

Plan

Logistics

Hashing Around the Clock

Concentration Inequalities

Load Balancing (Review)

Games!

Problem set due tomorrow

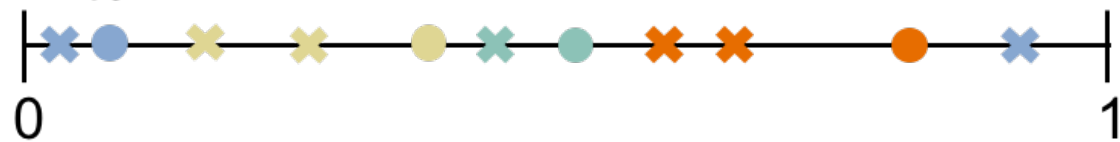
Not available tomorrow,
↳ ask me today!

↳ post on canvas

Hashing Around the Clock

● - server

✕ - data item



(1) $\mathbb{E}[\# \text{ requests to move}]$

(2) $\Pr(\text{any server "owns"} \geq c \text{ fraction}) \leq 1/10$

union bound

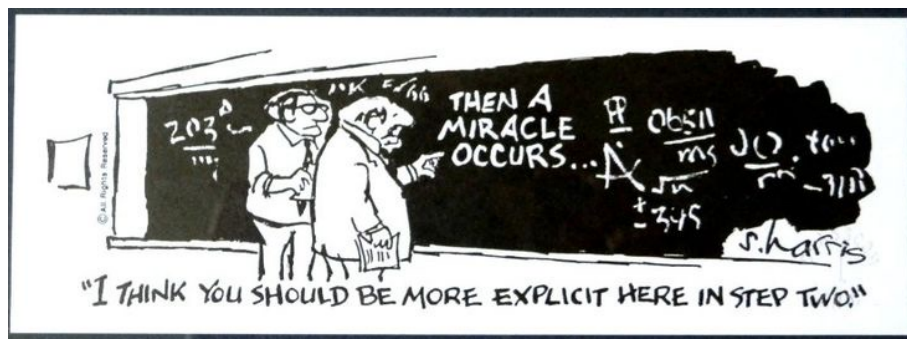
$$\Pr(\text{one server "owns"} \geq c) \leq \frac{1}{10n}$$

$$\Pr(\text{one server "owns"} \geq c)$$

$$= (1-c)^{n-1}$$

(then a miracle occurs)

