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Let X be non negative r.v. with finite expectation. Consider $a > 0$ a condition

Consider the inequality

$$a \mathbb{1}_{X \geq a} \leq X \quad \text{True, because}$$

$$\begin{aligned} \text{if } X \geq a \quad a(1) \in X &\Rightarrow X \geq a \\ X \leq a \quad P(0 \in X) &\Rightarrow X \geq a \\ &\text{by assumption} \end{aligned}$$

$$E[\mathbb{1}_{X \geq a}] \leq \mu$$

$$\Rightarrow a E[\mathbb{1}_{X \geq a}] \leq \mu \Rightarrow P(X \geq a) \leq \frac{\mu}{a}$$

$$\Rightarrow P(X \geq a) \leq \frac{\mu}{a}$$

Tail bound / Markov's inequality
 $P(X \geq a) \leq \frac{\mu}{a}$

$$\star \text{ let } a = a(\mu) \Rightarrow P(X \geq a(\mu)) \leq \frac{1}{a}$$

let h be ^{strictly} monotonically increasing function

$$h(a) \mathbb{1}_{h(X) \geq h(a)} \leq h(X)$$

$$\Rightarrow P(h(X) \geq h(a)) \leq \frac{E[h(X)]}{h(a)}$$

$$\Rightarrow P(X > a) \leq \frac{E[h(X)]}{h(a)}$$

$$h(x) = X^p \text{ s.t. } p > 1$$

$$P(X > a) \leq \frac{E[X^p]}{a^p}$$

* Recall Quantile $(X, p) = F_X^{-1}(p)$
inv. CDF

$$P(X > a) \leq \frac{\mu}{a}$$

$$\Rightarrow 1 - P(X \leq a) \leq \frac{\mu}{a} \Rightarrow 1 - F(a) \leq \frac{\mu}{a}$$

$$\text{let } a = F^{-1}(p) \Rightarrow 1 - P(F^{-1}(p)) \leq \frac{\mu}{F^{-1}(p)}$$

$$\Rightarrow 1 - p \leq \frac{\mu}{F_X(p)} \Rightarrow \text{Quantile}(X, p) \leq \frac{\mu}{p}$$

$$\Rightarrow \text{Mod}(X) = 2\mu$$

* Consider any r. v. X . $|X|$ is non negative

$$P(|X| > a) \leq \frac{E(|X|)}{a}$$

both tails

* Let X be any r.v. with finite μ , finite σ^2

let $Y = (X - \mu)^2$ note Y is non-negative

$$P(Y \geq a^2) \leq \frac{E(Y)}{a^2} = \frac{E[(X - \mu)^2]}{a^2} = \frac{\sigma^2}{a^2}$$

$$\Rightarrow P((X - \mu)^2 \geq a^2) \leq \frac{\sigma^2}{a^2}$$

$$\Rightarrow P(|X - \mu| \geq a) \leq \frac{\sigma^2}{a^2} \quad \text{Chebyshev's inequality}$$

let X any r.v.

let $Y = e^{tx}$ Y is non-negative

$$P(Y \geq c) \leq \frac{E(Y)}{c}$$

Note

$$M_X(t) = E[e^{tx}]$$

Moment generating function

let $c = e^{ta}$

$$\Rightarrow P(e^{tx} \geq e^{ta}) \leq \frac{E(e^{tx})}{e^{ta}} = \frac{M_X(t)}{e^{ta}}$$

$$\text{If } t > 0 \quad P(X \geq a) \leq e^{-ta} M_X(t)$$

$$\text{If } t < 0 \quad \Rightarrow P(X \geq a) \leq \min_{t > 0} \{ e^{-ta} M_X(t) \}$$

$$P(X \leq a) \leq e^{-ta} M_X(t)$$

Chernoff's Inequality

$$\Rightarrow P(X \leq a) \leq \min_{t < 0} \{ e^{-ta} M_X(t) \}$$

Application for each inequality

$$\rightarrow X \sim \text{Bin}(n, \frac{1}{4})$$

$$P(X \geq \frac{3}{4}n)$$

If n is large

$$X \approx N\left(\frac{1}{4}n, \left(\sqrt{\frac{3}{16}}n\right)^2\right)$$

$$P(X \geq \frac{3}{4}n) = P\left(\frac{X - \frac{1}{4}n}{\sqrt{\frac{3}{16}}n} \geq \frac{\frac{3}{4}n - \frac{1}{4}n}{\sqrt{\frac{3}{16}}n}\right)$$

$$= P(Z > \frac{2}{\sqrt{3}}\sqrt{n})$$

✓ Markov's $P(X \geq \frac{3n}{4}) \leq \frac{\frac{1}{4}n}{\frac{3}{4}n} = \frac{1}{3}$

