

## Plan

Logistics

Hashing Around the Clock

Concentration Inequalities

Load Balancing (Review-ish)

Thanks for coming to games!

Next wednesday, more pizza,  
more space, 6 or 7?

Problem set tomorrow

All but one said they liked pace,

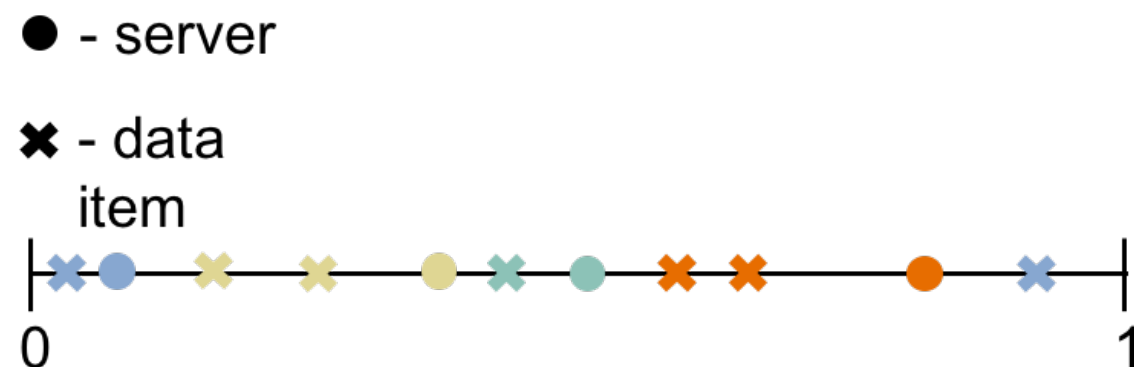
↳ I'll (try to) slow down

↳ more group activities

Not available tomorrow,

ask me today!!

# Hashing Around the Clock



union bound:

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

$$\Pr(\text{one server "owns" } c) = (1-c)^{n-1} \quad (\text{then a miracle occurs})$$

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

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

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

Review in lecture



# Concentration Inequalities

Chebyshev gave "disappointing" bound yesterday :

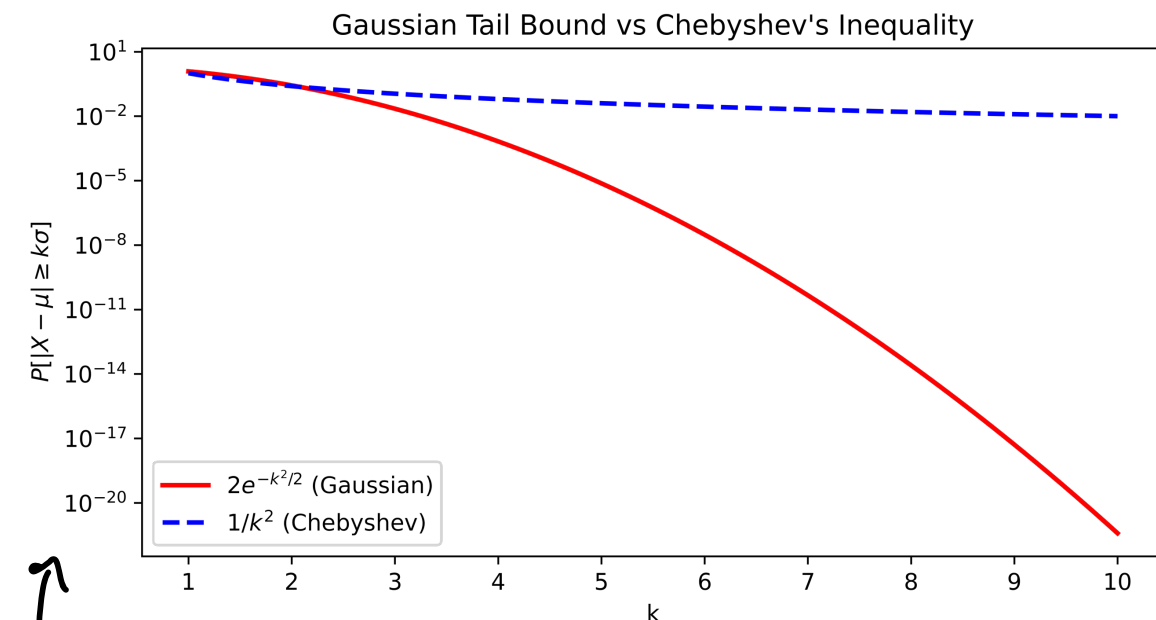
$$\mu = \mathbb{E}[X] \quad \sigma^2 = \text{Var}(X)$$

Chebyshev

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

Gaussian  $X$

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



↑  
Log scale!!

Is Chebyshev just bad?

