

①

a) ~~By~~ By Theorem 13.9 in course notes, the distribution of $X-Y$ for independent X and Y depends only on the distributions of X and Y . Since X and X' have same distributions, this fact implies that $X-X'$ and $X'-X$ have same distributions, i.e. $X-X'$ is symmetric.

$$\begin{aligned}
 \text{b) } \varphi_{X^S}(t) &= \mathbb{E} e^{it(X-X')} = \mathbb{E} e^{itX} \cdot \mathbb{E} e^{-itX'} && \text{by independence} \\
 &= \mathbb{E} e^{itX} \cdot \mathbb{E} e^{-itX} && \text{by identical distribution} \\
 &= \varphi_X(t) \cdot \varphi_X(t) \\
 &= \varphi_X(t) \cdot \overline{\varphi_X(t)} = |\varphi_X(t)|^2.
 \end{aligned}$$

$$\begin{aligned}
 \text{c) } \varphi_{(X+Y)^S}(t) &= |\varphi_{X+Y}(t)|^2 && \text{by (b)} \\
 &= |\varphi_X(t) \varphi_Y(t)|^2 && \text{by independence} \\
 &= |\varphi_X(t)|^2 \cdot |\varphi_Y(t)|^2 \\
 &= \varphi_{X^S}(t) \cdot \varphi_{Y^S}(t) && \text{by (b)} \\
 &= \varphi_{X^S+Y^S}(t) && \text{by independence.}
 \end{aligned}$$

Hence, by the Uniqueness Theorem (Corollary 22.5),
 $(X+Y)^S = X^S + Y^S$ in distribution.

(2)

$X - Y$ has the same distribution as X^S . By Problem 1,

$$\varphi_{X-Y}(t) = |\varphi_X(t)|^2$$

~~and~~, in particular, it is non-negative.

The random variable U uniformly distributed on $[t, 1]$ has ch.f.

$$\varphi_U(t) = \frac{\sin t}{t}$$

(see Section 21.1 of Course Notes). This function can have negative values, hence $X - Y$ can not be distributed identically with U .

(3)

Suppose that $|X_k| \leq K$ for all k .

Check Lyapunov's Condition with $p=4$ (for example):

$$\left(\sum_1^n \mathbb{E} |X_k|^4 \right)^{1/4} \leq \left(\sum_1^n K^2 \mathbb{E} X_k^2 \right)^{1/4} = K^{1/2} S_n^{1/2} = o(S_n) \quad \text{as } n \rightarrow \infty,$$

because $S_n \rightarrow \infty$ as $n \rightarrow \infty$. Therefore, CLT holds.

(4)

X_k has mean zero and variance $\sigma_k^2 = 15 \cdot \left(\frac{1}{16}\right)^k$. Hence

S_n has mean zero and variance $\sigma_{S_n}^2 = 15 \sum_{k=1}^n \left(\frac{1}{16}\right)^k = 1 - \left(\frac{1}{16}\right)^n \rightarrow 1$ as $n \rightarrow \infty$.

However, note that $\mathbb{P}(|S_n| \leq \frac{1}{2}) = 0$. Since $S_n \rightarrow 0$, it follows that

$$(*) \quad \mathbb{P}\left(\left|\frac{S_n}{S_n}\right| \leq \frac{1}{4}\right) = 0 \quad \text{for all sufficiently large } n.$$

Moreover, for a standard normal random variable,

$$(**) \quad \mathbb{P}(|N| \leq \frac{1}{4}) \text{ is some constant } > 0.$$

Then $(*)$ and $(**)$ show that $S_n/S_n \not\rightarrow N$ in distribution. QED

(5)

Statement of CLT: let $S_n = X_1 + \dots + X_n$, $S_n^2 = \text{var}(S_n)$.

Then

$$\frac{S_n}{S_n} \rightarrow N(0,1) \text{ in distribution, as } n \rightarrow \infty.$$

We check Lyapunov's condition. let $p > 2$, ^{even.} ~~It suffices~~

~~In view of~~ It suffices to prove that

$$\beta_n := \frac{\left(\sum_{k=1}^n \mathbb{E} X_k^p\right)^{1/p}}{\left(\sum_{k=1}^n \mathbb{E} X_k^2\right)^{1/2}} \rightarrow 0 \text{ as } n \rightarrow \infty.$$

~~Let~~ Note that $\mathbb{E} X_k^p = k^{\alpha p}$. Then

$$\begin{aligned} \left(\sum_{k=1}^n \mathbb{E} X_k^p\right)^{1/p} &= \left(\sum_{k=1}^n k^{\alpha p}\right)^{1/p} \leq C_{\alpha,p} \left(\int_1^n x^{\alpha p} dx\right)^{1/p} \leq C'_{\alpha,p} \left(n^{\alpha p + 1}\right)^{1/p} \\ &\leq C''_{\alpha,p} n^{\alpha + 1/p}. \end{aligned}$$

$$\text{Similarly, } \left(\sum_{k=1}^n \mathbb{E} X_k^2\right)^{1/2} = \left(\sum_{k=1}^n k^{2\alpha}\right)^{1/2} \leq \begin{cases} C_{\alpha,p} n^{\alpha + 1/2} & \text{if } \alpha > -1/2 \\ \frac{C_{\alpha,p}}{\sqrt{\log n}} & \text{if } \alpha = -1/2. \end{cases}$$

$$\text{Hence } \beta_n \leq \left. \begin{cases} C_{p,\alpha}''' \cdot n^{\frac{1}{p} - \frac{1}{2}} & \text{if } \alpha > -\frac{1}{2} \\ C_{p,\alpha}''' \cdot \frac{n^{\frac{1}{p} - \frac{1}{2}}}{\sqrt{\log n}} & \text{if } \alpha = -\frac{1}{2}. \end{cases} \right\} \rightarrow 0 \text{ as } n \rightarrow \infty.$$

Q.E.D.