

Cheernoff bound

$$X_1, \dots, X_n \text{ indep} \quad X_i \in [0, 1]$$

$$X = \sum_i X_i \quad \mu = E(X)$$

$$\Pr(X > \beta \mu) \leq e^{-\mu (\beta \ln \beta - \beta + 1)} \quad \beta > 1$$

$$\Pr(X < \beta \mu) \leq e^{-\mu (\beta \ln \beta - \beta + 1)} \quad \beta < 1$$

how good is this

$$\beta = 1 + \delta \quad \ln(1 + \delta) \sim \delta$$

$$\text{(e.g. } e^\delta = 1 + \delta + \dots)$$

$$\beta > 1 \text{ case bound} \quad e^{-\mu [(1 + \delta) \delta - (1 + \delta) + 1]}$$

$\beta < 1$ case

$$\beta = 1 - \delta \quad \ln(1 - \delta) \sim -\delta$$

bound

$$e^{-\mu [(1 - \delta) \cdot (-\delta) - (1 - \delta) + 1]}$$

Proving the bound

$$\mathbb{P}(Y \geq x) \leq E(Y)$$

$Y \geq 0$ variable, Markov \uparrow

Use this e^{tX}

$$\text{if } t \geq 0: X > \beta \mu \quad \text{iff} \quad e^{tX} > e^{t\mu\beta}$$

$$\text{if } t < 0: X < \beta \mu \quad \text{iff} \quad e^{tX} > e^{t\mu\beta}$$

What is

$$E(e^{tX}) = E(e^{t \sum_i X_i})$$

$$= E\left(\prod_i e^{tX_i}\right) = \prod_i E(e^{tX_i})$$

assume $E(X_i) = \mu_i$ ($\mu = \sum \mu_i$)

Claim Y_i $\Pr(Y_i=1) = \mu_i \neq 0$ or 1

$$E(e^{tY_i}) \geq E(e^{tX_i})$$

flip X_i then 0/1 win prob X_i to 1

[if X_i takes finite options]

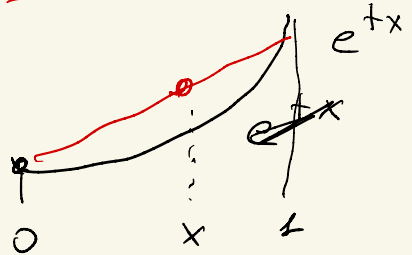
\Rightarrow prob new win 1

$$\sum_x \Pr(X_i=x) \cdot X_i = \mu_i$$

$$E(e^{tY_i} | X_i=x) \geq E(e^{tX_i} | X_i=x)$$

$$e^{tx} + 1(1-x) \geq e^{tx}$$

\Rightarrow



adding all $X_i=x$

$$E(e^{tY_i}) \geq E(e^{tX_i})$$

Using this

$$\begin{aligned} E(e^{tY_i}) &= e^{t\mu_i} + 1(1-\mu_i) \\ &= 1 + \mu_i(e^t - 1) \leq e^{\mu_i(e^t - 1)} \end{aligned}$$

Putting it together

$$\begin{aligned} E(e^{tX}) &= \prod_i E(e^{tX_i}) \leq \prod_i E(e^{tY_i}) \\ &\leq \prod_i e^{\mu_i(e^t - 1)} = e^{\mu(e^t - 1)} \end{aligned}$$

Using Markov

$$\begin{aligned} \beta > 1 \quad t > 0 \quad \Pr(X > \beta\mu) &= \Pr(e^{tX} > e^{\beta\mu t}) \\ &\leq \frac{E(e^{tX})}{e^{\beta\mu t}} \leq e^{\mu(e^t - 1) - \beta\mu t} \end{aligned}$$

if $\beta < 1$
 $t < 0$

$$\begin{aligned} \Pr(X < \beta\mu) &= \Pr(e^{tX} > e^{\beta\mu t}) \\ &\leq \frac{E(e^{tX})}{e^{\beta\mu t}} \leq e^{\mu(e^t - 1) - \beta\mu t} \end{aligned}$$

Optimizing μ of exponent

$$\mu \cdot e^t - \beta \mu \Rightarrow \text{set } t = \ln \beta$$

substituting into result,

bound is

$$e^{\mu(\beta-1) - \beta \mu \cdot \ln \beta}$$

Example X_1, \dots, X_n 0/1

$$E(X_i) = \frac{1}{n} \Rightarrow E(\sum X_i) = 1$$

Question

$$\Pr(\sum X_i > \alpha) \leq \frac{1}{N}$$

Our bound gives Pr estimate

$$e^{(\alpha-1) - \alpha \ln \alpha} < \frac{1}{N}$$

Claim: $\alpha > - \frac{\ln N}{\ln \ln N}$