$$\leq \frac{1}{10n}$$



Concentration Inequalities

Goal: Develop "stronger" inequalities

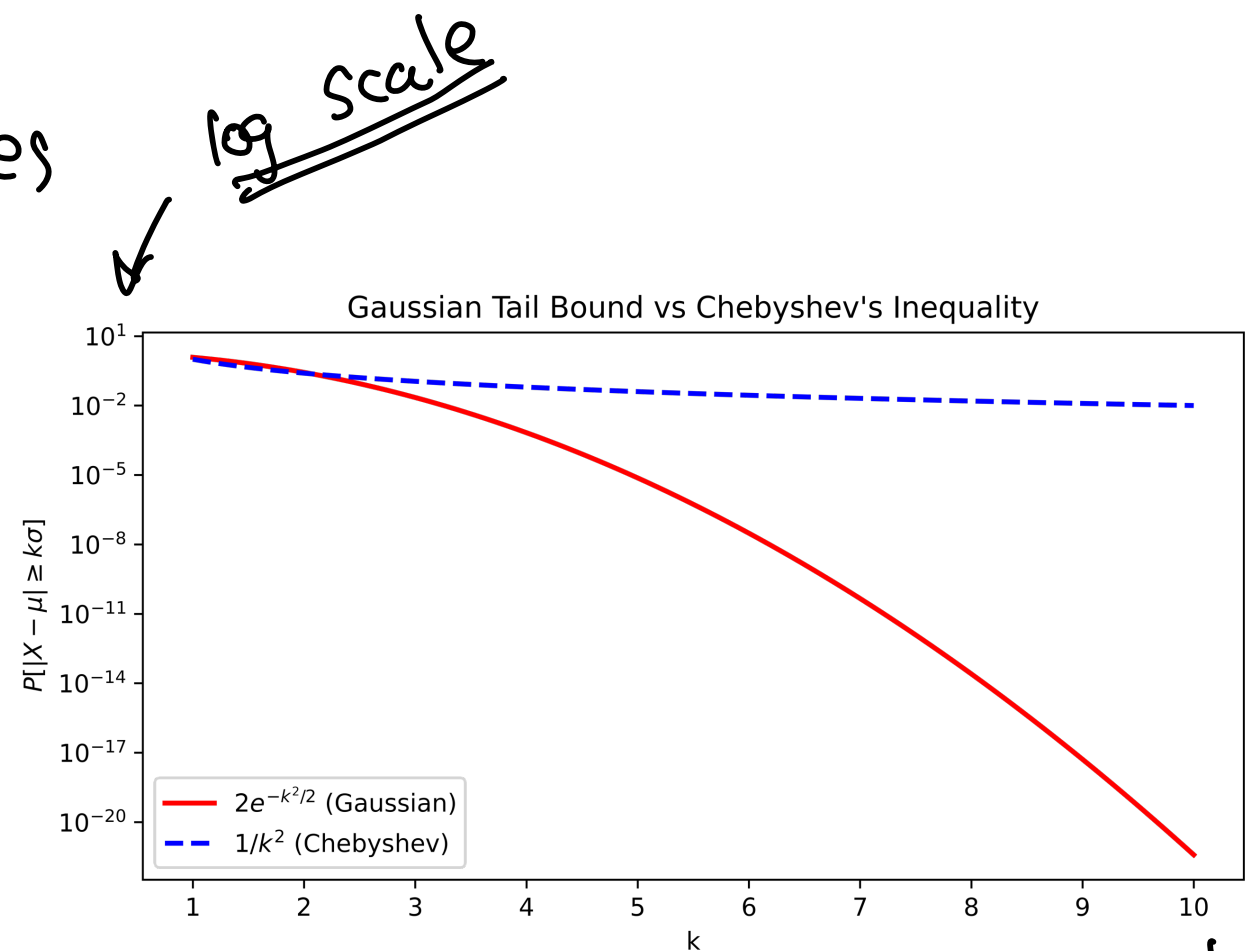
$$\mu \triangleq \mathbb{E}[X] \quad \sigma^2 \triangleq \text{Var}(X)$$

Chebyshev

$$\Pr(|X - \mu| \geq k \cdot \sigma) \leq \frac{1}{k^2}$$

Gaussian X

$$\Pr(|X - \mu| \geq k \cdot \sigma) \leq 2e^{-k^2/2} = 2 \cdot \frac{1}{e^{k^2/2}}$$



Q: Is Chebyshev just bad?

Chebyshev

$$\Pr(|X - \mu| \geq k \cdot \sigma) \leq \frac{1}{k^2}$$

$$\Rightarrow \begin{aligned} \mu &= 0 \\ \sigma &= 1 \end{aligned}$$

$$\Pr(|X| \geq k) \stackrel{\text{"tight!"}}{=} \frac{1}{k^2}$$

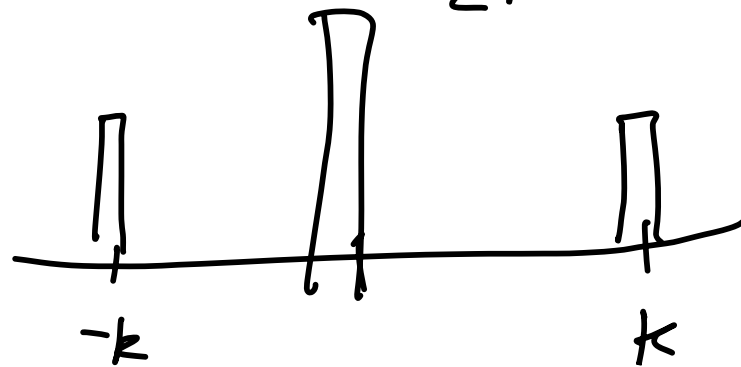
$$\Pr(X \geq k \text{ or } X \leq -k) = \frac{1}{k^2}$$

$$\Pr(X \geq k) + \Pr(X \leq -k) \stackrel{\text{"want equal"}}{=} \frac{1}{k^2}$$

$$\Pr(X \geq k) = \frac{1}{2k^2}$$

$$X = \begin{cases} k & \text{wp } \frac{1}{2k^2} \\ 0 & \text{wp } 1 - \frac{1}{k^2} \\ -k & \text{wp } \frac{1}{2k^2} \end{cases}$$

$$\mathbb{E}[X] = k \cdot \frac{1}{2k^2} + 0 \cdot 1 + (-k) \cdot \frac{1}{2k^2} = 0$$



$$\begin{aligned} \text{Var}(X) &= \mathbb{E}[X^2] - \mathbb{E}[X]^2 = \mathbb{E}[X^2] = (k)^2 \cdot \frac{1}{2k^2} + 0 \cdot 1 + (-k)^2 \cdot \frac{1}{2k^2} \\ &= \frac{k^2}{2k^2} + \frac{k^2}{2k^2} = 1 \end{aligned}$$

