

CS 473 Algorithms: Lecture 9

logistics: - midterm 1 Oct 7
- quicksort

last time: - concentration bounds
- limited independence

today: - hashing

Important notion Z

goal: - develop randomized algo w/ $\max_{x \in \mathcal{X}} E[T(x)] =: \bar{T}(n)$ small

- " vs $\Pr[T(x) > \bar{T}(n)] \leq \text{small}$ [concentration bound]

thm (Chernoff): $X_1, \dots, X_n \in [0, 1]$ indep. $X = \frac{\sum X_i}{n}, \epsilon \geq 0$

$$\Pr[X \geq 2E[X]] \leq e^{-\epsilon^2 n/3}$$

[among other forms]

rmk: - sums of independent random variables are concentrated [weak law of large numbers]

- fully independent random variables can be expensive

- randomness
- space [will see]
- algorithmically

Q: concentration bounds without full independence? [somehow independent?]

def: X_1, \dots, X_n are independent if $\forall \sigma_1, \dots, \sigma_n$

$$\Pr[X_1 = \sigma_1, \dots, X_n = \sigma_n] = \Pr[X_1 = \sigma_1] \cdots \Pr[X_n = \sigma_n]$$

are uniform if $\forall i, \Pr[X_i = \sigma_i] = \frac{1}{|\Sigma|}$

X_1, \dots, X_n are pairwise independent (and uniform) if $\forall i \neq j, \Pr[X_i = \sigma_i \wedge X_j = \sigma_j] = \Pr[X_i = \sigma_i] \cdot \Pr[X_j = \sigma_j]$

ex: $X, Y \in \{0, 1\}$ independent uniform

$\Rightarrow X, Y, X \oplus Y$ are pairwise independent and uniform [not independent]

$$\begin{aligned} \text{eg: } \Pr[X=1, X \oplus Y=0] &= \Pr[X \oplus Y=0 | X=1] \cdot \Pr[X=1] = \frac{1}{2} \\ &= \Pr[X \oplus Y=0] = \frac{1}{2} \\ &= \Pr[Y=1] = \frac{1}{2} \end{aligned}$$

[see]

def: $h: U \times \{0, 1\}^l \rightarrow T$ is a pairwise independent hash function

$$\text{if } \forall u_1 \neq u_2 \in U, \Pr_{\substack{s \in \{0, 1\}^l \\ t_1, t_2 \in T}}[h(u_1, s) = t_1 \wedge h(u_2, s) = t_2] = \frac{1}{|T|^2}$$

l = seed length

rmk: \exists the random vars $(h(u, s))_{u \in U}$ are pairwise independent and uniform

- fully random function takes $l = |U| \cdot \lg |T|$ [seed stores entire truth table?]

lem: $\mathbb{Z}_2 = \text{arithmetic modulo 2}$

$$h: \mathbb{Z}_2^n \times (\mathbb{Z}_2^{K \times K} \times \mathbb{Z}_2^K) \rightarrow \mathbb{Z}_2^K$$

$$h: (x, (A, b)) \mapsto A \cdot x + b \quad \leftarrow \boxed{A} \boxed{x}^T + \boxed{b}^K = \boxed{x}^T$$

is pairwise independent hash function, seed length $l = K \cdot (n+1)$

$$\Pr_{A, b}[\underbrace{Ax + b = c \wedge Ay + b = d}_{x \neq y \in \mathbb{Z}_2^n, c, d \in \mathbb{Z}_2^K}] = Ax + b = c$$

$$A(x-y) = c-d$$

$$= \Pr_{A,b} \left[\underbrace{Ax+b=c}_{\text{indeg of } b} \mid \underbrace{A(x-y)=c-d}_{A \neq 0} \right] \cdot \Pr_A [A(x-y) = c-d]$$

