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consider $\Phi'_x(t) = \frac{d}{dt} (E(e^{itx}))$

$$= \frac{d}{dt} \left[\int_{\mathbb{R}} e^{itx} f(x) dx \right] \stackrel{?}{=} \int_{\mathbb{R}} f(x) \frac{d}{dt} (e^{itx}) dx$$

does $\frac{d}{dt} \left(\int g(x, t) dx \right) \stackrel{?}{=} \int_{\mathbb{R}} \frac{\partial}{\partial t} [g(x, t)] dx$

conditions:

a) $\exists t \in A$ s.t. $\int_{\mathbb{R}} g(x, t) dx$ converges. $A = [a, b] \subset \mathbb{R}$

b) $g(x, t)$ continuous $\forall t \in A$

c) $g(x, t)$ continuous $\forall x \in \mathbb{R}$

d) $\forall t \in A$ $\int_{\mathbb{R}} \frac{\partial}{\partial t} g(x, t) dt$ converges uniformly.

$$\Phi'_x(t) = \int_{\mathbb{R}} f(x) ix e^{itx} dx$$

consider

$$\Phi'_x(0) = \int_{\mathbb{R}} f(x) ix dx = i \int_{\mathbb{R}} x f(x) dx = i E(x)$$

$$\Phi''_x(0) = \int_{\mathbb{R}} f(x) i^2 x^2 e^{itx} dx$$

$$\Phi''_x(0) = i^2 \int_{\mathbb{R}} x^2 f(x) dx = i^2 E(x^2)$$

$$\Phi'''_x(0) = i^3 E(x^3) \Rightarrow \left\{ \begin{aligned} E(x^n) &= \frac{\Phi^{(n)}_x(0)}{i^n} \end{aligned} \right.$$



$$\textcircled{6} P(X \in (a, b)) = \frac{1}{2\pi} \int_{\mathbb{R}} \frac{e^{-it a} - e^{-it b}}{it} \phi_X(t) dt$$

Motivation of $\phi_X \in L^1$

$$\Rightarrow f(x) = \frac{1}{2\pi} \int_{\mathbb{R}} e^{-itx} \phi_X(t) dt$$

$$P(X \in (a, b)) = \int_a^b f(x) dx = \int_a^b \frac{1}{2\pi} \int_{\mathbb{R}} e^{-itx} \phi_X(t) dt dx$$

$$= \frac{1}{2\pi} \int_{\mathbb{R}} \left(\int_a^b e^{-itx} dx \right) \phi_X(t) dt$$

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

$$\textcircled{7} \phi_X(t) = \phi_Y(t) \Leftrightarrow X \stackrel{d}{=} Y$$

$\textcircled{8}$ $\phi_{X_n}(t)$ is the ch. f for X_n

If $\lim_{n \rightarrow \infty} \phi_{X_n}(t) = \phi_X(t)$

$$\Rightarrow \lim_{n \rightarrow \infty} F_{X_n}(x) = F_X(x) \text{ also } \lim_{n \rightarrow \infty} X_n \stackrel{d}{=} X$$

or $X_n \xrightarrow{d} X$ convergence in distribution / law.

$$X \sim \text{Gamma}(k, \lambda) \quad \phi(t) = \int_0^{\infty} e^{itx} \frac{\lambda^k e^{-\lambda x} x^{k-1}}{\Gamma(k)} dx$$

$$= \frac{\lambda^k}{\Gamma(k)} \int_0^{\infty} x^{k-1} e^{(it-\lambda)x} dx$$

$$\text{let } u = (it - \lambda)x \rightarrow x = \frac{1}{it - \lambda} u$$

$$\text{let } u = (\lambda - it)x \Rightarrow x = \frac{1}{\lambda - it} u \Rightarrow dx = \frac{1}{\lambda - it} du$$

$$= \frac{\lambda^k}{\Gamma(k)} \int_0^{\infty} \frac{u^{k-1}}{(\lambda - it)^{k-1}} e^{-u} \frac{1}{\lambda - it} du$$

$$= \frac{\lambda^k}{\Gamma(k)(\lambda - it)^k} \int_0^{\infty} u^{k-1} e^{-u} du = \frac{\lambda^k}{\Gamma(k)(\lambda - it)^k} \cdot \Gamma(k)$$