⇒ We need assumptions for stronger concentration inequalities

Central Limit Theorem: Any sum of mutually independent and identically distributed random variables X_1, \dots, X_k with mean μ and finite variance σ^2 converges to a Gaussian random variable with mean $k \cdot \mu$ and variance $k \cdot \sigma^2$ as k goes to infinity. Formally, as $k \rightarrow \infty$!!

$$\sum_{i=1}^k X_i \sim \mathcal{N}(k\mu, k\sigma^2)$$

mean ✓ variance ✓

$$\mathbb{E}[X_i] = \mu$$

$$\mathbb{E}\left[\sum_{i=1}^k X_i\right] = \mu k$$

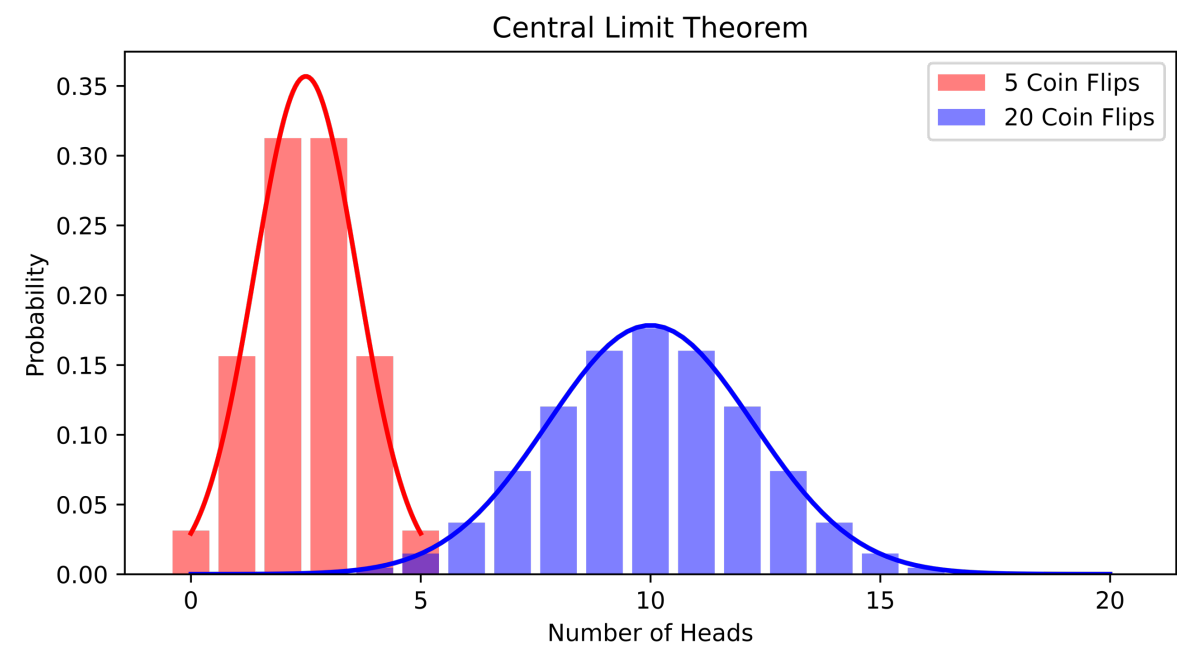
$$\text{Var}(X_i) = \sigma^2$$

$$\text{Var}\left(\sum_{i=1}^k X_i\right) = k \cdot \sigma^2$$

$$\Pr(X_1=x_1, \dots, X_k=x_k) \stackrel{!}{=} \Pr(X_1=x_1) \dots \Pr(X_k=x_k)$$

$$\stackrel{*}{=} \Pr(X_1=x_1) \cdot \Pr(X_2=x_2 | X_1=x_1) \Pr(X_3=x_3 | X_1=x_1, X_2=x_2) \dots$$

always



coin flip

$$H = \sum_{i=1}^{100} C_i$$

$$C_i = \begin{cases} 1 & \text{w.p. } 1/2 \\ 0 & \text{w.p. } 1/2 \end{cases}$$

$$E[H] = 50$$

$$\text{Var}(H) = 25 = \sigma^2 \quad \sigma = 5$$

$$\begin{aligned} \text{Var}(C_i) &= E[C_i^2] - E[C_i]^2 \\ &= 1/2 - 1/4 = 1/4 \end{aligned}$$

