GATE-EC2023

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Q65ST.2023: Let $\{X_n\}_{n\geq 1}$ be a sequence of independent and identically distributed random variables with mean 0 and variance 1, all of them defined on the same probability space. For n=1,2,3,..., let

$$Y_n = \frac{1}{n}(X_1X_2 + X_3X_4 + ... + X_{2n-1}X_{2n}) \tag{1}$$

Then which of the following statements is/are true?

(A) $\{\sqrt{n}Y_n\}_{n\geq 1}$ converges in distribution to a standard normal random variable.
(B) $\{Y_n\}_{n\geq 1}$ converges in 2nd mean to 0.
(C) $\{Y_n + \frac{1}{n}\}_{n\geq 1}$ converges in probability to 0.
(D) $\{X_n\}_{n\geq 1}$ converges almost surely to 0.

Solution: As X_i is a sequence of independent and identically distributed random variables,
Since X_1, X_2, X_3, X_4 are independent R.V ,

$$P_{X_1X_2X_3X_4}(x) = p_{X_1}(x)p_{X_2}(x)p_{X_3}(x)p_{X_4}(x) \tag{2}$$
$$= p_{X_1X_2}(x)p_{X_3X_4}(x) \tag{3}$$
$$= p_{(X_1X_2)(X_3X_4)}(x) \tag{4}$$

X_iX_{i+1} is a independent R.V
Also, X_1, X_2, X_3, X_4 are identical distributed R.V
Then, X_1X_2 and X_3X_4 are also identical distributed R.V
Thus,X_iX_{i+1} is a identical distributed R.V.
Hence $\sum_{i=1}^{2n-1} X_iX_{i+1}$ is iid.
Now, let $Z_i = X_iX_{i+1}$ then,

$$E(Z_i) = E(X_i)E(X_{i+1}) = 0 \tag{5}$$
$$Var(Z_i) = E(Z_i^2) - [E(Z_i)]^2 = 1 \tag{6}$$
$$\implies Z \sim \mathcal{N}(0, 1) \tag{7}$$
$$Y_n = \frac{1}{n}\left(\sum_{i=1}^{2n-1} Z_i\right) \tag{8}$$
$$\implies E(Y_n) = 0 \tag{9}$$
$$\implies Var(Y_n) = \frac{1}{n} \tag{10}$$
$$Y \sim \mathcal{N}\left(0, \frac{1}{\sqrt{n}}\right) \tag{11}$$

(A) Using (11),

$$\sqrt{n}Y \sim \mathcal{N}(0, 1) \tag{12}$$

For some $\{P_n\}_{n\geq 1}$ converges in distribution to P, $P_n \xrightarrow{d} P$, then for all p,

$$lim_{n\to\infty} F_{P_n}(x) = F_P(x) \tag{13}$$

where, $P \sim \mathcal{N}(0, 1)$

$$lim_{n\to\infty} F_{\sqrt{n}Y_n} = lim_{n\to\infty} \int_{-\infty}^{x} \frac{1}{\sqrt{2\pi\,(1)}} e^{-\frac{(x-0)^2}{2(1)^2}}\, dx \tag{14}$$
$$= lim_{n\to\infty} \int_{-\infty}^{x} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x)^2}{2}}\, dx \tag{15}$$
$$= F_P(x) \tag{16}$$

where, $P \sim \mathcal{N}(0, 1)$
So, option (A) is correct

(B) For 2nd mean to be converging to 0,

$$lim_{n\to\infty} E(|Y_n - 0|^2) = lim_{n\to\infty} E(Y_n^2) \tag{17}$$
$$\because \frac{1}{n} = E(Y_n^2) - [E(Y_n)]^2 \tag{18}$$
$$= lim_{n\to\infty} \frac{1}{n} + [E(Y_n)]^2 \tag{19}$$
$$\because [E(Y_n)]^2 = 0 \tag{20}$$
$$= lim_{n\to\infty} \frac{1}{n} = 0 \tag{21}$$

Thus,$\{Y_n\}_{n\geq 1}$ converges in 2nd mean to 0
Hence, option (B) is correct.

(C) For $\{Y_n + \frac{1}{n}\}_{n\geq 1}$ to be converging $Y_n + \frac{1}{n} \xrightarrow{p} 0$
Using (11) ,

$$E(Y_n + \frac{1}{n}) = \frac{1}{n} \tag{22}$$
$$Var(Y_n + \frac{1}{n}) = \frac{1}{n} \tag{23}$$
$$\tag{24}$$

$$lim_{n\to\infty}F_{Y_n+\frac{1}{n}} = lim_{n\to\infty}\int_{-\infty}^{x}\frac{1}{\sqrt{2\pi\left(\frac{1}{n}\right)}}e^{-\frac{\left(x-\frac{1}{n}\right)^2}{2\left(\frac{1}{n}\right)}}\,dx \tag{25}$$

$$= lim_{n\to\infty}\int_{-\infty}^{x}\frac{n}{\sqrt{2\pi}}e^{-\frac{\left(n^2x-n\right)^2}{2}}\,dx \tag{26}$$

$$= 0 \tag{27}$$

Thus, $Y_n + \frac{1}{n} \xrightarrow{d} 0 \implies Y_n + \frac{1}{n} \xrightarrow{p} 0$
Note: This condition is only true for converges at real constants.
Hence, option (C) is correct.

(D) let $\{X_n\}_{n\geq 1}$ almost surely converges to 0. Then, $X_n \xrightarrow{p} 1$.

$$lim_{n\to\infty}F_{X_n} = lim_{n\to\infty}\int_{-\infty}^{x}\frac{1}{\sqrt{2\pi\,(1)}}e^{\frac{-(x-0)^2}{2(1)^2}}\,dx \tag{28}$$

$$= F_X(x) \tag{29}$$

$$X_n \xrightarrow{d} X \tag{30}$$

$$\implies X_n \xrightarrow{p} X \tag{31}$$

Hence,option (D) is incorrect