

**CSE 312**

# **Foundations of Computing II**

**Lecture 21: Chernoff Bound & Union Bound**

## Review Tail Bounds

Putting a limit on the probability that a random variable is in the “tails” of the distribution (e.g., not near the middle).

Usually statements in the form of

$$P(\underline{X \geq a}) \leq b$$

or

$$P(|X - \mathbb{E}[X]| \geq a) \leq b$$

## Review Markov's and Chebyshev's Inequalities

**Theorem (Markov's Inequality).** Let  $X$  be a random variable taking only non-negative values. Then, for any  $t > 0$ ,

$$P(\underline{X} \geq \underline{t}) \leq \frac{\mathbb{E}[X]}{t}.$$

$$P(X \geq k \cdot \mathbb{E}[X]) \leq \frac{1}{k}$$

**Theorem (Chebyshev's Inequality).** Let  $X$  be a random variable. Then, for any  $t > 0$ ,

$$P(|X - \mathbb{E}[X]| \geq t) \leq \frac{\text{Var}(X)}{t^2}.$$

# Agenda

- Chernoff Bound ◀
- Example: Server Load, and the union bound

# Chebyshev & Binomial

$$\mathbb{P}(|X - \mathbb{E}[X]| \geq t) \leq \frac{\text{Var}(X)}{t^2}.$$

Reformulated:  $P(|X - \mu| \geq \delta\mu) \leq \frac{\sigma^2}{\delta^2\mu^2}$  where  $\mu = \mathbb{E}[X]$  and  $\sigma^2 = \text{Var}(X)$

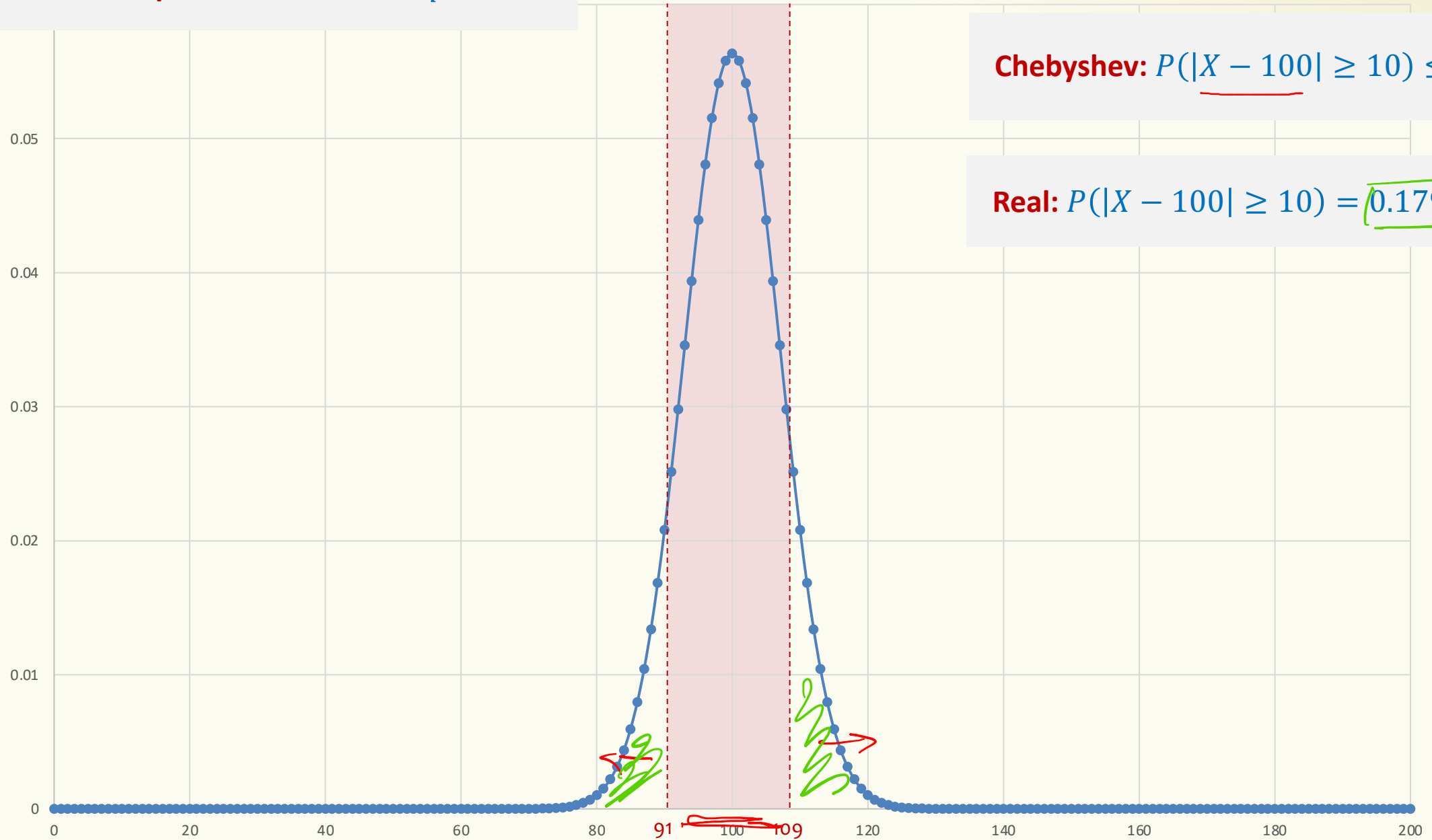
If  $X \sim \text{Bin}(n, p)$ , then  $\mu = np$  and  $\sigma^2 = np(1-p)$