Markov's

$$\Pr(X \geq 70) \leq 5/7$$

Chebyshev's

$$\Pr(|X - 50| \geq 20) \leq 0.0625$$

Gaussian

If CLT held exactly

$$\Pr(|X - 50| \geq k \cdot 5) \leq 2e^{-k^2/2}$$

$$k = 4$$

$$\Pr(|X - 50| \geq 20) \leq 2 \exp\left(-\frac{16}{2}\right)$$

$$\approx 0.00067$$

Let's be formal!

$$2 + \epsilon \leq 3 \text{ when } \epsilon < 1$$

$$\frac{1}{3} \leq \frac{1}{2 + \epsilon}$$

$$-\frac{1}{2 + \epsilon} \leq -\frac{1}{3}$$

↳ Different forms

↳ Use typed notes and/or wikipedia

↳ More assumptions → Better bounds

← indicator

Chernoff Bound: Let X_1, \dots, X_k be independent binary random variables. That is, $X_i \in \{0, 1\}$. Let $p_i = \mathbb{E}[X_i]$ where $0 < p_i < 1$. Choose a parameter $\epsilon > 0$. Then the sum $S = \sum_{i=1}^k X_i$, which has mean $\mu = \sum_{i=1}^k p_i$, satisfies

$$\Pr(S \geq (1 + \epsilon)\mu) \leq \exp\left(\frac{-\epsilon^2 \mu}{2 + \epsilon}\right)$$

and, if $0 < \epsilon < 1$,

$$\Pr(S \leq (1 - \epsilon)\mu) \leq \exp\left(\frac{-\epsilon^2 \mu}{2}\right).$$

$$e^x = \exp(x)$$

↳ ϵ can be large

$$\begin{aligned} & \Pr(|S - \mu| \geq \epsilon \mu) \\ &= \Pr(S \geq \mu + \epsilon \mu \text{ or } S \leq \mu - \epsilon \mu) \\ &= \Pr(S \geq (1 + \epsilon)\mu) + \Pr(S \leq (1 - \epsilon)\mu) \\ &\leq \exp\left(\frac{-\epsilon^2 \mu}{2 + \epsilon}\right) + \exp\left(\frac{-\epsilon^2 \mu}{2}\right) \\ &\leq \exp\left(\frac{-\epsilon^2 \mu}{3}\right) + \exp\left(\frac{-\epsilon^2 \mu}{3}\right) \\ &= 2 \exp\left(\frac{-\epsilon^2 \mu}{3}\right) \end{aligned}$$

Less restrictive!

any value between -1 and 1

Bernstein Inequality: Let X_1, \dots, X_k be independent random variables with each $X_i \in [-1, 1]$. Let $\mu = \sum_{i=1}^k \mathbb{E}[X_i]$ and $\sigma^2 = \sum_{i=1}^k \text{Var}[X_i]$. Then, for any $k \leq \frac{\sigma}{2}$, the sum $S = \sum_{i=1}^k X_i$ satisfies

$$\Pr(|S - \mu| > k\sigma) \leq 2 \exp\left(\frac{-k^2}{4}\right).$$

any value between a_i and b_i

Hoeffding's Inequality: Let X_1, \dots, X_k be independent random variables with each $X_i \in [a_i, b_i]$. Let $\mu = \sum_{i=1}^k \mathbb{E}[X_i]$. Then, for any $k > 0$, the sum $S = \sum_{i=1}^k X_i$ satisfies

$$\Pr(|S - \mu| > k) \leq 2 \exp\left(\frac{-k^2}{\sum_{i=1}^k (b_i - a_i)^2}\right).$$