$$= \left(\frac{\lambda}{\lambda - it} \right)^k = \left(1 - \frac{it}{\lambda} \right)^{-k}$$

$$X_1 \sim \text{Gamma}(k_1, \lambda) \text{ ind of } X_2 \sim \text{Gamma}(k_2, \lambda)$$

$$X_1 + X_2 \sim \text{Gamma}(k_1 + k_2, \lambda)$$

$$\phi_{X_1 + X_2}(t) = \phi_{X_1}(t) \cdot \phi_{X_2}(t) = \left(\frac{\lambda}{\lambda - it} \right)^{k_1} \left(\frac{\lambda}{\lambda - it} \right)^{k_2}$$

$$= \left(\frac{\lambda}{\lambda - it} \right)^{k_1 + k_2} = \text{Gamma}(k_1 + k_2, \lambda)$$

$X \sim \text{Poisson}(\lambda)$

PMF of Poisson $(\lambda e^{-\lambda})$

$$\phi_X(t) = \sum_{x=0}^{\infty} e^{itx} \frac{\lambda^x e^{-\lambda}}{x!} = \frac{e^{-\lambda}}{e^{-\lambda e^{it}}} \sum_{x=0}^{\infty} \frac{(\lambda e^{it})^x e^{-\lambda e^{it}}}{x!}$$

$$\frac{(e^{it})^x \lambda^x e^{-\lambda}}{x!} = \frac{(\lambda e^{it})^x e^{-\lambda e^{it}}}{x!} \cdot \frac{e^{-\lambda e^{it}}}{e^{-\lambda e^{it}}} = e^{-\lambda e^{it} + \lambda} = e^{\lambda(e^{it} - 1)}$$

$X_1 \sim \text{Poisson}(\lambda_1)$ and $X_2 \sim \text{Poisson}(\lambda_2)$

$X_1 + X_2 \sim \text{Poisson}(\lambda_1 + \lambda_2)$

$$\phi_{X_1+X_2}(t) = \phi_{X_1}(t) \cdot \phi_{X_2}(t) = e^{\lambda_1(e^{it} - 1)} \cdot e^{\lambda_2(e^{it} - 1)}$$

$$= e^{(\lambda_1 + \lambda_2)(e^{it} - 1)}$$

X_1, \dots, X_n iid same distribution with finite mean μ & finite variance σ^2

$$\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i \quad E(\bar{X}) = \mu, \quad \text{Var}(\bar{X}) = \frac{\sigma^2}{n}$$

$$\text{Define } Z_n = \frac{\bar{X}_n - \mu}{\frac{\sigma}{\sqrt{n}}} \quad E(Z_n) = 0, \quad \text{Var}(Z_n) = 1$$

Standardization ↓
E(Z_n)

$$\phi_x(t) \stackrel{\#2}{\geq} \phi_{\frac{t}{\sqrt{n}}} \stackrel{\#3}{\geq} \left(\phi_x\left(\frac{t}{\sqrt{n}}\right) \right)^n$$

$$\phi_x(t) \stackrel{\#3}{\geq} \phi_{\frac{t}{\sqrt{n}}} e^{it\left(\frac{-n}{\sqrt{n}}\right)} = \phi_{\frac{t}{\sqrt{n}}} e^{\frac{-itun}{\sqrt{n}}}$$

$$= \phi_{\frac{t}{\sqrt{n}}} e^{\frac{-itun}{\sqrt{n}}} = \left(\phi_x\left(\frac{t}{\sqrt{n}}\right) \right)^n e^{\frac{-itun}{\sqrt{n}}}$$

$$\lim_{n \rightarrow \infty} \phi_x(t) \geq \lim_{n \rightarrow \infty} \left(\phi_x\left(\frac{t}{\sqrt{n}}\right) \right)^n e^{\frac{-itun}{\sqrt{n}}} = \lim_{n \rightarrow \infty} e^{n \ln \left(\phi_x\left(\frac{t}{\sqrt{n}}\right) \right) - \frac{itun}{\sqrt{n}}}$$

