

Cauchy-Schwarz inequality: $|E(UV)| \leq \sqrt{E(U^2)} \sqrt{E(V^2)}$

Proof. Let $F(t) = E[(U+tV)^2] = E(U^2) + 2$
 $= c + 2bt + at^2 \geq 0$ for all t , $c = E(U^2)$, $b =$
Hence discriminant $= (2b)^2 - 4ac = 4(b^2 - ac) \geq 0$
 $|b| \leq \sqrt{ac}$.

Ex 1. Show that $E(|V|) \leq \sqrt{E(V^2)}$

Answer. Cauchy-Schwarz with $U = |V|$, $V = 1$.

More exercises

Ex 1. Let $\mu = E(V)$, $\sigma^2 = \text{Var}(V)$, $\sigma = \sqrt{\text{Var}(V)}$.

Confirm that

$$P(|V - \mu| \leq 3\sigma) \geq \frac{8}{9} \approx 0.89 \left(\begin{array}{l} \text{If } V \text{ is normal, then} \\ P(|V - \mu| \leq 3\sigma) = 0.997 \end{array} \right)$$

Answer. Chebyshev:

$$P(|V - \mu| > 3\sigma) \leq \frac{\text{Var}(V)}{9\sigma^2} = \frac{\sigma^2}{9\sigma^2} = \frac{1}{9}.$$

Ex 2. Let V, Z be independent, $Z \sim N(0,1)$,

$$P(V=1) = P(V=-1) = \frac{1}{2}.$$

a) Show that $X = VZ \sim N(0,1)$

Answer.

$$\begin{aligned} \phi_X(t) &= E(e^{itVZ}) = E(e^{itVZ} |_{V=1}) + E(e^{itVZ} |_{V=-1}) \\ &= P(V=1) E(e^{itZ}) + P(V=-1) E(e^{-itZ}) = \\ &= \frac{1}{2} e^{-t^2/2} + \frac{1}{2} e^{-(-t)^2/2} = e^{-t^2/2} = \phi_Z(t). \end{aligned}$$

b) Find $\text{Cor}(X, Z)$

Answer. $\text{Cor}(X, Z) = E(XZ) = E(VZ^2) = E(V)E(Z^2) = 0$.

c) Are X, Z independent?

Answer. $P(X=Z) = P(VZ=Z) = P(V=1) = \frac{1}{2} > 0$.
No.

d) Are $X = VZ$ and V independent?

Answer

$$\left. \begin{aligned} E(e^{isVZ} | V=1) &= e^{-s^2/2} \\ E(e^{isVZ} | V=-1) &= e^{-s^2/2} \end{aligned} \right\} E(e^{isVZ} | V) = e^{-s^2/2}$$

Joint cf: $E(e^{(isVZ + it)V}) = E[e^{itV} E(e^{isVZ} | V)] = e^{-s^2/2} E(e^{itV}) = E(e^{isVZ}) E(e^{itV})$. Yes.

Ex 3. Let $Y_n \sim \text{geometric}(p = \frac{1}{n})$: $G_{Y_n}(s) = \frac{ps}{1-ps}$
 $P(Y_n = k) = q^{k-1} p, k=1, 2, \dots$

Show that $\frac{Y_n}{n} \xrightarrow{D} Y \sim \text{exponential}(\lambda)$

Answer. $\phi_{Y_n}(t) = \frac{pe^{it}}{1-(1-p)e^{it}} = \frac{pe^{it}}{e^{it}(p-(1-e^{-it}))} = \frac{\frac{\lambda}{n}}{\frac{\lambda}{n} + (e^{-it/n} - 1)}$

$\phi_{\frac{Y_n}{n}}(t) = \phi_{Y_n}\left(\frac{t}{n}\right) = \frac{\lambda}{\lambda + n(e^{-it/n} - 1)} = \frac{\lambda}{\lambda + n(e^{-it/n} - 1)}$

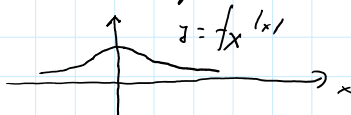
$M_Y(t) = \frac{\lambda}{\lambda - t}, \quad \phi_Y(t) = \frac{\lambda}{\lambda - it}$

