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Chernoff Bound

"The probability that $X_1 + \dots + X_n$ differs from $\mathbb{E}[X_1 + \dots + X_n]$ by more than $t\sqrt{n}$ tends to zero exponentially in t , when X_1, \dots, X_n are independent."

The cumulant generating function of a distribution

If X is a \mathbb{R} -valued random var, its cumulant generating function is

$$K_X(t) = \ln \mathbb{E}[e^{tx}]$$

If X, Y are independent

$$\begin{aligned} K_{X+Y}(t) &= \ln \left(\mathbb{E}[e^{t(X+Y)}] \right) \\ &= \ln \left(\mathbb{E}[e^{tX} e^{tY}] \right) \\ &= \ln \left(\mathbb{E}[e^{tX}] \mathbb{E}[e^{tY}] \right) \\ &= \ln \mathbb{E}[e^{tX}] + \ln \mathbb{E}[e^{tY}] \\ &= K_X(t) + K_Y(t) \end{aligned}$$

If $X \sim N(0,1)$, for any t ,

$$\begin{aligned} \mathbb{E}[e^{tX}] &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{tx} e^{-\frac{1}{2}x^2} dx \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(x^2 - 2tx)} dx \end{aligned}$$

$$= \frac{1}{\sqrt{2\pi}} e^{\frac{1}{2}t^2} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(x-t)^2} dx$$

$$= e^{\frac{1}{2}t^2}$$

$$K_x(t) = \ln(e^{\frac{1}{2}t^2}) = \frac{1}{2}t^2.$$

If X_1, \dots, X_n are identically distributed
 when $\mathbb{E}X_i = 0 \quad \forall i$ and $K_{X_i}(t) = K(t) \dots$

$$K(t) = \ln(\mathbb{E}(e^{tX_i}))$$

$$= \ln(\mathbb{E}(1 + tX_i + \frac{1}{2}t^2X_i^2 + \dots))$$

$$= \ln(1 + t\mathbb{E}(X_i) + \frac{1}{2}t^2\mathbb{E}(X_i^2) + \dots)$$

$$= t\mathbb{E}(X_i) + \frac{1}{2}t^2\text{Var}(X_i) + O(t^3).$$

$$\text{So, } \text{Var}(X_i) = \sigma^2. \quad K_{X_i}(t) = \frac{1}{2}\sigma^2 t^2 + O(t^3).$$

$$Y = \frac{1}{\sqrt{n}}(X_1 + \dots + X_n)$$

$$K_Y(t) = \ln(\mathbb{E}(e^{\frac{1}{\sqrt{n}}(X_1 + \dots + X_n)t}))$$

$$= K_{X_1 + \dots + X_n}\left(\frac{t}{\sqrt{n}}\right)$$

$$= n \cdot K\left(\frac{t}{\sqrt{n}}\right)$$

$$= n \cdot \frac{1}{2}\sigma^2 \left(\frac{t}{\sqrt{n}}\right)^2 + O\left(n \cdot \left(\frac{t}{\sqrt{n}}\right)^3\right)$$

$$= \frac{1}{2}\sigma^2 t^2 + O\left(\frac{t^3}{\sqrt{n}}\right)$$

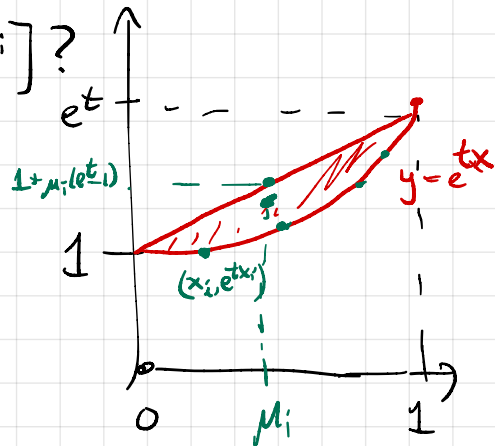
Say X_1, X_2, \dots, X_n all take values in $[0, 1]$.

Then what can we say about $K_{X_i}(t)$?

Let $\mu_i = \mathbb{E}X_i$.

What can we say about $\mathbb{E}[e^{tX_i}]$?

$$e^{t\mu_i} \leq \mathbb{E}[e^{tX_i}] \leq 1 + \mu_i(e^t - 1)$$



Using $\ln(1+x) \leq x$,

$$K_{X_i}(t) \leq \mu_i (e^t - 1).$$

Suppose $X = X_1 + \dots + X_n$ and X_1, \dots, X_n indep.

$$\begin{aligned} K_X(t) &= \sum_i K_{X_i}(t) \leq \left(\sum_i \mu_i \right) (e^t - 1) \\ &= \mathbb{E}[X] (e^t - 1). \end{aligned}$$

What can we say about $\Pr(X > (1+\epsilon)\mathbb{E}[X])$?

If $t > 0$, $X > (1+\epsilon)\mathbb{E}[X]$

\Downarrow

$$e^{tX} > e^{t(1+\epsilon)\mathbb{E}[X]}.$$

$$\Pr\left(e^{tX} > e^{t(1+\epsilon)\mathbb{E}[X]}\right) \leq \frac{\mathbb{E}[e^{tX}]}{e^{t(1+\epsilon)\mathbb{E}[X]}} \quad (\text{Markov's})$$

$$= \frac{e^{K_X(t)}}{e^{t(1+\varepsilon)E(X)}}$$

$$= e^{K_X(t) - (1+\varepsilon)E(X)t}$$

$$\leq e^{(e^t - 1 - (1+\varepsilon)t)E(X)}$$

Set t to minimize $e^t - 1 - (1+\varepsilon)t$:

$$e^t = 1 + \varepsilon$$

$$t = \ln(1 + \varepsilon)$$

$$e^t - 1 - (1+\varepsilon)t = \varepsilon - (1+\varepsilon)\ln(1+\varepsilon)$$

$$= \varepsilon - (1+\varepsilon)\left(\varepsilon - \frac{1}{2}\varepsilon^2 + \frac{1}{3}\varepsilon^3 - \frac{1}{4}\varepsilon^4 + \dots\right)$$

$$= \varepsilon - \left(\varepsilon + \frac{1}{2}\varepsilon^2 - \frac{1}{6}\varepsilon^3 + \frac{1}{12}\varepsilon^4 - \frac{1}{20}\varepsilon^5 + \dots\right)$$

$$\leq -\frac{1}{3}\varepsilon^2 \quad \text{for } 0 \leq \varepsilon < 1.$$

Chernoff. If X_1, \dots, X_n are indep, $[0, 1]$ -valued

$$\Pr(X_1 + \dots + X_n > (1+\varepsilon)E(X_1 + \dots + X_n))$$

$$< e^{-\frac{1}{3}\varepsilon^2 E(X_1 + \dots + X_n)}$$

$$\Pr(X_1 + \dots + X_n < (1-\varepsilon)E(X_1 + \dots + X_n))$$

$$< e^{-\frac{1}{2}\varepsilon^2 E(X_1 + \dots + X_n)}$$