

## Hoeffding's inequality

Let  $X_1, X_2, \dots, X_n$  be independent random variables such that  $X_i \in [a_i, b_i]$  with probability one. Then, for all  $\epsilon > 0$ ,

$$\textcircled{1} \Pr \left[ \sum_{i=1}^n X_i \geq E \left[ \sum_{i=1}^n X_i \right] + \epsilon \right] \leq e^{-2\epsilon^2 / \sum_{i=1}^n (b_i - a_i)^2}$$

$$\textcircled{2} \Pr \left[ \sum_{i=1}^n X_i \leq E \left[ \sum_{i=1}^n X_i \right] - \epsilon \right] \leq e^{-2\epsilon^2 / \sum_{i=1}^n (b_i - a_i)^2}$$

$$\textcircled{3} \Pr \left[ \left| \sum_{i=1}^n X_i - E \left[ \sum_{i=1}^n X_i \right] \right| \geq \epsilon \right] \leq 2e^{-2\epsilon^2 / \sum_{i=1}^n (b_i - a_i)^2}$$

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$\textcircled{1} \Rightarrow \textcircled{2}$  : Apply to  $\textcircled{1}$  to

$-X_1, -X_2, \dots, -X_n$

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① AND ②  $\Rightarrow$  ③ : Union bound

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It suffices to prove ①

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Detour #1 :

Markov's inequality

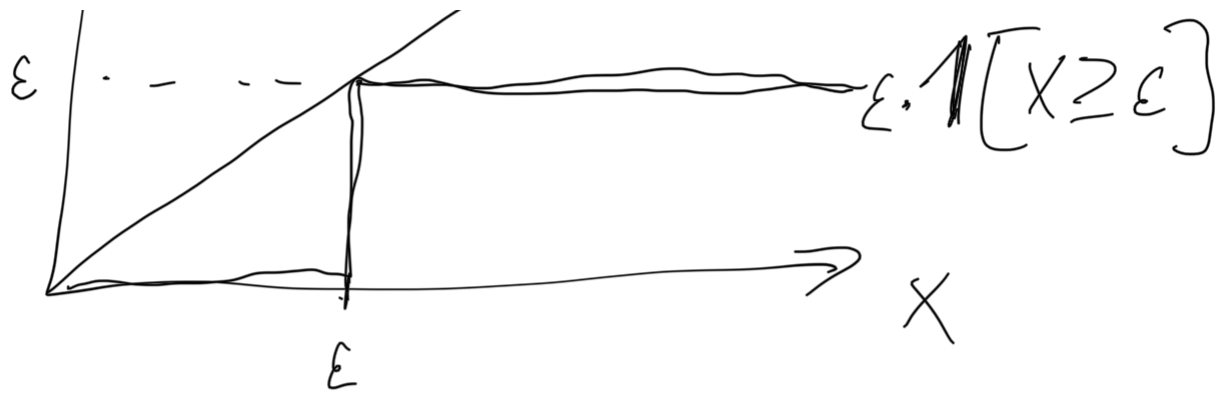
Let  $X$  be a non-negative random variable. Then for any  $\epsilon > 0$ ,

$$P[X \geq \epsilon] \leq \frac{E[X]}{\epsilon}$$

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Proof of Markov's inequality:  $X$

↑



$$\varepsilon \cdot \mathbb{1}[X \geq \varepsilon] \leq X$$

$$\varepsilon \cdot E[\mathbb{1}[X \geq \varepsilon]] \leq E[X]$$

$$P_X[X \geq \varepsilon]$$

$$\varepsilon \cdot P_X[X \geq \varepsilon] \leq E[X] \quad // \frac{1}{\varepsilon}$$

$$P_X[X \geq \varepsilon] \leq \frac{E[X]}{\varepsilon}$$



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Proof Hoeffding's inequality (1):

Use Chernoff's bounding technique. The idea is to apply Markov's inequality to moment generating function of  $\sum_{i=1}^n X_i$ .