We find limit of $n(e^{-it/n} - 1) = \frac{e^{-it/n} - 1}{\frac{it}{n}} \cdot it =$
 $\left(z = \frac{it}{n}\right) \rightarrow -2-1$ L'Hospital $-it$

We find limit of $n(e^{it/n} - 1) = \frac{it}{n} \dots =$
 $\lim_{n \rightarrow \infty} \frac{e^{-it/n} - 1}{-it/n} \xrightarrow{\text{L'Hospital}} -it \text{ as } z \rightarrow 0$

Hence $\phi_{Y_n}(t) \rightarrow \phi_Y(t) \text{ as } n \rightarrow \infty.$

Ex 4. Let X_1, \dots, X_n be independent Cauchy

Recall $f_X(x) = \frac{1}{\pi} \cdot \frac{1}{1+x^2}, -\infty < x < \infty$ 

$\phi_X(t) = e^{-|t|}, t \in \mathbb{R}.$

a) Find $\phi_{\bar{X}_n}(t)$

Answer. Recall $\phi_{\bar{X}_n}(t) = \phi_X\left(\frac{t}{n}\right)^n = \left(e^{-|t/n|}\right)^n = e^{-|t|} = \phi_X(t).$

b) Is there a constant c so that $\bar{X}_n \rightarrow c$ in probability?

Answer. No: $\bar{X}_n \rightarrow c \text{ in prob.} \Leftrightarrow \phi_{\bar{X}_n}(t) \rightarrow e^{itc}.$

Ex 5. Let $X \sim N(0, \sigma^2), Y \sim N(aX, \tilde{\sigma}^2) \mid \text{given } X=x,$

$Y \sim N(a_x, \tilde{\sigma}^2).$

a) Find $E(e^{itY} \mid X), \text{Cor}(X, Y)$
Answer. $E(e^{itY} \mid X=x) = e^{ita_x} e^{-\tilde{\sigma}^2 t^2 / 2}$
 $E(e^{itY} \mid X) = e^{itaX} e^{-\tilde{\sigma}^2 t^2 / 2}$ $\left. \begin{array}{l} \text{Cor}(X, Y) = E(XY) \\ = E[X E(Y \mid X)] = \\ = E(aX^2) = a\sigma^2. \end{array} \right\}$

b) Find $\phi_Y(t)$. How is Y distributed

Answer. $\phi_Y(t) = E[E(e^{itY} \mid X)] = e^{-\tilde{\sigma}^2 t^2 / 2} E(e^{itaX})$
 $= \exp\left\{-\frac{t^2}{2} (\tilde{\sigma}^2 + a^2 \sigma^2)\right\}: Y \sim N(0, \tilde{\sigma}^2 + a^2 \sigma^2)$

c) Is X, Y normal bivariate?

Answer. joint cf:

$$\begin{aligned}\phi(s, t) &= E(e^{isX + itY}) = E\left[e^{isX} E(e^{itY} | X)\right] \\ &= e^{-\tilde{\sigma}^2 t^2 / 2} E(e^{i(s+ta)X}) = e^{-\tilde{\sigma}^2 t^2 / 2} e^{-(s+ta)^2 \sigma^2 / 2} \\ &= \exp\left\{-\frac{1}{2} \left[t^2 (\tilde{\sigma}^2 + a^2 \sigma^2) - 2sta\sigma^2 + s^2 \frac{\sigma^2}{2} \right]\right\}\end{aligned}$$

Yes: $B = \begin{pmatrix} \sigma^2 & a\sigma^2 \\ a\sigma^2 & \tilde{\sigma}^2 + a^2\sigma^2 \end{pmatrix}$ is covariance matrix.

In general $X = (X_1, X_2) \sim N(\mu, B)$ if

$$\phi_X(t) = \exp\left\{i t \mu' - \frac{1}{2} t B t'\right\} \quad , t = (t_1, t_2) \in \mathbb{R}^2$$

$$B = \left(\text{Cov}(X_i, X_j) \right)_{1 \leq i, j \leq 2}.$$