Chebyshev's $P(X \geq \frac{3n}{4})$

$$= P(X - \frac{1}{4}n \geq \frac{3n}{4} - \frac{1}{4}n)$$

$$= P(X - \frac{1}{4}n \geq \frac{1}{2}n) + P(\frac{1}{4}n - X \geq \frac{1}{2}n)$$

$$= P(X - \frac{1}{4}n \geq \frac{1}{2}n) \cup P(\frac{1}{4}n - X \geq \frac{1}{2}n)$$

$$= P(|X - \frac{1}{4}n| \geq \frac{1}{2}n) \leq \frac{\frac{3}{16}n}{\frac{1}{4}n^2} = \frac{3}{4n}$$

$X \sim \text{Bin}(n, p)$

$$M_X(t) = E[e^{tx}] = \sum_{x=0}^n e^{tx} \binom{n}{x} p^x (1-p)^{n-x}$$

$$= \sum_{x=0}^n \binom{n}{x} (ep)^x (1-p)^{n-x} = (1-p + pe)^n$$

Bin. Theor.

$$X \sim \text{Bin}(n; \frac{1}{4}) \Rightarrow M_X(t) = \left(\frac{3}{4} + \frac{1}{4} e^t \right)^n$$

$$P(X \geq \frac{3}{4}n) \leq \min_{t > 0} \left\{ e^{-t(\frac{3}{4}n)} \left(\frac{3}{4} + \frac{1}{4} e^t \right)^n \right\}$$

$$= \min_{t > 0} \left\{ \left(\frac{3}{4} e^{-\frac{3}{4}t} + \frac{1}{4} e^{\frac{1}{4}t} \right)^n \right\}$$

side

~~$$\Rightarrow n \left(-\frac{1}{4} e^{-\frac{3}{4}t} + \frac{1}{4} e^{\frac{1}{4}t} \right) = 0$$~~

$$\Rightarrow e^{-\frac{1}{4}t} = 9 e^{\frac{3}{4}t}$$

$$\Rightarrow \frac{1}{4}t = \ln(9) - \frac{3}{4}t$$

$$\Rightarrow t_{\min} = \ln(9)$$

Now go back to min formula

$$= \left(\frac{3}{4} e^{-\frac{3}{4}\ln(9)} + \frac{1}{4} e^{\frac{1}{4}\ln(9)} \right)^n$$

$$\rightarrow z = \left(\frac{3}{4} 9^{-\frac{3}{4}} + \frac{1}{4} 9^{\frac{1}{4}} \right)^n$$

$$= \frac{\sqrt[4]{9}}{4^n} \left(\frac{3}{9^{\frac{3}{4}}} + 1 \right)^n = \frac{\sqrt[4]{9} \cdot (1.004)^n}{4^n} \approx 0$$

→ Exponentially fast

Consider any 2 r.v. X, Y with finite μ_x, σ_x^2
 let $W = (X - cY)^2$, $c \in \mathbb{R}$ W is non negative

$\Rightarrow E(W) \geq 0$
 $\Rightarrow E(X - cY)^2 \geq 0 \Rightarrow E(X^2 - 2cXY + c^2Y^2) \geq 0$
 $\Rightarrow E(X^2) - 2cE[XY] + c^2E[Y^2] \geq 0$

Pick $c = \frac{E[XY]}{E[X^2]}$

$$\Rightarrow E[W] = E[X^2] - 2 \frac{E[XY]}{E[X^2]} E[XY] + \frac{E[XY]^2}{E[X^2]^2} E[X^2]$$

$$\Rightarrow E[X^2] E[Y^2] - 2 E[XY]^2 + E[X^2] \geq 0$$

$$\Rightarrow E[XY]^2 \leq E[X^2] E[Y^2]$$

$$\Rightarrow |E[XY]| \leq \sqrt{E[X^2] E[Y^2]}$$

Cauchy-Schwarz Inequality

what is the conclusion

$$\text{Corr}[X, Y] = \text{Corr}[cY, Y] = \frac{\text{Corr}[cY, Y]}{SE[cY] SE(Y)}$$

$$= \frac{c \text{Corr}[Y, Y]}{|c| SE(Y)^2}$$

$$= \frac{c}{|c|} = \begin{cases} 1 & \text{if } c > 0 \\ -1 & \text{if } c < 0 \end{cases}$$