$$\begin{aligned} & \Pr \left[ \sum X_i \geq E \sum X_i + \epsilon \right] \leq e^{-s\epsilon} \\ &= \Pr \left[ e^{s \sum X_i} \geq e^{s E \sum X_i + s\epsilon} \right] \\ &\leq \frac{E \left[ e^{s \sum X_i} \right]}{e^{s E \sum X_i + s\epsilon}} \end{aligned}$$

$$= \frac{\prod_{i=1}^n E[e^{sX_i}]}{\prod_{i=1}^n e^{sE[X_i]}}$$

$e^{sX_1}, e^{sX_2}, \dots$   
are independent

$$= e^{-s\epsilon} \prod_{i=1}^n E\left[e^{s(X_i - E[X_i])}\right]$$

Let  $Z_i = X_i - E[X_i]$   $Z_i \in [a_i - E[X_i], b_i - E[X_i]]$

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Detail #2

Hoeffding's lemma

Let  $Z$  be random variable

such that  $E[Z] = 0$

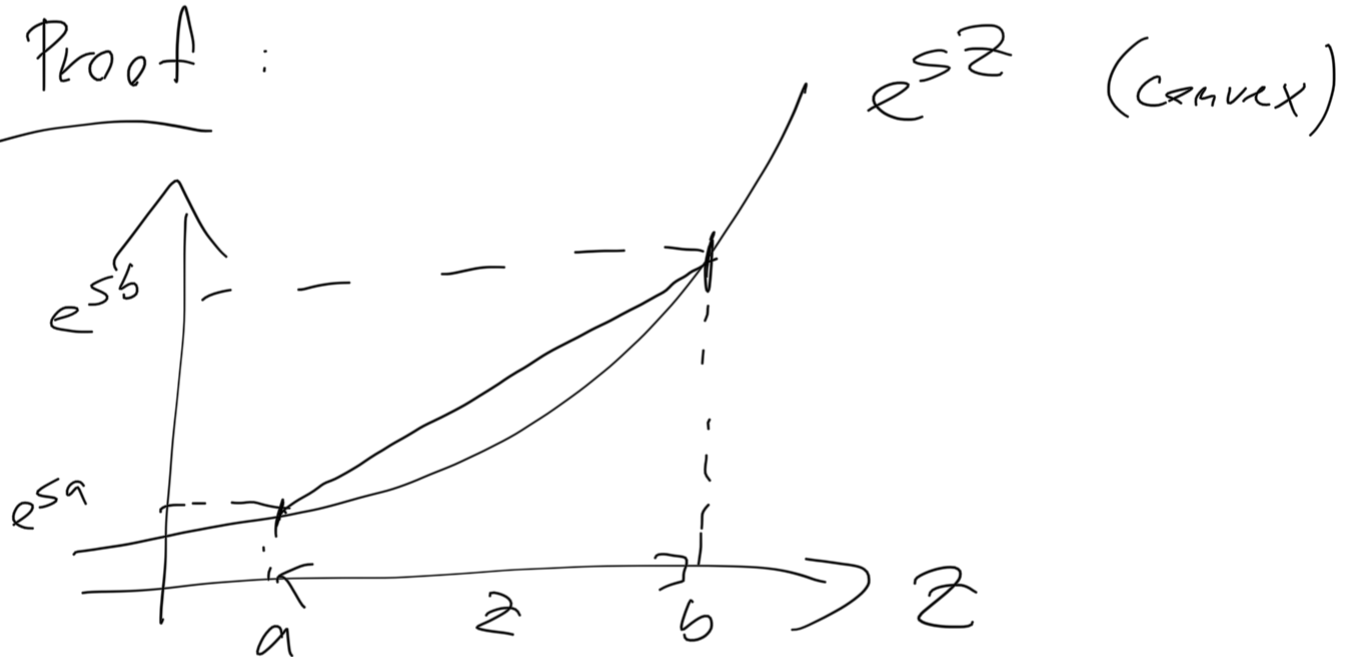
and  $Z \in [a, b]$  w.p. one.

