

# Randomness and Computation

or, “Randomized Algorithms”

Mary Cryan

School of Informatics  
University of Edinburgh



RC (2018/19) – Lecture 7 – slide 1

## Chernoff Bounds from the book

*Poisson trials* - sequence of Bernoulli variables  $X_i$  with varying  $p_i$ s.

### Theorem (4.4)

Let  $X_1, \dots, X_n$  be independent Poisson trials such that  $\Pr[X_i = 1] = p_i$  for all  $i \in [n]$ . Let  $X = \sum_{i=1}^n X_i$ , and  $\mu = \mathbb{E}[X]$ . We have the following Chernoff bounds:

1. For any  $\delta > 0$ ,

$$\Pr[X \geq (1 + \delta)\mu] \leq \left( \frac{e^\delta}{(1 + \delta)^{1 + \delta}} \right)^\mu;$$

2. For any  $0 < \delta \leq 1$ ,

$$\Pr[X \geq (1 + \delta)\mu] \leq e^{-\mu\delta^2/3};$$

3. For  $R \geq 6\mu$ ,

$$\Pr[X \geq R] \leq 2^{-R}.$$



RC (2018/19) – Lecture 7 – slide 3

## Bounding deviation

We already have ...

### Theorem (3.1, Markov's Inequality)

Let  $X$  be any random variable that takes only non-negative values. Then for any  $a > 0$ ,

$$\Pr[X \geq a] \leq \frac{\mathbb{E}[X]}{a}.$$

And also ...

### Theorem (3.2, Chebyshev's Inequality)

For every  $a > 0$ ,

$$\Pr[|X - \mathbb{E}[X]| \geq a] \leq \frac{\text{Var}[X]}{a^2}.$$

These are *generic*. Chernoff/Hoeffding bounds (specific) give tighter bounds for *sums of independent 0/1 variables* and related distributions.



RC (2018/19) – Lecture 7 – slide 2

## Chernoff Bounds from the book

### Lemma

For  $n$  independent Poisson trials  $X_1, \dots, X_n$  and  $X = \sum_{i=1}^n X_i$ ,  $\mu = \mathbb{E}[X]$ ,

$$\mathbb{E}[e^{tX}] \leq e^{\mu(e^t - 1)}.$$

### Proof.

To prove the result, we will consider  $\mathbb{E}[e^{tX}]$  for  $t > 0$ .

This is  $\mathbb{E}[e^{t(\sum_{i=1}^n X_i)}] = \mathbb{E}[\prod_{i=1}^n e^{tX_i}]$ . The  $X_i$  and hence the  $e^{tX_i}$  are mutually independent, so by Thm 3.3,  $\mathbb{E}[e^{tX}] = \prod_{i=1}^n \mathbb{E}[e^{tX_i}]$ .

Each  $e^{tX_i}$  has expectation

$$\begin{aligned} \mathbb{E}[e^{tX_i}] &= p_i \cdot e^t + (1 - p_i) \cdot 1 \\ &= 1 + p_i(e^t - 1) \\ &\leq e^{p_i(e^t - 1)} \quad \text{by } 1 + x \leq e^x \text{ for } x \in \mathbb{R} \end{aligned}$$

$$\mathbb{E}[e^{tX}] \leq \prod_{i=1}^n e^{p_i(e^t - 1)} = e^{\sum_{i=1}^n p_i(e^t - 1)} = e^{\mu(e^t - 1)}.$$



RC (2018/19) – Lecture 7 – slide 4

## Chernoff Bounds from the book

### Proof of Thm 4.4 (1.)

Interested in events when  $X \geq (1 + \delta)\mu$ .

Identical to when  $e^X \geq e^{(1+\delta)\mu}$ , or for any  $t > 0$ , when  $e^{tX} \geq e^{t(1+\delta)\mu}$ .

$$\begin{aligned}\Pr[X \geq (1 + \delta)\mu] &= \Pr[e^{tX} \geq e^{t(1+\delta)\mu}] \\ &\leq \frac{\mathbb{E}[e^{tX}]}{e^{t(1+\delta)\mu}} && \text{by Markov's Inequality} \\ &\leq \frac{e^{\mu(e^t-1)}}{e^{t(1+\delta)\mu}} && \text{by Lemma just proved}\end{aligned}$$

Now take  $t = \ln(1 + \delta)$  (and note this is  $> 0$ ) to see

$$\begin{aligned}\Pr[X \geq (1 + \delta)\mu] &\leq \frac{e^{\mu(e^{\ln(1+\delta)}-1)}}{e^{\ln(1+\delta)(1+\delta)\mu}} \\ &= \frac{e^{\mu\delta}}{(1 + \delta)^{(1+\delta)\mu}} = \left(\frac{e^\delta}{(1 + \delta)^{(1+\delta)}}\right)^\mu.\end{aligned}$$

RC (2018/19) – Lecture 7 – slide 5

## Chernoff Bounds from the book

### Proof of Thm 4.4 (2.)

Already have

$$\Pr[X \geq (1 + \delta)\mu] \leq \left(\frac{e^\delta}{(1 + \delta)^{(1+\delta)}}\right)^\mu.$$

