

①

$$\Phi_x(\omega) = e^{jm\omega - \frac{\sigma^2 \omega^2}{2}}$$

$$N(m, \sigma^2) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-m)^2}{2\sigma^2}}$$

$$\Phi_x(\omega) = \mathbb{E}_x[e^{j\omega x}] = \int_{-\infty}^{\infty} \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-m)^2}{2\sigma^2}} \cdot e^{j\omega x} dx$$

$$= \frac{1}{\sigma \sqrt{2\pi}} \int_{-\infty}^{\infty} e^{j\omega x - \frac{(x-m)^2}{2\sigma^2}} dx$$

hint ①

$$= \frac{1}{\sigma \sqrt{2\pi}} \int_{-\infty}^{\infty} e^{\frac{-(x-k)^2 + 2mj\omega\sigma^2 - \omega^2\sigma^4}{2\sigma^2}} dx \text{ where } k = m + j\omega\sigma^2$$

$$= e^{\frac{2mj\omega\sigma^2 - \omega^2\sigma^4}{2\sigma^2}} \underbrace{\int_{-\infty}^{\infty} \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-k)^2}{2\sigma^2}} dx}_{k = m + j\omega\sigma^2}$$

Gaussian w/ mean $m + j\omega\sigma^2$
var σ^2

$$\text{so } \int_{-\infty}^{\infty} N(m + j\omega\sigma^2, \sigma^2) = 1$$

$$= e^{\frac{2mj\omega\sigma^2 - \omega^2\sigma^4}{2\sigma^2}}$$

$$= e^{\frac{2mj\omega\sigma^2}{2\sigma^2} - \frac{\omega^2\sigma^4}{2\sigma^2}}$$

$$\boxed{\Phi_x(\omega) = e^{(jm\omega - \frac{\omega^2\sigma^2}{2})}}$$

2.

$$Y = aX + b$$

$$\frac{Y-b}{a} = X$$

$$\frac{\partial X}{\partial Y} = \frac{1}{a}$$

$$f_Y(y) = f_X\left(\frac{Y-b}{a}\right) \cdot \left|\frac{\partial X}{\partial Y}\right|$$

$$= f_X\left(\frac{Y-b}{a}\right) \cdot \left|\frac{1}{a}\right|$$

$$\Phi_Y(\omega) = \mathbb{E}[e^{j\omega Y}]$$

$$= \mathbb{E}[e^{j\omega(ax+b)}]$$

$$= \mathbb{E}[e^{j\omega ax + j\omega b}]$$

$$= e^{j\omega b} \mathbb{E}[e^{j\omega ax}]$$

$$= e^{j\omega b} \int_{-\infty}^{\infty} e^{aj\omega x} \cdot \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-m)^2}{2\sigma^2}} dx$$

$$= \frac{e^{j\omega b}}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{aj\omega x - \frac{(x-m)^2}{2\sigma^2}} dx$$

$$\frac{e^{j\omega b}}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{aj\omega x - \frac{x^2 + 2xm - m^2}{2\sigma^2}} dx$$

$$\frac{e^{j\omega b}}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{\frac{2\sigma^2 aj\omega x - x^2 + 2xm - m^2}{2\sigma^2}} dx$$

$$\frac{e^{j\omega b}}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{\frac{-x^2 + 2x(\sigma^2 aj\omega + m) - m^2}{2\sigma^2}} dx$$

$$\frac{e^{j\omega b}}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{\frac{-x^2 + 2x(\sigma^2 aj\omega + m) - (\sigma^2 aj\omega + m)^2 + (\sigma^2 aj\omega + m)^2 - m^2}{2\sigma^2}} dx$$

$$\frac{e^{j\omega b}}{\sqrt{2\pi}\sigma} \cdot e^{\frac{(\sigma^2 aj\omega + m)^2 - m^2}{2\sigma^2}} \int_{-\infty}^{\infty} e^{\frac{-x^2 + 2x(\sigma^2 aj\omega + m) - (\sigma^2 aj\omega + m)^2}{2\sigma^2}} dx$$

$$e^{j\omega b} \cdot e^{\frac{(\sigma^2 aj\omega + m)^2 - m^2}{2\sigma^2}} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma} e^{\frac{-(x - [\sigma^2 aj\omega + m])^2}{2\sigma^2}} dx$$

$$e^{j\omega b} \cdot e^{\frac{-\sigma^4 a^2 \omega^2 + 2\sigma^2 aj\omega m + m^2 - m^2}{2\sigma^2}} \text{ Gaussian}$$

$$= e^{j\omega b + aj\omega m - \frac{\sigma^2 a^2 \omega^2}{2}} = \boxed{e^{j\omega(b+m) - \frac{\sigma^2 a^2 \omega^2}{2}} = \Phi_Y(\omega)}$$

You may conclude if $Y = aX + b$ and X is gaussian then

$$\begin{aligned} m_Y &= m_X + b & \text{and} & \quad \sigma_Y = \sigma_X a \\ \sigma_Y^2 &= \sigma_X^2 a^2 \end{aligned}$$

3)

a.

$$E[X] = \int p(x) x \, dx$$

:

$$= \int [p(x|X \geq a) + p(x|X < a)] x \, dx$$

$$= \int [p(x|X \geq a)p(X \geq a) + p(x|X < a)p(X < a)] x \, dx$$

$$= \int p(x|X \geq a) x \, dx p(X \geq a) + \int p(x|X < a) x \, dx p(X < a)$$

$$E[X] = E[X|X \geq a]p(X \geq a) + E[X|X < a]p(X < a)$$

b.