Then for all  $s \geq 0$ ,

$$E[e^{sz}] \leq e^{s^2(b-a)^2/8}$$

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Proof:



$$e^{sz} \leq e^{sa} + (z-a) \cdot \frac{(e^{sb} - e^{sa})}{b-a}$$

$$E[e^{sz}] \leq e^{sa} + \frac{a(e^{sa} - e^{sb})}{b-a}$$

$$= \frac{b}{b-a} \cdot e^{sa} - \frac{a}{b-a} e^{sb}$$

$b-a$  $b-a$ 

$$p = -\frac{a}{b-a}$$

$$a = -p(b-a) = p(a-b)$$

$$p \geq 0 = (1-p)e^{sa} + pe^{sb}$$

$$= e^{sa} (1-p + pe^{s(b-a)})$$

$$= e^{-ps(b-a)} (1-p + pe^{s(b-a)})$$

$$u = s(b-a)$$

$$= e^{-pu} (1-p + pe^u)$$

$$\phi(u) = -pu + \log(1-p + pe^u)$$

$$= e^{\phi(u)}$$

Need to prove  $\phi(u) \leq s^2(b-a)^2/8$ .

$$\phi(u) = \phi(0) + u \cdot \phi'(0) + \frac{u^2}{2} \cdot \phi''(v)$$

for some  $v \in [0, u]$

$$\phi(0) = 0$$

$$\phi'(u) = -p + \frac{pe^u}{1-p+pe^u}$$

$$\phi'(0) = 0$$

$$\phi''(u) = \frac{pe^u(1-p+pe^u) - (pe^u)^2}{(1-p+pe^u)^2}$$

$$= \frac{p(1-p)e^u}{(1-p+pe^u)^2} = \frac{AB}{(A+B)^2} \leq \frac{1}{4}$$

$$\sqrt{AB} \leq \frac{A+B}{2} \Rightarrow \frac{AB}{(A+B)^2} \leq \frac{1}{4}$$

$$\phi(u) \leq \frac{u^2}{2} \cdot \phi''(v) \leq \frac{u^2}{8} = \frac{e^2(b-a)^2}{8}$$

□



$$\begin{aligned} \Pr[\dots] &\leq e^{-s\varepsilon} \prod_{i=1}^n E[e^{s z_i}] \\ &\leq e^{-s\varepsilon} \prod_{i=1}^n e^{s^2 (b_i - a_i)^2 / 8} \end{aligned}$$

We can choose  $s$  that minimizes the last expression

$$A(s) = -s\varepsilon + \frac{s^2}{8} \sum_{i=1}^n (b_i - a_i)^2$$

$$\Pr[\dots] \leq e^{A(s)}$$

Minimizer of  $A(s)$  is at  $A'(s) = 0$ .

$$A'(s) = -\varepsilon + \frac{s}{4} \sum_{i=1}^n (b_i - a_i)^2$$

Minimizer is at

$$s^* = \frac{4\varepsilon}{\sum_{i=1}^n (b_i - a_i)^2}$$

$$A(s^*) = \frac{4\varepsilon^2}{\sum_{i=1}^n (b_i - a_i)^2} + \frac{16\varepsilon^2}{8 \left( \sum_{i=1}^n (b_i - a_i)^2 \right)^2} \cdot \sum_{i=1}^n (b_i - a_i)^2$$

$$4\varepsilon^2$$

$$2\varepsilon^2$$

$$= - \frac{\sum_{i=1}^n (b_i - a_i)^2}{\sum_{i=1}^n (b_i - a_i)^2} + \frac{\sum_{i=1}^n (b_i - a_i)^2}{\sum_{i=1}^n (b_i - a_i)^2}$$

$$= - \frac{2\varepsilon^2}{\sum_{i=1}^n (b_i - a_i)^2}$$

~~(M)~~

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