$b = c - Ax$
 $\frac{n}{R_2}$
 $\sqrt{\frac{1}{2^n}} = \frac{1}{2^{\frac{n}{2}}}$
 $= \binom{n}{k}^{\frac{n}{2}}$

exercise: $\frac{n}{R_2} = \frac{1}{2^{\frac{n}{2}}}$

rank: seed length $l = k(n+1) = O(k+n) \ll k \cdot 2^n$

- can achieve $O(k+n)$

(Chrysler)
prop: $X_1, \dots, X_n \in \{0,1\}$ pairwise indep. $X = \frac{\sum X_i}{n} \approx 0$

$$\Pr[X - \mathbb{E}[X] \geq \varepsilon] \leq \frac{1}{\varepsilon^2 n}$$

Sketch: exercise: X_1, \dots, X_n pairwise indep. $\Rightarrow \text{Var}(\sum X_i) = \sum \text{Var}(X_i)$

prop: $h: \mathbb{Z}_p \times (\mathbb{Z}_p \times \mathbb{Z}_p) \rightarrow \mathbb{Z}_p$ p prime

$h: x \times (a, b) \mapsto ax + b$ is pairwise independent hash family

rank: seed length $l \geq 2 \lceil \log p \rceil \ll (\log p)^2$

- requires finding prime numbers

def: a field \mathbb{F} is a set st. - $(\mathbb{F}, +)$ commutative group \mathbb{F} has size p
associativity
- $(\mathbb{F}, \cdot, 1, \alpha)$ commutative group associativity

eg.: $\mathbb{R}, \mathbb{C}, \mathbb{Q}$

- distributivity $\alpha(\beta + \gamma) = \alpha \cdot \beta + \alpha \cdot \gamma$

- finite fields have $|\mathbb{F}| < \infty$.

Prop: $\mathbb{Z}_p = \{0, 1, \dots, p-1\} =: \mathbb{F}_p$ is a field w/ arithmetic mod p

If: $(\mathbb{F}_p, +)$: exercise

commutativity: exercise

$(\mathbb{F}_p, \cdot, 1)$: identity: 1

associativity: exercise

inverses?

lem: p prime, $0 \neq x \in \mathbb{F}_p$. $\exists! y \in \mathbb{F}_p$ s.t. $xy \equiv 1 \pmod{p}$

Pf: define $M_x: \mathbb{F}_p \rightarrow \mathbb{F}_p$
 $y \mapsto xy$

Clm: M_x injective

$$\text{pf: } y_1, y_2 \in \mathbb{F}_p \quad M_x(y_1) = M_x(y_2) \quad \equiv \quad xy_1 = xy_2 \quad (\pmod{p})$$

$$x(y_1 - y_2) \equiv 0 \quad (\pmod{p})$$

$$p \mid x(y_1 - y_2) \quad \text{but } p \nmid x$$

$$p \mid (y_1 - y_2)$$

$$y_1 = y_2 \quad (\pmod{p}) \Rightarrow y_1 = y_2 \text{ in } \mathbb{F}_p$$

Clm: M_x bijection

Pf: $\mathbb{F}_p \xrightarrow{M_x} \mathbb{F}_p \xrightarrow{\text{bijection}}$
 M_x injective

$\Rightarrow (\mu_x)^{-1}: \mathbb{F}^p \rightarrow \mathbb{F}^p$ is bijective

$\Rightarrow (\mu_x)^{-1}(z) = y$ is unique y s.t. $\underbrace{\mu_x(y)}_{=x} = z$

rmk: - any prime p , $k \in \mathbb{Z}$, exist unique \mathbb{F} w/ $|\mathbb{F}| = p^k$ [did $k=1$]

1st - $h: \mathbb{F}_p \times (\mathbb{F}_p \times \mathbb{F}_p) \rightarrow \mathbb{F}_p$ is pairwise indep

$$x(a, b) \mapsto ax + b$$

$$\Pr_{\substack{a, b \\ c, d \in \mathbb{F}_p}} [ax + b = c \wedge ay + b = d] = \Pr_{\substack{a \\ c, d}} [ax + b = c \wedge a(x-y) = c-d]$$

$$= \Pr_{\substack{a, b \\ c, d}} [ax + b = c] \cdot \Pr_{\substack{a \\ c, d}} [a(x-y) = c-d]$$

$$= \Pr_{\substack{a \\ c, d}} [a = \frac{c-d}{x-y}] = \gamma_p$$

$$= \gamma_p^2$$

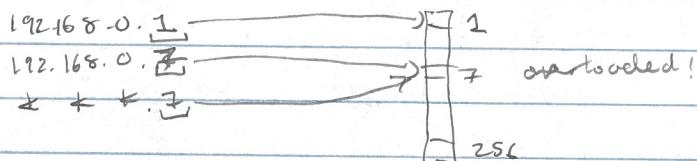
□

Rmk: - can define t -wise independence

- $h: \mathbb{F}_p \times (\mathbb{F}_p)^{t+1} \rightarrow \mathbb{F}_p$ defined by $x \mapsto a_0 + a_1 x + a_2 x^2 + \dots + a_t x^t$