Markov inequality proof

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

$$E[X] = E[X|X < a]p(X < a) + E[X|X \geq a]p(X \geq a)$$

$$\frac{E[X] - \overbrace{E[X|X < a]p(X < a)}^{> 0}}{E[X|X \geq a]} = p(X \geq a)$$

$$E[X|X \geq a]$$

$$\frac{E[X]}{E[X|X \geq a]} \geq p(X \geq a)$$

$$E[X|X \geq a]$$

$$E[X|X \geq a] \geq a$$

so

$$\frac{E[X]}{a} \geq p(X \geq a)$$

$E[X|X < a]p(X < a) > 0$ by definition of problem $X > 0$ and $a > 0$ and $p(\cdot) \geq 0$

divide by less than before gives bigger number!

④

$$Y = \frac{X}{n} \quad X \text{ is Binomial w/ } n \text{ trials}$$

$$a=1$$

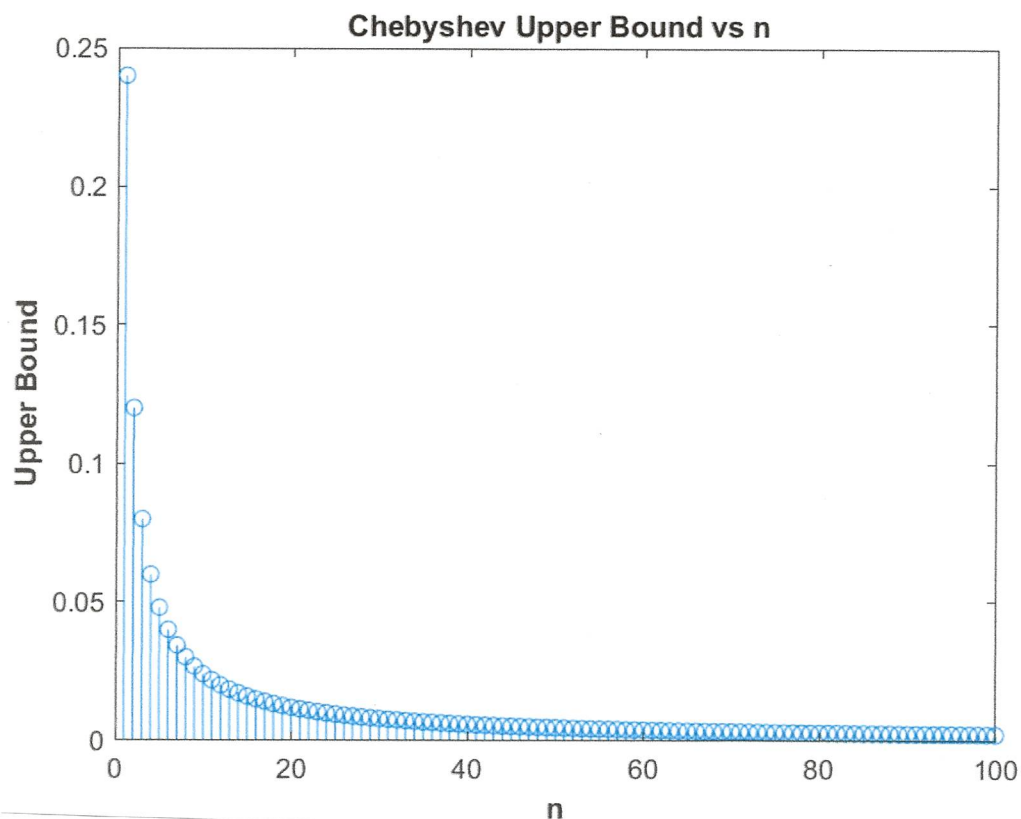
$$E[Y] = \frac{np}{n} = p$$

$$\text{Var}[Y] = \frac{1}{n^2} \cdot np(1-p) = \frac{1}{n}(1-p)p$$

Chebyshev bound

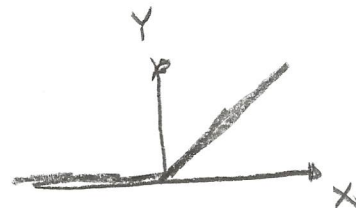
$$P(|Y-p| > a) \leq \left(\frac{1}{n}(1-p)p \right)$$

as $n \rightarrow \infty$ then the bound $\frac{1}{n}(1-p)p \rightarrow 0$



(3)

(a)



for $y \geq 0$

$$f_Y(y) = \left| \frac{dx}{dy} \right| f_X(x)$$

$$f_Y(y) = \frac{1}{\sqrt{2\pi} \cdot 2} e^{-\frac{(y-2)^2}{8}} \quad y > 0$$

for $y < 0$

$$\Phi\left(\frac{0-2}{2}\right)$$

$$f_Y(y) = u(y) \cdot \frac{1}{2\sqrt{2\pi}} e^{-\frac{(y-2)^2}{8}} + \delta(\max(y, 0)) \Phi(-1)$$

step fn

delta fn

b) Y is not a continuous random variable due to the delta function... It is a mixed random variable.