$$P(|X - \mu| \geq \delta\mu) \leq \frac{np(1-p)}{\delta^2 n^2 p^2} = \frac{1}{\delta^2} \cdot \frac{1}{n} \cdot \frac{1-p}{p}$$

E.g.,  $\delta = 0.1$ ,  $p = 0.5$ :  $n = 200$ :  $P(|X - \mu| \geq \delta\mu) \leq 0.5$   $\approx \frac{100}{n}$   $\frac{1-p}{p} = 1$   $\frac{1}{\delta^2} = 100$   
 $n = 800$ :  $P(|X - \mu| \geq \delta\mu) \leq 0.125$

**How good is it?**

Binomial with parameter  $n = 200, p = 0.5$

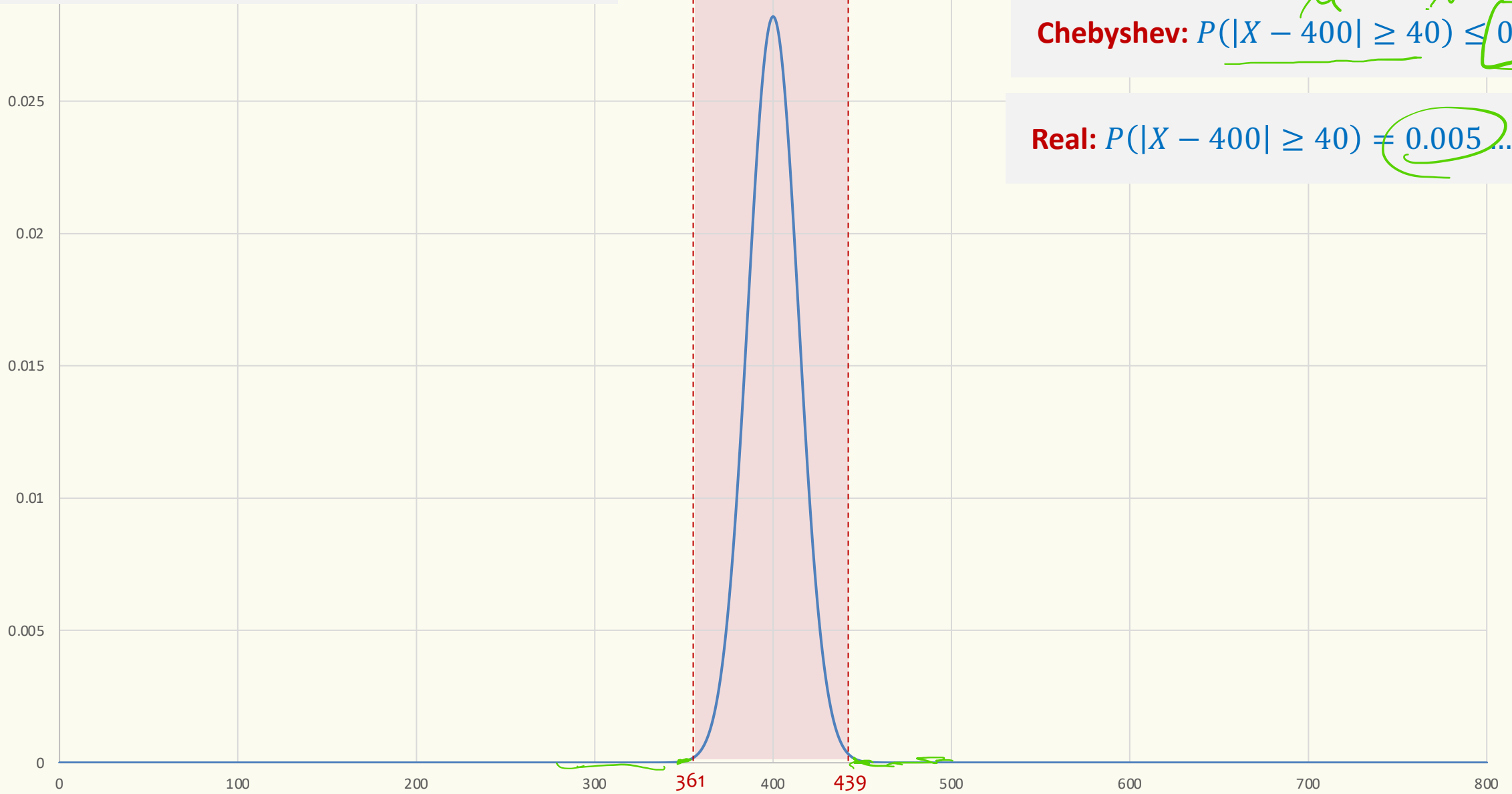


Chebyshev:  $P(|X - 100| \geq 10) \leq \frac{1}{2}$

Real:  $P(|X - 100| \geq 10) = 0.179 \dots$

PMF

Binomial with parameter  $n = 800, p = 0.5$



**Chebyshev:**  $P(|X - 400| \geq 40) \leq 0.125$

**Real:**  $P(|X - 400| \geq 40) = 0.005 \dots$

# Chernoff-Hoeffding Bound

**Theorem.** Let  $X = X_1 + \dots + X_n$  be a sum of independent RVs, each taking values in  $[0,1]$ , such that  $\mathbb{E}[X] = \mu$ . Then, for every  $\delta \in [0,1]$ ,

$$P(|X - \mu| \geq \delta \cdot \mu) \leq e^{-\frac{\delta^2 \mu}{4}} \quad \Theta(\delta^2 \mu)$$

Herman Chernoff, Herman Rubin, Wassily Hoeffding