- concentration bounds for t -wise indep get stronger as $t \rightarrow \infty$

Q: how to load balance on the internet?



goal: minimize the maximum load

def: A dictionary is a data structure that maintains $S \subseteq U$. [Keys that we store]

1) $\text{lookup}(x)$: is $x \in S$?

2) $\text{insert}(x)$: add x to S [no op if $x \in S$]

3) $\text{delete}(x)$: remove x from S [no op if $x \notin S$]

is static: all insertions before lookup [no delete]

dynamic: intermingle insert/lookup/delete

ex: - sorted array = $O(\lg |S|)$ lookup

- self-balancing binary search tree = $O(\lg |S|)$ operations

Q: do "better"?

A. relax: $|S|$ space $\mapsto O(|S|)$ space

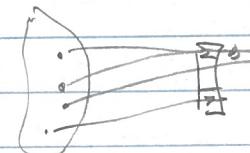
deterministic \mapsto Monte Carlo randomize [same probability of error]

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def: A hash table w/ chaining is a dictionary defined by

hash. $h: U \rightarrow T$, $|T|=m$ $x \in S$ is stored as $h(x)$

linked list



rank: other collision strategies exist

- dictionary operators take $\leq \max_{x \in S} |T[h(x)]|$

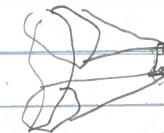
ex: $h: U \rightarrow U$ identity

goal: min \overline{r}

- max load = 1
 - no space savings b/c $|S| \ll |U|$

lem: $h: U \rightarrow T$ fixed. $\exists S \subseteq U$ w/ $h(S) = \mathbb{Z}_t$, $|S| \geq \frac{|U|}{|T|}$

pf: pigeonhole



\Rightarrow in worst case cannot use single fixed function

lem. $S \subseteq U$ \mathbb{I} static $h: U \rightarrow T$ random

\mathbb{I} in practice can hope this does not happen
 OR we crypto to make worst case behavior hard to find \mathbb{I}

$$\forall x \in S \quad \mathbb{E}[|T[h(x)]|] = 1 + \frac{|S|-1}{|T|}$$

$$\text{pf.} \quad = \Pr[x \in T[h(x)]] + \sum_{y \neq x} \Pr[y \in T[h(x)]] = 1 + \frac{|S|-1}{|T|}$$

rank: $n=m \Rightarrow O(1)$ \mathbb{I} load op time \mathbb{I}

- storing h takes $|U| \cdot |S|T|$ bits $\gg |S|$

lem. $\forall h: U \times \{0,1\}^d \rightarrow T$ pairwise independent

$$\text{pf.} \quad \Pr[y \in T[h(x,y)]] = \Pr[h(y,s) = h(x,s)] = \frac{1}{|T|}$$

rank: - can use $h(x) = Ax+b$ or $h(x) = ax+b$ \mathbb{I} fast evaluation
 - cheap to store h

- only need $\Pr[h(x,s) = h(y,s)] \leq \frac{1}{|T|}$ "universal" hash function

- can extend to dynamic hashing \mathbb{I} w/ limitations \mathbb{I}

fact: $\frac{|S|}{|T|} = n$: pairwise indep $\Rightarrow \mathbb{E} \max_{t \in T} |T[t]| \leq O(\sqrt{n})$ \mathbb{I} right

h random

h $O(\frac{\lg n}{\lg \lg n})$ indep

still cheaper than

$O(\frac{\lg n}{\lg \lg n})$

$O(\frac{\lg n}{\lg \lg n})$

logistics: - pros due w/o
 - middle 1 Oct 7

today: - limited independence
 - hashing

review: - randomized algo