* $h : U \rightarrow [m]$.

Store array L , where

$L[i] = \text{linked list over } \{y \in S \mid h(y) = i\}$.

$L[i]$ 存的是 hash 到 i 的所有输入



→ We store an array where index of i in this array is a head of a linked lists that contains all the elements in our sets that hashed to that element.

Each operation takes $\mathcal{O}(|L[h(x)]| + 1)$ time.

先找 $h(x)$ 有 index
然后有链表，并操作

所有可能的输入 y

Hashing with chaining

Theorem

$$\text{For } x \notin S, \mathbb{E}[|L[h(x)]|] \leq 1$$

Then, we store an array where the index of i in this array is a head of a linked list that contains all the elements in our sets that hashed to that element.

这三个方法所花费时间都和链表长度成正比。↓

Each operation take $\mathcal{O}(|L[h(x)]| + 1)$ time. And we need to prove the former part is a constant time.

- **Theorem - 1**

For $x \notin S$, $\mathbb{E}[|L[h(x)]|] \leq 1$. => 找不存在于集合里的 x 所花费的时间。某种程度上算是最差情况，如果最差情况也被bound住，那一般情况肯定在bound里。

Proof.

$$\begin{aligned}
 \mathbb{E}[|L[h(x)]|] &= \mathbb{E}[|\{y \in S | h(y) = h(x)\}|] \Leftarrow \text{By definition} \\
 &= \mathbb{E}\left[\sum_{y \in S} [h(y) = h(x)]\right] \Leftarrow \text{Indicator variable} \\
 &= \sum_{y \in S} \mathbb{E}[[h(y) = h(x)]] \Leftarrow \text{Linearity of expectation} \\
 &= \sum_{y \in S} \Pr[h(y) = h(x)] \Leftarrow \text{Expectation of indicator variable} \\
 &\leq |S| \frac{1}{m} \Leftarrow \text{Since } x \neq y \Rightarrow \text{Universal} \\
 &= \frac{n}{m} \leq 1
 \end{aligned}$$

This actually proves that hashing with chaining and expectation you use only constant time per operation.

- **Signatures => Universal Hashing**

Application: Signatures

Problem: Assign a unique “signature” to each $x \in S \subseteq U$, $|S| = n$.

Solution: Use universal hash function $s : U \rightarrow [n^3]$. \rightarrow The probability of getting a collision among your chosen signatures is very small.
Then by a “union bound”

$$\Pr[\exists \{x, y\} \subseteq S | s(x) = s(y)] \leq \sum_{\{x, y\} \subseteq S} \Pr[s(x) = s(y)] \stackrel{\frac{1}{n^3}}{\rightarrow} \text{It's with high probability that we have no collisions.}$$

$\nearrow p(A \cup B) = p(A) + p(B) - p(A \cap B)$
 $\nearrow \binom{n}{2}$ 有 n 个元素，选择 2 个元素的组合数
 $\nearrow \binom{n^3}{2}$ universal
 $\nearrow \frac{1}{2n}$ 由上式得

Thus with “high probability” we have no collisions.

universal hash function
 \rightarrow cheaper to compute

- **Multiply-mod-prime (2-approximately strongly universal)**

It is the most classic but not the fastest. However, it is good enough for some applications.

Multiply-mod-prime

Let $U = [u]$ and pick prime $p \geq u$. For any $a, b \in [p]$, and $m < u$, let $h_{a,b}^m : U \rightarrow [m]$ be

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

Choose $a, b \in [p]$ independently and uniformly at random, and let $h(x) := h_{a,b}^m(x)$.
Then $h : U \rightarrow [m]$ is a 2-approximately strongly universal hash function.

所以要定义随机
选择
必须保证每次运行程序前都得随机
随机数

Multiply-shift \rightarrow how to work with

Let $U = [2^w]$ and $m = 2^\ell$. For any odd $a \in [2^w]$ define

$$h_a(x) := \left\lfloor \frac{(ax) \bmod 2^w}{2^{w-\ell}} \right\rfloor \xrightarrow{\text{向下取整}} \text{Computer how to work with power of}$$

Choose odd $a \in [2^w]$ uniformly at random, and let $h(x) := h_a(x)$.
Then $h : U \rightarrow [m]$ is a 2-approximately universal hash function.

通过 chaining 和 signature
不要担心溢出，因为他会自动对 m mod

Exercise 3.4 asks you to show if there is some constant c so it is c -approximately strongly universal.

超级 cheap to compute

- **Multiply-shift (2-approximately universal) \Rightarrow Universal Hashing**

Extremely cheaper to compute.

- **Strong Multiply-shift \Rightarrow Strongly Universal Hashing**

It is a strongly universal hash function.

- **Coordinated sampling \Rightarrow Strongly Universal Hashing**

Application: Coordinated sampling

Suppose we have a bunch of *agents* that each observe some set of events from some universe U . Let $A_i \subseteq U$ denote the set of events seen by agent i , and suppose $|A_i|$ is large so only a small sample $S_i \subseteq A_i$ is actually stored.

If each agent independently just samples a random subset of the seen events, there is very little chance that two agents that see an event make the same decision.

\Rightarrow The samples are incomparable.

Coordinated sampling means that all agents that see an event make the same decision about whether to store it.

\Rightarrow Samples can be combined, i.e.

- $S_i \cup S_j$ is a sample of $A_i \cup A_j$
- $S_i \cap S_j$ is a sample of $A_i \cap A_j$

以下有代理人设一个标准(一件事),
如果该事被采样,都存在 S_i 中,不然
都不存.

