Assignment 6

CS20BTECH11028

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1 Problem UGC/MATH Dec 2018 104

Let $X_1, X_2, \cdots$ be i.i.d. $N(0,1)$ random variables.Let $S_n = X_1^2+X_2^2+\cdots+X_n^2.\forall n \geq 1$.Which of the following statements are correct?

(A) $\frac{S_n-n}{\sqrt{2}} \sim N(0,1)$ for all $n \geq 1$

(B) For all $\epsilon > 0$,$\Pr\left(\left|\frac{S_n}{n} - 2\right| > \epsilon\right) \to 0$ as $n \to \infty$

(C) $\frac{S_n}{n} \to 1$ with probability 1

(D) $\Pr(S_n \leq n + \sqrt{n}x) \to \Pr(Y \leq x) \forall x \in R$,where $Y \sim N(0,2)$

2 Solution

Definition 2.1 (Almost sure convergence)**.** *A sequence of random variables $\{X_n\}_{n\in N}$ is said to converge almost surely or with probability 1 (denoted by a.s or w.p 1) to X if*

$$\Pr(\omega|X_n(\omega) \to X(\omega)) = 1 \tag{2.0.1}$$

Definition 2.2 (Convergence in probability)**.** *A sequence of random variables $\{X_n\}_{n\in N}$ is said to converge in probability (denoted by i.p) to X if*

$$\lim_{n\to\infty} \Pr(|X_n - X| > \epsilon) = 0, \forall \epsilon > 0 \tag{2.0.2}$$

Theorem 2.1 (Weak law of large numbers)**.** *Let $X_1, X_2, \cdots$ be i.i.d random variables with same expectation(μ) and finite variance(σ^2).Let $S_n = X_1 + X_2 + \cdots X_n$,Then as $n \to \infty$*

$$\frac{S_n}{n} \xrightarrow{i.p} \mu, \tag{2.0.3}$$

in probability.

Theorem 2.2 (Strong law of large numbers)**.** *Let $X_1, X_2, \cdots$ be i.i.d random variables with same expectation(μ) and finite variance(σ^2).Let $S_n = X_1 + X_2 + \cdots X_n$,Then as $n \to \infty$*

$$\frac{S_n}{n} \xrightarrow{a.s} \mu, \tag{2.0.4}$$

almost surely.

Theorem 2.3 (Central limit theorem)**.** *The Central limit theorem states that the distribution of the sample approximates a normal distribution as the sample size becomes larger,given that all the samples are equal in size,regardless of the distribution of the individual samples.*

Given $X_1, X_2, \cdots$ follow normal distribution with mean 0 and variance 1.

$$f_{X_i}(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}, i \in \{1, 2, \cdots\} \tag{2.0.5}$$

As $X_1, X_2, \cdots$ are i.i.d random variables therefore $X_1^2, X_2^2, \cdots$ are also identical and independent. We can write

$$E(X^2) = Var(X) \tag{2.0.6}$$

Given

$$S_n = X_1^2 + X_2^2 + \cdots + X_n^2.\forall n \geq 1 \tag{2.0.7}$$

Hence from theorem 2.1,2.2 we can write

$$\frac{S_n}{n} \to Var(X) \tag{2.0.8}$$

$$\implies \frac{S_n}{n} \to 1 \tag{2.0.9}$$

in probability and almost surely. Hence **Option B is false ,Option C is true**

$$E\left(\frac{S_n - n}{\sqrt{2}}\right) = E\left(\frac{\sum_i (X_i^2 - 1)}{\sqrt{2}}\right) \tag{2.0.10}$$

$$= \frac{\sum_i E(X_i^2 - 1)}{\sqrt{2}} \tag{2.0.11}$$

From (2.0.6) we can write

$$E\left(\frac{S_n - n}{\sqrt{2}}\right) = 0 \tag{2.0.12}$$

$$Var\left(\frac{S_n - n}{\sqrt{2}}\right) = Var\left(\frac{\sum_i (X_i^2 - 1)}{\sqrt{2}}\right) \tag{2.0.13}$$

$$= \frac{\sum_i Var(X_i^2 - 1)}{\sqrt{2}} \tag{2.0.14}$$

$$Var(X_i^2 - 1) = \int_{-\infty}^{\infty} (X_i^2 - 1)^2 f_{X_i}(x)dx \tag{2.0.15}$$

$$= \int_{-\infty}^{\infty} (X_i^4 + 1 - 2X_i^2) f_{X_i}(x)dx \tag{2.0.16}$$

$$= 2 \tag{2.0.17}$$

$$Var\left(\frac{S_n - n}{\sqrt{2}}\right) = n\sqrt{2} \tag{2.0.18}$$

Hence from theorem 2.2 as $n \to \infty$

$$\left(\frac{S_n - n}{\sqrt{2}}\right) \sim N(0, n\sqrt{2}) \tag{2.0.19}$$

Hence **Option A is false.**

$$E\left(\frac{S_n - n}{\sqrt{n}}\right) = 0 \tag{2.0.20}$$

using (2.0.6) and (2.0.11).

$$Var\left(\frac{S_n - n}{\sqrt{n}}\right) = \frac{2n}{\sqrt{n}} \tag{2.0.21}$$

$$= 2\sqrt{n}. \tag{2.0.22}$$

using (2.0.17).From theorem 2.3 we can write,

$$\left(\frac{S_n - n}{\sqrt{n}}\right) \sim N(0, 2\sqrt{n}) \tag{2.0.23}$$

$$\Pr\left(\frac{S_n - n}{\sqrt{n}} \le x\right) = \Pr\left(S_n \le n + \sqrt{n}x\right) \tag{2.0.24}$$

Hence using (2.0.23), **Option D is false.**