

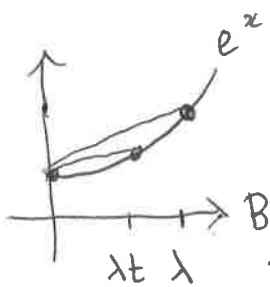
## SOME CONCENTRATION INEQUALITIES

Thm 1 (Chernoff bound) Let  $X_1, \dots, X_n$  be indep. r.v.s with  $0 \leq X_i \leq 1 \quad \forall i \leq n$ . Let  $S = X_1 + \dots + X_n$ ,  $\mu = \mathbb{E}S$ . Then,

$$\mathbb{P}(S > \mu + t) \leq e^{-t} \left(\frac{\mu}{\mu+t}\right)^{\mu+t}, \quad 0 \leq t \leq n - \mu.$$

Proof For  $\lambda > 0$ , we have

$$\begin{aligned} \mathbb{P}(S > \mu + t) &= \mathbb{P}(\lambda S > \lambda(\mu + t)) \leq e^{-\lambda(\mu+t)} \mathbb{E}e^{\lambda S} \\ &= e^{-\lambda(\mu+t)} \prod_{i=1}^n \mathbb{E}e^{\lambda X_i}. \end{aligned}$$



By convexity,  $\forall 0 \leq t \leq 1$

$$\frac{e^{\lambda t} - 1}{\lambda t} \leq \frac{e^\lambda - 1}{\lambda}$$

$$e^{\lambda t} \leq 1 + t(e^\lambda - 1),$$

so  $\mathbb{E}e^{\lambda X_i} \leq 1 + (e^\lambda - 1)\mathbb{E}X_i$ , so

$$\begin{aligned} \prod_{i=1}^n \mathbb{E}e^{\lambda X_i} &\leq \prod_{i=1}^n (1 + (e^\lambda - 1)\mathbb{E}X_i) \\ &\stackrel{\text{AM-GM}}{\leq} \left( \frac{\sum (1 + (e^\lambda - 1)\mathbb{E}X_i)}{n} \right)^n \\ &= \left( 1 + \frac{e^\lambda - 1}{n} \mu \right)^n, \end{aligned}$$

$$\mathbb{P}(S > \mu + t) \leq e^{-\lambda(\mu+t)} \left( 1 + \frac{e^\lambda - 1}{n} \mu \right)^n.$$

Choosing  $\lambda > 0$  which minimises the RHS yields ...

We choose  $e^{-\lambda} = \frac{\mu}{\mu+t}$ , which gives

$$\begin{aligned} \mathbb{P}(S > t + \mu) &\leq \left(\frac{\mu}{\mu+t}\right)^{\mu+t} \left(1 + \frac{\mu}{n} \left(\frac{\mu+t}{\mu} - 1\right)\right)^n \\ &= \left(\frac{\mu}{\mu+t}\right)^{\mu+t} \left(1 + \frac{t}{n}\right)^n \leq e^{t \left(\frac{\mu}{\mu+t}\right)^{\mu+t}}. \quad \square \end{aligned}$$



1) The Chernoff bound is usually written as (set  $t = \delta\mu$ )

$$\mathbb{P}(S > (1+\delta)\mu) \leq \left(\frac{e^\delta}{(1+\delta)^{1+\delta}}\right)^\mu, \quad \delta > 0$$

$$\leq \begin{cases} e^{-\frac{\delta^2 \mu}{3}}, & 0 \leq \delta \leq 1 \\ e^{-\frac{\delta \mu}{3}}, & \delta \geq 1 \end{cases}$$

2) The same arguments give bounds for lower tails

$$\begin{aligned} \mathbb{P}(S < \mu - t) &\stackrel{\lambda < 0}{=} \mathbb{P}(\lambda S > \lambda(\mu - t)) \leq e^{-\lambda(\mu - t)} \mathbb{E} e^{\lambda S} \\ &\leq e^{-t} \left(\frac{\mu}{\mu - t}\right)^{\mu - t}, \quad \mu \geq t > 0. \end{aligned}$$

3) By  $x + \frac{x^2}{2(1+x/3)} \leq (1+x) \log(1+x)$ ,  $x > 0$ , we also get

$$\mathbb{P}(S > \mu + t) \leq \exp\left\{-\frac{t^2}{2(\mu + t/3)}\right\}, \quad t > 0,$$

which is another consequence of the Chernoff bound,

useful in applications.