**Example:** If  $X \sim \text{Bin}(n, p)$ , then  $X = X_1 + \dots + X_n$  is a sum of independent  $\{0,1\}$ -Bernoulli variables, and  $\mu = np$

**Note:** More accurate versions are possible, but with more cumbersome right-hand side (e.g., see textbook) —



# Chernoff-Hoeffding Bound – Binomial Distribution

**Theorem. (CH bound, binomial case)** Let  $X \sim \text{Bin}(n, p)$ . Let  $\mu = np = \mathbb{E}[X]$ . Then, for any  $\delta \in [0, 1]$ ,

$$P(|X - \mu| \geq \delta \cdot \mu) \leq e^{-\frac{\delta^2 np}{4}}.$$

## Example:

$$p = 0.5$$

$$\delta = 0.1$$

Chebyshev Chernoff

$n$	$\frac{1}{\delta^2} \cdot \frac{1}{n} \cdot \frac{1-p}{p}$	$e^{-\frac{\delta^2 np}{4}}$
800	0.125	0.3679
2600	0.03846	0.03877
8000	0.0125	0.00005
80000	0.00125	$3.72 \times 10^{-44}$

# Chernoff Bound – Example

$$\mathbb{P}(|X - \mu| \geq \delta, \mu) \leq e^{-\frac{\delta^2 np}{4}}$$

Alice tosses a fair coin  $n$  times, what is an upper bound for the probability that she sees heads at least  $0.75 \times n$  times?