The rhs will be  $\leq e^{-\mu\delta^2/3}$  if and only if (taking  $\mu$ -th root, then  $\ln$ )

$$\delta - (1 + \delta) \ln(1 + \delta) < -\delta^2/3$$

We will show the following  $f$  is always negative for  $\delta \in (0, 1)$

$$f(\delta) =_{\text{def}} \delta - (1 + \delta) \ln(1 + \delta) + \delta^2/3$$

Differentiating,

$$\begin{aligned}f'(\delta) &= 1 - \ln(1 + \delta) - (1 + \delta) \frac{1}{1 + \delta} + \frac{2\delta}{3} \\ &= -\ln(1 + \delta) + \frac{2\delta}{3}.\end{aligned}$$

RC (2018/19) – Lecture 7 – slide 6

## Chernoff Bounds from the book

### Proof of Thm 4.4 (2.) cont'd.

$$f'(\delta) = -\ln(1 + \delta) + \frac{2\delta}{3}.$$

Differentiating again

$$f''(\delta) = -\frac{1}{1 + \delta} + \frac{2}{3} = -\frac{1}{1 + \delta} + \frac{2}{3}$$

Note

$$f''(\delta) \begin{cases} < 0 & \text{for } 0 < \delta < 1/2 \\ 0 & \delta = 1/2 \\ > 0 & \delta > 1/2 \end{cases}$$

Also  $f'(0) = 0$ ,  $f'(1) < 0$  (check  $\delta = 1$  in top equation), and by  $f'$  decreasing first, then increasing from  $1/2$   $f'(\delta) < 0$  on  $(0, 1)$ .

By  $f(0) = 0$ , this implies that  $f(\delta) \leq 0$  in all of  $[0, 1]$ .

Hence  $\delta - (1 + \delta) \ln(1 + \delta) < -\delta^2/3$ , proving 2. □

RC (2018/19) – Lecture 7 – slide 7

## Chernoff Bounds from the book (other direction)

### Theorem (4.5)

Let  $X_1, \dots, X_n$  be independent Poisson trials such that  $\Pr[X_i = 1] = p_i$  for all  $i \in [n]$ . Let  $X = \sum_{i=1}^n X_i$ , and  $\mu = \mathbb{E}[X]$ . For any  $0 < \delta < 1$ , we have the following Chernoff bounds:

1.

$$\Pr[X \leq (1 - \delta)\mu] \leq \left(\frac{e^{-\delta}}{(1 - \delta)^{1-\delta}}\right)^\mu;$$

2.

$$\Pr[X \leq (1 - \delta)\mu] \leq e^{-\mu\delta^2/2};$$

► Proof is similar to Thm 4.4.

► Bound of 2. is slightly *better* than for the  $\geq (1 + \delta)\mu$  bound.

► No 3. Why?

RC (2018/19) – Lecture 7 – slide 8

## Concentration

### Corollary (4.6)

Let  $X_1, \dots, X_n$  be independent Poisson trials such that  $\Pr[X_i = 1] = p_i$  for all  $i \in [n]$ . Let  $X = \sum_{i=1}^n X_i$ , and  $\mu = \mathbb{E}[X]$ . Then for any  $\delta, 0 < \delta < 1$ ,

$$\Pr[|X - \mu| \geq \delta\mu] \leq 2e^{-\mu\delta^2/3}.$$

- ▶ For almost all applications, we will want to work with a *symmetric* version like the Corollary.
- ▶ We “threw away” a bit in moving from the  $\left(\frac{e^{\pm\delta}}{(1\pm\delta)^{1\pm\delta}}\right)^\mu$  versions, but they are tricky to work with.



RC (2018/19) – Lecture 7 – slide 9

## References

- ▶ Chapter 4 of “Probability and Computing”
- ▶ We will continue with Chernoff Bounds on Friday
- ▶ We may not have time to cover the packet routing analysis of 4.5. But it’s worth reading (but not examinable in the exam).



RC (2018/19) – Lecture 7 – slide 11

## Analysing a collection of coin flips

Suppose we have  $p_i = 1/2$  for all  $i \in [n]$ .

We have  $\mu = \mathbb{E}[X] = \frac{n}{2}$ ,  $\text{Var}[X] = \frac{n}{4}$ .

Consider the probability of being further than  $5\sqrt{n}$  from  $\mu$ .

**Chebyshev**  $\Pr[|X - \mu| \geq 5\sqrt{n}] \leq \frac{\text{Var}[X]}{25n} = \frac{1}{100}$

**Chernoff** Work out the  $\delta$  - we need  $\mu\delta = 5\sqrt{n}$ , so need  $\delta = 5\sqrt{n}/\mu = 10\sqrt{n}/n = \frac{10}{\sqrt{n}}$ . Then by Chernoff

$$\Pr[|X - \mu| \geq 5\sqrt{n}] \leq 2e^{-\mu\delta^2/3} = 2e^{\frac{-10^2 \cdot n}{2 \cdot 3 \cdot \sqrt{n}^2}} = 2e^{-16.6\dots}$$

This is much smaller than the Chebyshev bound (though note it doesn’t depend on  $n$ ).

Get much improved bounds because Chernoff uses specialised analysis for sums of independent Bernoulli variables.



RC (2018/19) – Lecture 7 – slide 10