We need assumptions! Hint?

$$\Pr(X_1 = x_1, \dots, X_k = x_k) \\ = \Pr(X_1 = x_1) \dots \Pr(X_k = x_k)$$

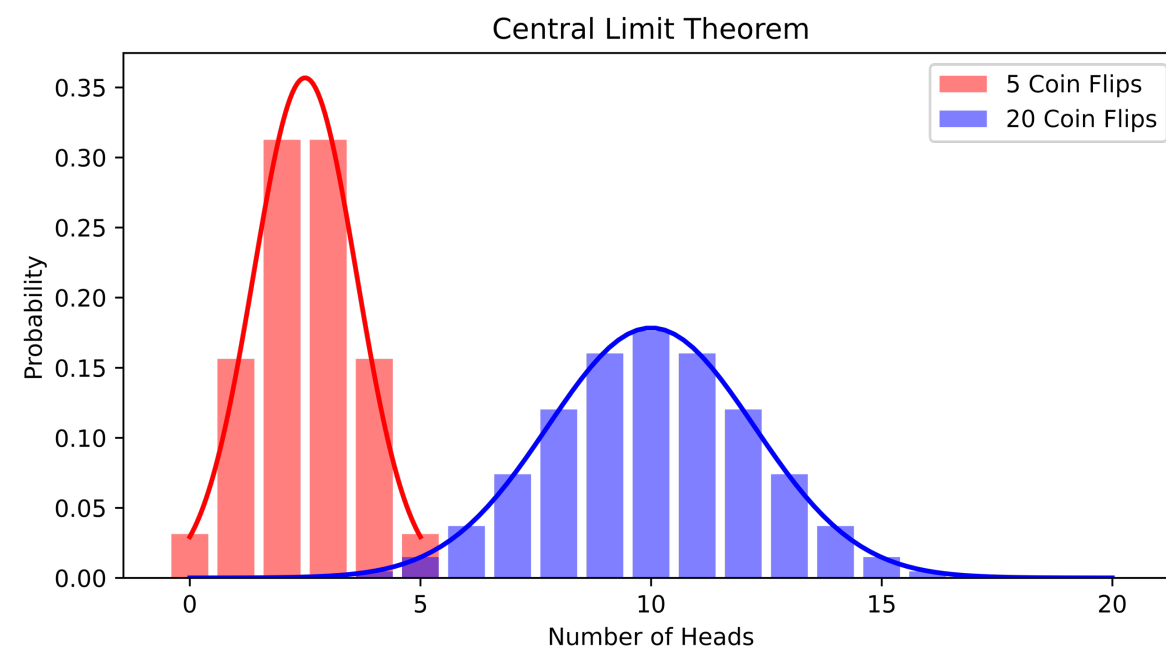
**Central Limit Theorem:** Any sum of mutually independent and identically distributed random variables  $X_1, \dots, X_k$  with mean  $\mu$  and finite variance  $\sigma^2$  converges to a Gaussian random variable with mean  $k \cdot \mu$  and variance  $k \cdot \sigma^2$  as  $k$  goes to infinity.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

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

!!

linearity of variance

linearity of expectation



← better approximation

coin flip

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

$$E[H] = 50$$

$$\text{Var}(H) = 25$$

Chebyshev:

$$\Pr(|X - 50| \geq 20) \leq .0625$$

If CLT held exactly,

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

$$k = 4$$

$$\begin{aligned} \Pr(|X - 50| \geq 20) &\leq 2 \exp(-16/2) \\ &= .00067 \end{aligned}$$

Let's be formal!

↳ Different forms

↳ Use typed notes and/or wikipedia

↳ Different assumptions  $\Rightarrow$  different 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

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

←  $\epsilon$  could be large!  $\Rightarrow$

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

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

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

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$$\left[ \begin{array}{l} 2 + \epsilon \leq 3 \quad \text{when } \epsilon < 1 \\ \frac{1}{3} \leq \frac{1}{2 + \epsilon} \\ -\frac{1}{2 + \epsilon} \leq -\frac{1}{3} \Rightarrow \end{array} \right. \quad \exp\left(-\frac{\epsilon^2\mu}{2 + \epsilon}\right) \stackrel{*}{\leq} \exp\left(-\frac{\epsilon^2\mu}{3}\right)$$

Less restrictive?

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

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

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

$$S = \sum_{i=1}^k X_i \quad \leftarrow \quad = \begin{cases} 1 & \text{wp } b \\ 0 & \text{wp } 1-b \end{cases}$$

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

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

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



## Load Balancing

m requests to n servers



$$\Pr(\max_i S_i \geq c) \leq 1/10$$

$$\Pr(S_i \geq c) \leq \frac{1}{10n}$$

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

↙ binary independent sum!

## Chernoff

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

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

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

$$-\frac{\epsilon}{2} = \log(1/10n)$$

$$\epsilon = 2 \log(10n)$$

$$\Pr(S_i \geq 1 + \sqrt{3 \log(10n)}) \leq \frac{1}{10n}$$

$$\Pr(S_i \geq O(\log n)) \leq \frac{1}{10n}$$

$$\Rightarrow \Pr(\max_i S_i \geq O(\log n)) \leq \frac{1}{10n}$$

Practice: Hash to 2 servers and choose least loaded

$O(\log n)$  or  $O(\log \log n)$  or  $O(1)$  maximum load?

Log log n on desmos! crazy!

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

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