Coin Flips

$$X_i = \begin{cases} 1 & \text{wp } b \\ 0 & \text{wp } 1-b \end{cases}$$

$$S = \sum_{i=1}^k X_i$$

Choose $k \geq \frac{3 \log(2/\delta)}{\epsilon^2}$

(1) $\mathbb{E}[S] = bk = \mu$

(2) $\Pr(|S - bk| \geq \epsilon k) \leq \delta$

Chernoff Bound: Let X_1, \dots, X_k be independent binary random variables. That is, $X_i \in \{0, 1\}$. Let $p_i = \mathbb{E}[X_i]$ where $0 < p_i < 1$. Choose a parameter $\epsilon > 0$. Then the sum $S = \sum_{i=1}^k X_i$, which has mean $\mu = \sum_{i=1}^k p_i$, satisfies

$$\Pr(S \geq (1 + \epsilon)\mu) \leq \exp\left(\frac{-\epsilon^2 \mu}{2 + \epsilon}\right)$$

and, if $0 < \epsilon < 1$,

$$\Pr(S \leq (1 - \epsilon)\mu) \leq \exp\left(\frac{-\epsilon^2 \mu}{2}\right).$$

$$\Pr(|S - \mu| \geq \epsilon \mu) \leq 2 \exp\left(\frac{-\epsilon^2 \mu}{3}\right)$$

$$\Pr(|S - \mu| \geq \epsilon' \mu) \leq 2 \exp\left(-\frac{\epsilon'^2 \mu}{3}\right)$$

$$\mu = bk$$

$$\Pr(|S - bk| \geq \epsilon' \cdot bk) \leq 2 \exp\left(-\frac{\epsilon'^2 bk}{3}\right)$$

$$\epsilon k = \epsilon' bk$$

$$\epsilon' = \frac{\epsilon}{b}$$

$$\Pr(|S - bk| \geq \epsilon k) \leq 2 \exp\left(-\frac{\epsilon^2}{b^2} \cdot \frac{bk}{3}\right)$$

$$\leq 2 \exp\left(-\frac{\cancel{\epsilon^2}}{b^2} \cdot \frac{3 \log(2/\delta)}{\cancel{\epsilon^2}}\right)$$

$$\leq 2 \exp\left(-\frac{1}{b} \log(2/\delta)\right)$$

$$\leq 2 \exp\left(-\frac{1}{b} \log(2/\delta)\right) = 2 \left(\frac{2}{\delta}\right)^{-1/b} = 2 \cdot \frac{\delta}{2}$$

$$b \leq 1$$

$$\frac{1}{1} \leq \frac{1}{b}$$

$$-1 \leq -\frac{1}{b}$$

$$k \geq \frac{3 \log(2/\delta)}{\epsilon^2}$$

Load Balancing

m requests to n servers



union \rightarrow

$$\Pr(\max_i S_i \geq c) \stackrel{\text{want}}{\leq} \frac{1}{10}$$
$$\Pr(S_i \geq c) \stackrel{\text{want}}{\leq} \frac{1}{10n}$$

$$S_i = \sum_{j=1}^m \mathbb{I}[j \text{ goes to } i]$$

assumption $m=n$

$$\mathbb{E}[S_i] = \mu = \frac{m}{n} = 1$$

Chernoff

$$\Pr(S_i \geq (1+\epsilon) \cdot \mu) \leq \exp\left(-\frac{\epsilon^2 \cdot \mu}{2+\epsilon}\right)$$

$$\Pr(S_i \geq (1+\epsilon)) \leq \exp\left(-\frac{\epsilon^2}{2+\epsilon}\right)$$

$$\exp\left(-\frac{\epsilon^2}{2+\epsilon}\right) \stackrel{\text{want}}{\leq} \frac{1}{10n}$$

$$\exp\left(-\frac{\epsilon^2}{2+\epsilon}\right) \stackrel{*}{\leq} \exp\left(-\frac{\epsilon^2}{2\epsilon}\right)$$
$$= \exp\left(-\frac{\epsilon}{2}\right) \stackrel{\text{want}}{=} \frac{1}{10n}$$

$$\epsilon \geq 2 \quad -\frac{\epsilon}{2} = \log\left(\frac{1}{10n}\right)$$
$$2+\epsilon \stackrel{*}{\leq} 2\epsilon \quad \epsilon = 2 \log(10n)$$

Practice: Hash request to 2 servers then
choose the server with smaller load

$O(\log n)$ or $O(\log \log n)$ or $O(1)$ max load
why

atoms in universe $\approx 10^{82}$

$$\log_{10} \log_{10} 10^{82} = \log_{10} 82 \approx 1.91$$