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consider RV's X_1, X_2, \dots, X_n iid but PMF/PDF is unknown but we know it has expectation μ and variance σ^2

$$\text{let } T_n = X_1 + X_2 + \dots + X_n$$

$$\text{let } \bar{X}_n = \frac{T_n}{n} = \frac{X_1 + X_2 + \dots + X_n}{n}$$

From 4.21, we know $E[\bar{X}_n] = \mu$, $\text{Var}[\bar{X}_n] = \frac{\sigma^2}{n}$

$$\text{let } Z_n = \frac{\bar{X}_n - \mu}{\frac{\sigma}{\sqrt{n}}} = \frac{\sqrt{n}}{\sigma} \bar{X}_n + \frac{-\sqrt{n}}{\sigma} \mu$$

"X standardization" $E[Z_n] = 0$

$$\text{Var}[Z_n] = 1 = \text{SD}[Z_n]$$

$$\phi_{T_n}(t) \stackrel{(P_3)}{=} \phi_{X_1}(t) \phi_{X_2}(t) \dots \phi_{X_n}(t) \stackrel{(P_1)}{=} \phi_X(t)^n$$

$$\phi_{\bar{X}_n}(t) \stackrel{(P_3)}{=} \phi_{\frac{1}{n} T_n}(t) \stackrel{(P_2)}{=} \phi_X\left(\frac{t}{n}\right)^n$$

$$\phi_{Z_n}(t) \stackrel{(P_2)}{=} e^{itb} \phi_{\bar{X}_n}(at) = e^{\frac{-it\mu\sqrt{n}}{\sigma}} \phi_X\left(\frac{\sqrt{n}}{\sigma} t\right)^n$$

$$= e^{\frac{-it\mu n}{\sigma\sqrt{n}}} e^{\ln\left(\phi_X\left(\frac{t}{\sigma\sqrt{n}}\right)^n\right)} = e^{\frac{-it\mu n}{\sigma\sqrt{n}} + n \ln\left(\phi_X\left(\frac{t}{\sigma\sqrt{n}}\right)\right) \frac{1}{n}}$$

$$= e^{\frac{-it\mu}{\sigma\sqrt{n}} + \ln\left(\phi_X\left(\frac{t}{\sigma\sqrt{n}}\right)\right) \cdot \frac{t^2}{\frac{t^2}{\sigma^2}} = \frac{t^2}{\sigma^2} \left(\frac{\ln\left(\phi_X\left(\frac{t}{\sigma\sqrt{n}}\right)\right) - \frac{-it\mu}{\sigma\sqrt{n}}}{\frac{t^2}{\sigma^2 n}} \right)} = \phi_{Z_n}(t)$$

We want to investigate now $\lim_{n \rightarrow \infty} \phi_{Z_n}(t) = ?$

$$= e^{\frac{t^2}{2\sigma^2} \lim_{n \rightarrow \infty} \frac{\ln(\phi_X(\frac{t}{\sigma\sqrt{n}})) - \frac{it\mu}{\sigma\sqrt{n}}}{\frac{t^2}{\sigma^2 n}}} \rightarrow e^{\frac{t^2}{2\sigma^2} \lim_{u \rightarrow 0} \frac{\ln(\phi_X(u)) - i\mu u}{u^2}}$$

let $u = \frac{t}{\sigma\sqrt{n}} \Rightarrow n \rightarrow \infty \Rightarrow u \rightarrow 0$

1) L'Hôpital's \downarrow

$$= e^{\frac{t^2}{2\sigma^2} \lim_{u \rightarrow 0} \frac{\frac{\phi_X'(u)}{\phi_X(u)} - i\mu}{u}} = e^{\frac{t^2}{2\sigma^2} \lim_{u \rightarrow 0} \frac{\phi_X'(u)\phi_X''(u) - \phi_X'(u)^2}{\phi_X(u)^2}}$$

$$= e^{\frac{t^2}{2\sigma^2} \frac{\phi_X'(0)\phi_X''(0) - \phi_X'(0)^2}{\phi_X(0)}} \stackrel{(P_6)}{=} e^{\frac{t^2}{2\sigma^2} (\phi_X''(0) - \phi_X'(0)^2)}$$

$$\stackrel{(P_7)}{=} e^{\frac{t^2}{2\sigma^2} (1^2 E[X^2] - (iE[X])^2)} = e^{\frac{t^2}{2\sigma^2} (E[X^2] - E[X]^2)} = e^{-\frac{t^2}{2}} = \phi_Z(t)$$

$\stackrel{(P_8)}{\Rightarrow} Z_n \xrightarrow{d} Z$ where Z has chf $\phi_Z(t) = e^{-\frac{t^2}{2}}$.

$$Z \sim f_Z(Z) = ?$$

use (P_6) to find PDF of Z , check $\phi_Z(t) \in L^1 \Rightarrow$

$$\Rightarrow \int_{\mathbb{R}} e^{-t^2/2} dt = \sqrt{\pi} < \infty \quad \text{yes!}$$

