Assignment 2

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Download all latex-tikz codes from

https://github.com/ImNamitaKumari/Probabilityand-Random-Variables/blob/main/ Assignment4/Assignment4.tex

1 CSIR UGC NET - June 2015 Q. 105

Suppose $X_1, X_2, ...$ are independent random variables. Assume that X_1, X_3, \dots are independently distributed with mean μ_1 and variance σ_1^2 , while X_2, X_4, \dots are independently distributed with mean μ_2 and variance σ_2^2 . Let $S_n = X_1 + X_2 + ... + X_n$. Then $\frac{S_n - a_n}{h_n}$ converges in distribution to N(0, 1) if

1)
$$a_n = \frac{n(\mu_1 + \mu_2)}{2}$$
 and $b_n = \sqrt{n} \sqrt{\frac{\sigma_1^2 + \sigma_2^2}{2}}$
2) $a_n = \frac{n(\mu_1 + \mu_2)}{2}$ and $b_n = \frac{n(\sigma_1 + \sigma_2)}{2}$
3) $a_n = n(\mu_1 + \mu_2)$ and $b_n = \sqrt{n} \frac{(\sigma_1 + \sigma_2)}{2}$

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$$a_n = \frac{n(\mu_1 + \mu_2)}{2}$$
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$$a_n = n(\mu_1 + \mu_2)$$
 and $b_n = \sqrt{n} \sqrt{\frac{\sigma_1^2 + \sigma_2^2}{2}}$

2 Solution

For X and Y being independent random variables,

$$E(X + Y) = E(X) + E(Y)$$
 (2.0.1)

$$Var(X + Y) = Var(X) + Var(Y)$$
 (2.0.2)

$$Var(aX + b) = a^{2}Var(X)$$
 (2.0.3)

$$S_n = X_1 + X_2 + \dots + X_n$$
 (2.0.4)

Using equation (2.0.1),

$$\implies E(S_n) = E(X_1) + E(X_2) + ... + E(X_n)$$
 (2.0.5)

$$\implies E(S_n) = \begin{cases} \frac{n(\mu_1 + \mu_2)}{2} & n = \text{even} \\ \frac{n(\mu_1 + \mu_2)}{2} + \frac{\mu_1 - \mu_2}{2} & n = \text{odd} \end{cases}$$
 (2.0.6)

Given,

$$\lim_{n \to \infty} E\left(\frac{S_n - a_n}{b_n}\right) = 0 \tag{2.0.7}$$

For all options, a_n and b_n are fixed numbers dependent only on n and not a random variable. So, $E(a_n) = a_n$ and $E(b_n) = b_n$.

$$\implies \lim_{n \to \infty} \frac{E(S_n) - a_n}{b_n} = 0 \tag{2.0.8}$$

(2.0.9)

For all given options,

$$\lim_{n \to \infty} b_n = \infty \tag{2.0.10}$$

So,

$$\lim_{n \to \infty} E(S_n) - a_n = k, k \in \mathbb{R}, \text{ free of } n \qquad (2.0.11)$$

$$\implies \frac{n(\mu_1 + \mu_2)}{2} - a_n = k_1 \qquad (2.0.12)$$

And,
$$\frac{n(\mu_1 + \mu_2)}{2} + \frac{\mu_1 - \mu_2}{2} - a_n = k_2$$
 (2.0.13)

$$\implies a_n = \frac{n(\mu_1 + \mu_2)}{2} + k_3 \qquad (2.0.14)$$

Using equation (2.0.2)

$$Var(S_n) = Var(X_1) + Var(X_2) + \dots + Var(X_n)$$

(2.0.15)

$$\implies Var(S_n) = \begin{cases} \frac{n(\sigma_1^2 + \sigma_2^2)}{2} & n = \text{even} \\ \frac{n(\sigma_1^2 + \sigma_2^2)}{2} + \frac{\sigma_1^2 - \sigma_2^2}{2} & n = \text{odd} \end{cases}$$
(2.0.16)

Given,

$$\lim_{n \to \infty} Var \left(\frac{S_n - a_n}{b_n} \right) = 1 \tag{2.0.17}$$

From equation (2.0.3),

$$\implies \lim_{n \to \infty} \frac{Var(S_n)}{b_n^2} = 1 \quad (2.0.18)$$

$$\implies \frac{n(\sigma_1^2 + \sigma_2^2)}{2b_n^2} = 1, n = \text{even } (2.0.19)$$

And,
$$\frac{n(\sigma_1^2 + \sigma_2^2)}{2b_n^2} + \frac{\sigma_1^2 - \sigma_2^2}{2b_n^2} = 1, n = \text{odd}$$
 (2.0.20)

$$\implies b_n = \sqrt{n} \sqrt{\frac{\sigma_1^2 + \sigma_2^2}{2}}, n = \text{even}$$

$$(2.0.21)$$
And, $b_n = \sqrt{\frac{n(\sigma_1^2 + \sigma_2^2) + (\sigma_1^2 - \sigma_2^2)}{2}}, n = \text{odd}$

$$(2.0.22)$$

Hence from equation (2.0.21) and (2.0.14), the correct answer is option 1).