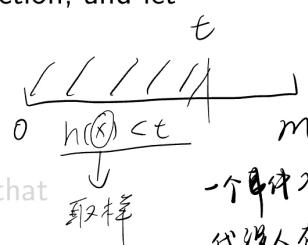
Let $h : U \rightarrow [m]$ be a strongly universal hash function, and let $t \in \{0, \dots, m\}$. Send h and t to all the agents.

Each agent samples $x \in U$ iff $h(x) < t$.

Thus if an agent sees the set $A \subseteq U$, the set

$S_{h,t}(A) := \{x \in A \mid h(x) < t\}$ is sampled. Note that

- $S_{h,t}(A_i) \cup S_{h,t}(A_j) = S_{h,t}(A_i \cup A_j)$
- $S_{h,t}(A_i) \cap S_{h,t}(A_j) = S_{h,t}(A_i \cap A_j)$



一个事件 x 如果被多个代理人看到,
代理人有相同的 h 和 t . 如果 $h(x) < t$
则所有人都存在 S_i 中作为样本.

Application: Coordinated sampling

Let $h : U \rightarrow [m]$ be a strongly universal hash function, and let $t \in \{0, \dots, m\}$. Send h and t to all the agents.

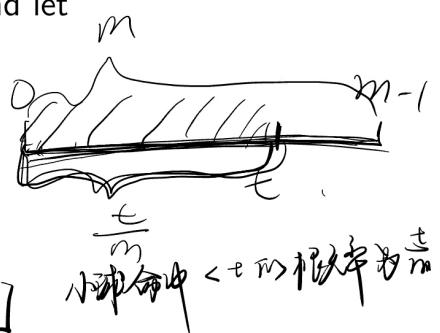
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$$[h(x) = t]$$



Each $x \in A$ is sampled with probability $\Pr[h(x) < t] = \frac{t}{m}$.

Why? Strong universality $\Rightarrow h(x)$ uniform in $[m]$

For any $A \subseteq U$, $\mathbb{E}[|S_{h,t}(A)|] = |A| \cdot \frac{t}{m}$.

$$\text{即 } [样本大小] = |A| \cdot \frac{t}{m}$$

Thus we have an unbiased estimate $|A| \approx \frac{m}{t} \cdot |S_{h,t}(A)|$.

How good is this estimate?

- Lemma

Lemma

Let $X = \sum_{a \in A} X_a$ where the X_a are pairwise independent 0–1 variables.
 Let $\mu = \mathbb{E}[X]$. Then $\text{Var}[X] \leq \mu$, and for any $q > 0$,

$$\Pr[|X - \mu| \geq q\sqrt{\mu}] \leq \frac{1}{q^2}$$

Proof (not curriculum).

For $a \in A$ let $p_a = \Pr[X_a = 1]$. Then $p_a = \mathbb{E}[X_a]$ and

$$\begin{aligned}\text{Var}[X_a] &= \mathbb{E}[(X_a - p_a)^2] = (1 - p_a)(0 - p_a)^2 + p_a(1 - p_a)^2 \\ &= (p_a^2 + p_a(1 - p_a))(1 - p_a) = p_a(1 - p_a) \leq p_a \\ \text{Var}[X] &= \text{Var}\left[\sum_{a \in A} X_a\right] = \sum_{a \in A} \text{Var}[X_a] \leq \sum_{a \in A} p_a = \mu\end{aligned}$$

Finally, since $\sigma_X = \sqrt{\text{Var}[X]} \leq \sqrt{\mu}$ we get:

$$\begin{aligned}\Pr[|X - \mu| \geq q\sqrt{\mu}] &\leq \Pr[|X - \mu| \geq q\sigma_X] \\ &\leq \frac{1}{q^2} \quad (\text{Chebyshev's ineq.}) \square\end{aligned}$$

- **How good is the unbiased estimate with Lemma?**

Application: Coordinated sampling

Let's apply this lemma to the estimate $|A| \approx \frac{m}{t}|S_{h,t}(A)|$ from our coordinated sampling.

Let $X = |S_{h,t}(A)|$ and for $a \in A$ let $X_a = [h(a) < t]$. Then $X = \sum_{a \in A} X_a$ and for any $a, b \in A$, X_a and X_b are independent. Also, let $\mu = \mathbb{E}[X] = \frac{t}{m}|A|$.

Then for any $q > 0$,

$$\begin{aligned}\Pr\left[\left|\frac{m}{t}|S_{h,t}(A)| - |A|\right| \geq q\sqrt{\frac{m}{t}|A|}\right] \\ = \Pr\left[\left||S_{h,t}(A)| - \frac{t}{m}|A|\right| \geq q\sqrt{\frac{t}{m}|A|}\right] \\ = \Pr[|X - \mu| \geq q\sqrt{\mu}] \leq \frac{1}{q^2}\end{aligned}$$

We needed strong universality in two places for this to work.
 Where? ***h* must be uniform to get unbiased estimate, and pairwise independent for the lemma.**

Todays topic was hashing, and we have covered

- ▶ What is a random hash function, and what properties do we want.
- ▶ Two applications of universal hashing — unordered sets and signatures. ✓
- ▶ Some concrete universal or strongly universal hash functions. ✓
- ▶ An application of strongly universal hashing coordinated sampling. ✓
- ▶ Next time: An ordered set data structure that is not comparison based, and an application of hash tables.