Gaussian Integrl

$$f_Z(z) = \frac{1}{2\pi} \int_{\mathbb{R}} e^{-it^2} \phi_Z(t) dt$$

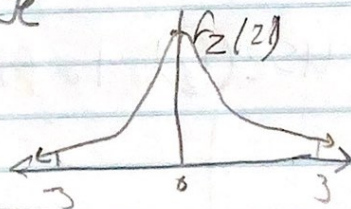
$$= \frac{1}{2\pi} \int_{\mathbb{R}} e^{-it^2} e^{-\frac{t^2}{2}} dt = \frac{1}{2\pi} \int_{\mathbb{R}} e^{-(it^2 + \frac{t^2}{2})} dt$$

$$\begin{aligned} \frac{t^2}{2} + itz &= \left(\frac{t}{\sqrt{2}} + \frac{\sqrt{2}it}{2} \right)^2 - \left(\frac{\sqrt{2}it}{2} \right)^2 \\ &= \frac{t^2}{2} + 2 \frac{\sqrt{2}it}{2} \frac{t}{\sqrt{2}} + \frac{i^2 z^2}{2} - \frac{i^2 z^2}{2} \\ &= \frac{1}{2\pi} \int_R e^{-\left(\frac{t}{\sqrt{2}} + \frac{\sqrt{2}it}{2}\right)^2} e^{-\frac{z^2}{2}} dt = \frac{1}{2\pi} e^{-\frac{z^2}{2}} \int_R e^{-\left(\frac{t}{\sqrt{2}} + \frac{\sqrt{2}it}{2}\right)^2} dt \end{aligned}$$

$$\text{let } y = \frac{t}{\sqrt{2}} + \frac{\sqrt{2}it}{2} \Rightarrow \frac{dy}{dt} = \frac{1}{\sqrt{2}}, \quad t \rightarrow \infty \Rightarrow y \rightarrow \infty, \quad t \rightarrow -\infty \Rightarrow y \rightarrow -\infty$$

$$\begin{aligned} &= \frac{1}{2\pi} e^{-\frac{z^2}{2}} \int_R e^{-y^2} \sqrt{2} dy \stackrel{\text{Gaussian Integral}}{=} \frac{1}{2\pi} e^{-\frac{z^2}{2}} \sqrt{2} \sqrt{\pi} = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} \\ &\quad \text{standard normal} = N(0,1) \end{aligned}$$

This completes the proof of the "central limit theorem" (CLT), the crown jewel of a basic probability class, one of the most useful results that probability has given to the world at large



$$\boxed{\text{CLT} = X_1, \dots, X_n \text{ i.i.d. } \mu, \sigma^2, \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} \xrightarrow{d} N(0,1)}$$

$$\begin{aligned} &\text{let } \sigma > 0 \\ &Z \sim N(0,1), \quad X = \mu + \sigma Z \sim f_X(x) = ? \\ &\left. \begin{aligned} \phi_X(t) &\stackrel{P2}{=} e^{it\mu} \phi_Z(\sigma t) \\ &= e^{it\mu - \sigma^2 t^2/2} \end{aligned} \right\} \end{aligned}$$

$$f_X(x) = \frac{1}{\sigma} \phi\left(\frac{x-\eta}{\sigma}\right) = \frac{1}{\sigma} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\eta}{\sigma}\right)^2} = \frac{1}{\sqrt{2\pi}\sigma^2} e^{-\frac{1}{2\sigma^2}(x-\eta)^2} = N(\eta, \sigma^2)$$

$$E[Z] = \frac{\phi'_2(0)}{i} = 0, \quad \text{Var}[Z] = E[Z^2] - E[Z]^2$$

$$\phi'_2(t) = \frac{d}{dt} [e^{-t^2/2}] = -t e^{-t^2/2}$$

$$\phi''_2(t) = \frac{d}{dt} [-t e^{-t^2/2}] = -[e^{-t^2/2} - t^2 e^{-t^2/2}] = t^2 e^{-t^2/2} - e^{-t^2/2}$$

$$= \frac{\phi''_2(0)}{i^2} = 1$$

$$E[X] = E[\eta + \sigma Z] = \eta, \quad \text{Var}[X] = \text{Var}[\eta + \sigma Z] = \sigma^2, \quad \text{SD}[X] = \sigma$$

$$X_1 \sim N(\eta_1, \sigma_1^2) \text{ indep of } X_2 \sim N(\eta_2, \sigma_2^2), \quad T = X_1 + X_2 \sim f_T(t) = ?$$

$$\phi_T(t) \stackrel{(P_2)}{=} \phi_{X_1}(t) \phi_{X_2}(t) = e^{it\eta_1 - \sigma_1^2 t^2/2} e^{it\eta_2 - \sigma_2^2 t^2/2}$$

$$= e^{it(\eta_1 + \eta_2) - (\sigma_1^2 + \sigma_2^2)t^2/2} \stackrel{(P_1)}{\Rightarrow} T \sim N(\eta_1 + \eta_2, \sigma_1^2 + \sigma_2^2)$$

$$X \sim N(\eta, \sigma^2), \quad Y = e^X \sim f_Y(y) = ?$$

$$g'(y) = \ln(y) \quad \left| \frac{d}{dy} (g^{-1}(y)) \right| = \frac{1}{|y|}$$

$$f_Y(y) = f_X(\ln y) \frac{1}{|y|} = \frac{1}{\sqrt{2\pi}\sigma^2} e^{-\frac{1}{2\sigma^2}(\ln(y)-\eta)^2} \frac{1}{|y|} = \frac{1}{\sqrt{2\pi}\sigma^2 y^2} e^{-\frac{1}{2\sigma^2}(\ln(y)-\eta)^2}$$

$$= \frac{1}{\sqrt{2\pi}\sigma^2 y^2} e^{-\frac{1}{2\sigma^2}(\ln(y)-\eta)^2}$$

$$= \text{Log-N}(\eta, \sigma^2)$$

Log-Normal distribution