$X \approx \# \text{ heads}$        $P(X \geq 0.75n)$        $X \sim \text{Bin}(n, 1/2)$   
 $\mu = n/2$

$= P(X - \mu \geq 0.25n)$   
 $\leq P(|X - \mu| \geq 0.25n)$   
 $\leq e^{-\frac{(0.25n)^2}{n \cdot (1/4)}} = e^{-\frac{0.0625n^2}{0.25n}} = e^{-0.25n}$

Poll: [pollev.com/stefanotessararo617](http://pollev.com/stefanotessararo617)

a.  $e^{-n/64}$     2  
 b.  $e^{-n/32}$     1  
 c.  $e^{-n/16}$     1  
 d.  $e^{-n/8}$     4

# Chernoff vs Chebyshev – Summary

$$\frac{1}{\delta^2} \cdot \frac{1}{n} \cdot \frac{1-p}{p}$$

Chebyshev,  
linear  
decrease in  $n$

VS

Chernoff, exponential  
decrease in  $n$

$$e^{-\frac{\delta^2 np}{4}}$$

# Why is the Chernoff Bound True?

Proof strategy (upper tail): For any  $t > 0$ :

- $P(\underline{X} \geq (1 + \delta) \cdot \mu) = P(\underline{e^{tX}} \geq \underline{e^{t(1+\delta) \cdot \mu}})$
- Then, apply Markov + independence:

$$P(e^{tX} \geq e^{t(1+\delta) \cdot \mu}) \leq \frac{\mathbb{E}[e^{tX}]}{e^{t(1+\delta)\mu}} = \frac{\mathbb{E}[e^{tX_1}] \cdots \mathbb{E}[e^{tX_n}]}{e^{t(1+\delta)\mu}}$$

- Find  $t$  minimizing the right-hand-side.

# Brain Break



# Agenda

- Chernoff Bound
- Example: Server Load, and the union bound ◀

# Application – Distributed Load Balancing

We have  $k$  processors, and  $n \gg k$  jobs.

We want to distribute jobs evenly across processors.

**Strategy:** Each job assigned to a randomly <sup>uniformly</sup> chosen processor!

$X_i$  = load of processor  $i$        $X_i \sim \text{Binomial}(n, 1/k)$        $\mathbb{E}[X_i] = n/k$

$X = \max\{X_1, \dots, X_k\}$  = max load of a processor

**Question:** How close is  $X$  to  $n/k$ ?

# Distributed Load Balancing

**Claim. (Load of single server)** If  $n > 16k \ln k$ , then

$$P\left(\underline{X_i} > \frac{n}{k} + 4\sqrt{\frac{n \ln k}{k}}\right) \leq 1/k^4.$$

## Example:

- $n = 10^6 \gg k = 1000$
- $\frac{n}{k} + 4\sqrt{n \ln k / k} \approx 1332$
- “The probability that server  $i$  processes more than 1332 jobs is at most 1-over-one-trillion!”



# Distributed Load Balancing

**Claim. (Load of single server)** If  $n > 16k \ln k$ , then

$$P\left(X_i > \frac{n}{k} + 4\sqrt{\frac{n \ln k}{k}}\right) = P\left(X_i > \frac{n}{k} \left(1 + 4\sqrt{\frac{k \ln k}{n}}\right)\right) \leq 1/k^4.$$

**Proof.** Set  $\mu = \mathbb{E}[X_i] = \frac{n}{k}$  and  $\delta = 4\sqrt{\frac{k}{n} \ln k} < 4\sqrt{\frac{k}{16k \ln k} \ln k} = 1$

$$P\left(X_i > \mu \left(1 + 4\sqrt{\frac{k \ln k}{n}}\right)\right) = P(X_i > \mu(1 + \delta))$$

$$\leq P(|X_i - \mu| > \mu\delta)$$

$$\leq e^{-\frac{\delta^2 \mu}{4}} = e^{-4 \ln k} = \frac{1}{k^4}$$

$n > 16k \ln k$

## What about the maximum load?

**Claim. (Load of single server)** If  $n > 16k \ln k$ , then

$$P\left(X_i > \frac{n}{k} + 4\sqrt{\frac{n \ln k}{k}}\right) \leq 1/k^4.$$

What about  $X = \max\{X_1, \dots, X_k\}$ ?

