

④ (a) for $E(X)$ is finite & $E(X^2)$ is not finite

$E(X) = \sum_{\forall x} P(X=x) \cdot x$ to find
we need $P(X=x)$ such that
 $E(X) = \sum_{\forall x} P(X=x) \cdot x < \infty$ &

$E(X^2) = \sum_{\forall x} P(X=x) \cdot x^2$ is not finite

atleast, $\Rightarrow P(X=x)$ is the form of $\frac{1}{x^3}$

~~$\Rightarrow P(X=x) = \frac{1}{x^3}$~~ $\Rightarrow P(X=x)$ can be $\frac{1}{z} \cdot \frac{1}{x^{1+\epsilon}}$ $\epsilon \in (1, 2)$

proved here z is req. const, such that $\frac{1}{z} \sum_{\forall x} P(X=x) = 1$

(b) $E(X)$ is finite & $E(X^2)$ is not finite

we need to find $f(x)$ such that

$$E(X) = \int_{\forall x} f(x) \cdot x dx < \infty \text{ &}$$

$$E(X^2) = \int_{\forall x} f(x) \cdot x^2 dx \text{ is not finite}$$

atleast $P(X=f(x)) \approx \frac{1}{x^3}$

$\Rightarrow f(x)$ can be $\frac{1}{z} \cdot \frac{1}{x^{1+\epsilon}}$ $\epsilon \in (1, 2)$
where z is req. const, such that

$$\frac{1}{z} \int_{\forall x} f(x) dx = 1$$

proved

(c) e^{-x} is convex,

Jensen inequality says

$$E(e^{-x}) \geq e^{-E(x)} = e^{-1} = \frac{1}{e}$$

$$E(e^{-x}) \geq \frac{1}{e} > \frac{1}{3}$$

so we cannot have $E(e^{-x}) < \frac{1}{3}$

if $E(x) = 1$

also disproved

⑥ Let ~~X, Y be a uniformly~~
 Let X, Y be random variables which are independent & uniformly distributed over $[0, d]$

$$\Rightarrow f(x) = \begin{cases} \frac{1}{d}, & 0 \leq x \leq d \\ 0, & \text{o.w} \end{cases} \quad \text{||} \quad f(y) = \begin{cases} \frac{1}{d}, & 0 \leq y \leq d \\ 0, & \text{o.w} \end{cases}$$

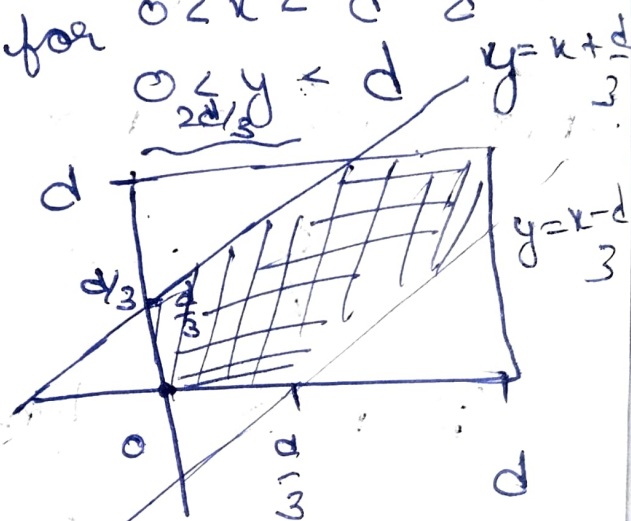
we want $P(|X-Y| < \frac{d}{3})$

$|X-Y|$ give $\frac{d}{3}$ line segment

$$f(x, y) = \frac{1}{d^2} \quad \text{for } 0 < x < d \text{ \& } 0 < y < d$$

$$|X-Y| < \frac{d}{3}$$

$$y = x + \frac{d}{3} \quad y = x - \frac{d}{3}$$



area of shaded part

$$= d^2 - \frac{1}{2} \left(\frac{2d}{3} \right) \left(\frac{2d}{3} \right) - \frac{1}{2} \left(\frac{2d}{3} \right) \left(\frac{2d}{3} \right)$$

$$= \frac{5d^2}{9}$$

probability = volume below of shaded part below $f(x, y)$

$$= \frac{1}{d^2} \cdot \frac{5d^2}{9} = \frac{5}{9}$$

$$\Rightarrow \text{req prob} = \frac{5}{9}$$

④ by def

$$H(x) = \int_{-\infty}^{\infty} F(x-y) G(y) dy$$

$H(x)$ should be \rightarrow ① non decreasing
② right continuous

③ $\lim_{x \rightarrow -\infty} H(x) = 0$ & $\lim_{x \rightarrow \infty} H(x) = 1$

① for $x_1 < x_2$

F is also non decreasing $\Rightarrow F(x_1 - y) < F(x_2 - y)$

$$\Rightarrow H(x_1) < H(x_2)$$

② $\lim_{x \rightarrow x_0^+} H(x) = \int_{-\infty}^{\infty} \lim_{x \rightarrow x_0^+} F(x-y) G(y) dy$

It can be taken inside \int & F is right continuous

$$= H(x_0)$$

③ $\lim_{x \rightarrow -\infty} H(x) = \int_{-\infty}^{\infty} \lim_{x \rightarrow -\infty} F(x-y) G(y) dy = \int_{-\infty}^{\infty} 0 G(y) dy = 0$

$$\Rightarrow \lim_{x \rightarrow -\infty} H(x) = 0$$

$$\lim_{x \rightarrow +\infty} H(x) = \int_{-\infty}^{\infty} \lim_{x \rightarrow +\infty} F(x-y) G(y) dy = \int_{-\infty}^{\infty} 1 \cdot G(y) dy = 1$$

$$\Rightarrow \lim_{x \rightarrow +\infty} H(x) = 1$$

$\Rightarrow H(x)$ is also a

distributive function

Similarly can be prove for discrete case too

$$(ii) (i) E(e^{ux}) = e^{ux + \frac{1}{2} u^2 \sigma^2}$$

$E(e^{ux})$ is moment generating function

$$\Rightarrow E(e^{ux}) = \int_{-\infty}^{\infty} \frac{1}{\sigma\sqrt{2\pi}} e^{ux} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx$$

$$\frac{x-\mu}{\sigma} = z$$

$$\Rightarrow \frac{dx}{\sigma} = dz$$

$$\Rightarrow E(e^{ux}) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{(u+\sigma z)\mu - \frac{z^2}{2}} dz$$

$$E(e^{ux}) = \frac{e^{u\mu}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(z^2 - 2\sigma z u + (\sigma u)^2)} dz$$

$$E(e^{ux}) = \frac{e^{u\mu + \sigma^2 \frac{u^2}{2}}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(z - \sigma u)^2} dz$$

$$E(e^{ux}) = \frac{e^{u\mu + \sigma^2 \frac{u^2}{2}}}{e} \quad (\text{proved})$$

no need to prove it is just MGF of $N(\mu, \sigma^2)$

(ii) $E[\phi(x)] \geq \phi[E(x)]$ Jensen's inequality

$$E[e^{ux}] \geq e^{u[E(x)]} \geq e^{u\mu}$$

$$\Rightarrow e^{u\mu + \frac{1}{2} u^2 \sigma^2} \geq e^{u\mu}$$

$\sigma > 0$ we have

$$E[e^{ux}] \geq e^{u[E(x)]} \quad (\text{inequality holds})$$