$$= \lim_{n \rightarrow \infty} e^{n \ln \left(\phi_x\left(\frac{t}{\sqrt{n}}\right) \right) - \frac{itun}{\sqrt{n}}}$$

$$= \lim_{n \rightarrow \infty} e^{n \ln \left(\phi_x\left(\frac{t}{\sqrt{n}}\right) \right) - \frac{itun}{\sqrt{n}}}$$

$$= \lim_{n \rightarrow \infty} e^{n \left(\ln \left(\phi_x\left(\frac{t}{\sqrt{n}}\right) \right) - \frac{itun}{\sqrt{n}} \right)}$$

$$= e^{\lim_{n \rightarrow \infty} n \left(\ln \left(\phi_x\left(\frac{t}{\sqrt{n}}\right) \right) - \frac{itun}{\sqrt{n}} \right)}$$

$$= e^{\lim_{n \rightarrow \infty} \frac{\ln \left(\phi_x\left(\frac{t}{\sqrt{n}}\right) \right) - \frac{itun}{\sqrt{n}}}{\frac{1}{n}} \cdot \frac{\frac{t^2}{\sqrt{n}}}{\frac{t^2}{\sqrt{n}}}}$$

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

$$= e^{\frac{1}{\sigma^2} \lim_{u \rightarrow 0} \frac{\ln(\phi_x(u)) - iu\mu}{u^2}}$$

$$= \frac{\frac{t^2}{2\sigma^2} \lim_{u \rightarrow 0} \frac{\phi'(u)}{\phi(u)} - iu}{u} = e^{\frac{t^2}{2\sigma^2} \lim_{u \rightarrow 0} \frac{d}{du} \left[\frac{\phi'(u)}{\phi(u)} \right]}$$

$$\text{since } \lim_{u \rightarrow 0} \frac{d}{du} \left[\frac{\phi'(u)}{\phi(u)} \right] = \lim_{u \rightarrow 0} \frac{\phi''(u)\phi(u) - (\phi'(u))^2}{\phi(u)^2}$$

$$= \frac{\phi''(0)\phi(0) - (\phi'(0))^2}{\phi(0)^2} = \frac{i^2 E(x^2)(1) - (iu)^2}{1}$$

$$= i^2 (E(x^2) - u^2) = -\sigma^2$$

Now we have

$$\Rightarrow e^{\frac{t^2}{2\sigma^2} \lim_{u \rightarrow 0} \frac{d}{du} \left[\frac{\phi'(u)}{\phi(u)} \right]} = e^{\frac{t^2}{2\sigma^2} (-\sigma^2)} = e^{-\frac{t^2}{2}} = \phi_{\frac{1}{\sigma}}(t)$$

$$f(x) = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{-tx} \phi(t) dt = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{-tx} e^{-\frac{t^2}{2}} dt$$

$$= \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{-(tx + \frac{t^2}{2})} dt = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{-\left(\frac{t}{\sqrt{2}} + \frac{\sqrt{2}tx}{2}\right)^2 + \frac{x^2}{2}} dt$$

$$= \frac{1}{\sqrt{2\pi}} e^{\frac{x^2}{2}} \int_{\mathbb{R}} e^{-\left(\frac{t}{\sqrt{2}} + \frac{\sqrt{2}tx}{2}\right)^2} dt = \frac{1}{\sqrt{2\pi}} e^{\frac{x^2}{2}} \int_{\mathbb{R}} e^{-y^2} \sqrt{2} dy$$

$$= \frac{1}{\pi\sqrt{2}} \int_{\mathbb{R}} e^{-y^2} dy = \frac{1}{\sqrt{2\pi}} e^{\frac{x^2}{2}}$$

Gaussian Integral $N(0, 1)$

standard normal

$$\text{Let } \frac{\partial^2}{\partial x^2} + \partial \partial x = \left(-\frac{1}{\sqrt{2}} + \frac{\sqrt{2} \partial x}{2} \right)^2 + \frac{x^2}{2}$$

$$\frac{\partial^2}{\partial x^2} \cdot \frac{\sqrt{2} \partial x}{2} = \partial \partial x \left(\frac{\sqrt{2} \partial x}{2} \right)^2 = -\frac{x^2}{2}$$