Note:  $X_1, \dots, X_k$  are not (mutually) independent!

In particular:  $X_1 + \dots + X_k = n$

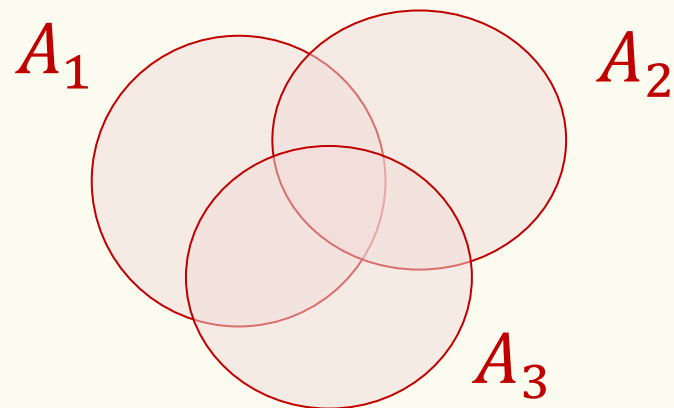
*When non-trivial outcome of one RV can be derived from other RVs, they are non-independent.*

## Detour – Union Bound

**Theorem (Union Bound).** Let  $A_1, \dots, A_n$  be arbitrary events. Then,

$$P\left(\bigcup_{i=1}^n A_i\right) \leq \sum_{i=1}^n P(A_i)$$

Intuition (3 evts.):



## Detour – Union Bound - Example

Suppose we have  $N = 200$  computers, where each one fails with probability  $0.001$ .

What is the probability that at least one server fails?

Let  $A_i$  be the event that server  $i$  fails.

Then event that at least one server fails is  $\bigcup_{i=1}^N A_i$

$$P\left(\bigcup_{i=1}^N A_i\right) \leq \sum_{i=1}^N P(A_i) = 0.001N = 0.2$$


## What about the maximum load?

**Claim. (Load of single server)** If  $n > 16k \ln k$ , then

$$P\left(X_i > \frac{n}{k} + 4\sqrt{\frac{n \ln k}{k}}\right) \leq 1/k^4.$$

What about  $X = \max\{X_1, \dots, X_k\}$ ?

$$\begin{aligned} P\left(X > \frac{n}{k} + 4\sqrt{n \ln k / k}\right) &= P\left(\left\{X_1 > \frac{n}{k} + 4\sqrt{n \ln k / k}\right\} \cup \dots \cup \left\{X_k > \frac{n}{k} + 4\sqrt{n \ln k / k}\right\}\right) \\ &\leq P\left(X_1 > \frac{n}{k} + 4\sqrt{\frac{n \ln k}{k}}\right) + \dots + P\left(X_k > \frac{n}{k} + 4\sqrt{n \ln k / k}\right) \\ &\leq \frac{1}{k^4} + \dots + \frac{1}{k^4} = k \times \frac{1}{k^4} = \frac{1}{k^3} \end{aligned}$$

**Union bound** 

## What about the maximum load?

**Claim. (Load of single server)** If  $n > 16k \ln k$ , then

$$P\left(X_i > \frac{n}{k} + 4\sqrt{\frac{n \ln k}{k}}\right) \leq 1/k^4.$$

**Claim. (Max load)** Let  $X = \max\{X_1, \dots, X_k\}$ . If  $n > 16k \ln k$ , then

$$P\left(\underbrace{X}_{\rightarrow} > \frac{n}{k} + 4\sqrt{\frac{n \ln k}{k}}\right) \leq \underbrace{1/k^3}_{=} = u \cdot \frac{1}{u^3}$$