E.g. 1)  $S \sim \text{Bin}(\frac{1}{2}, n)$

$$\mathbb{P}(S > (1+\delta)\frac{n}{2}) \leq \left(\frac{e^\delta}{(1+\delta)^{1+\delta}}\right)^{n/2}$$

$$\delta = \frac{1}{2}$$

$$\mathbb{P}(S > \frac{3n}{4}) \leq \left(\frac{e^{1/2}}{(3/2)^{3/2}}\right)^{n/2}$$

2)  $S \sim \text{Bin}(\frac{1}{n}, n)$  (Poisson(1))

$$\mathbb{P}(S > 1+\delta) \leq \frac{e^\delta}{(1+\delta)^{1+\delta}}$$

$$\mathbb{P}(S = k) \leq \frac{1}{e} \left(\frac{e}{k}\right)^k$$

$$\left(\mathbb{P}(\text{Poisson}(1) = k) = \frac{1}{e} \frac{1}{k!} \sim \frac{1}{e} \left(\frac{e}{k}\right)^k \frac{1}{\sqrt{2\pi k}}\right)$$

Thm 2 (Bernstein bound) Let  $X_1, \dots, X_n$  be indep. r.v.s with

$|X_i| \leq 1$ ,  $\mathbb{E}X_i = 0$ . Let  $S = X_1 + \dots + X_n$ ,  $\sigma^2 = \text{Var}(S)$ . Then,

$$\mathbb{P}(S > t) \leq \exp\left\{-\frac{t^2}{2(\sigma^2 + t/3)}\right\}, \quad t > 0.$$

Proof For  $\lambda > 0$ ,  $\mathbb{P}(S > t) = \mathbb{P}(\lambda S > \lambda t) \leq e^{-\lambda t} \mathbb{E}e^{\lambda S}$   
 $= e^{-\lambda t} \prod_{i=1}^n \mathbb{E}e^{\lambda X_i}$

We have,  $\mathbb{E}e^{\lambda X_i} = \mathbb{E}\left(1 + \lambda X_i + \frac{1}{2!}(\lambda X_i)^2 + \dots\right)$

$$\leq 1 + \sum_{k=2}^{\infty} \frac{\lambda^k}{k!} \mathbb{E}|X_i|^k$$

For  $k \geq 2$ ,  $\mathbb{E}|X_i|^k = \mathbb{E} \underbrace{|X_i|^{k-2}}_{\leq 1} |X_i|^2 \leq \mathbb{E}X_i^2$ , so

$$\begin{aligned}\mathbb{E}e^{\lambda X_i} &\leq 1 + \left( \sum_{k=2}^{\infty} \frac{\lambda^k}{k!} \right) \mathbb{E}X_i^2 \\ &= 1 + (e^\lambda - \lambda - 1) \mathbb{E}X_i^2 \\ &\leq \exp \{ (e^\lambda - \lambda - 1) \mathbb{E}X_i^2 \},\end{aligned}$$

$$\mathbb{P}(S > t) \leq e^{-\lambda t} \exp \{ (e^\lambda - \lambda - 1) \sigma^2 \} = \exp \left\{ \begin{array}{l} -\lambda(t + \sigma^2) \\ + e^\lambda \sigma^2 - \sigma^2 \end{array} \right\}.$$

Set  $e^\lambda = 1 + \frac{t}{\sigma^2}$ , to get

$$\begin{aligned}\mathbb{P}(S > t) &\leq \exp \left\{ -\left( \frac{t}{\sigma^2} \right) \log \left( 1 + \frac{t}{\sigma^2} \right) + t \right\} \\ &= e^t \left( \frac{\sigma^2}{\sigma^2 + t} \right)^{t + \sigma^2}.\end{aligned}$$

As in  $\triangle 3$ ) we get the assertion.  $\square$