

Assignment 6

CS20BTECH11028

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1 PROBLEM UGC/MATH DEC 2018 104

Let X_1, X_2, \dots be i.i.d. $N(0, 1)$ random variables. Let $S_n = X_1^2 + X_2^2 + \dots + X_n^2, \forall n \geq 1$. 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 $\epsilon > 0, \Pr\left(\left|\frac{S_n}{n} - 1\right| > \epsilon\right) \rightarrow 0$ as $n \rightarrow \infty$
- (C) $\frac{S_n}{n} \rightarrow 1$ with probability 1
- (D) $\Pr(S_n \leq n + \sqrt{n}x) \rightarrow \Pr(Y \leq x) \forall x \in \mathbb{R}$, where $Y \sim N(0, 2)$

2 SOLUTION

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

$$\frac{S_n}{n} \rightarrow \mu, \quad (2.0.1)$$

in probability.

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

$$\frac{S_n}{n} \rightarrow \mu, \quad (2.0.2)$$

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, \dots follow normal distribution with mean 0 and variance 1.

$$f_{X_i}(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}, i \in \{1, 2, \dots\} \quad (2.0.3)$$

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

$$E(X^2) = \text{Var}(X) \quad (2.0.4)$$

Given

$$S_n = X_1^2 + X_2^2 + \dots + X_n^2, \forall n \geq 1 \quad (2.0.5)$$

Hence from theorem 2.1, 2.2 we can write

$$\frac{S_n}{n} \rightarrow \text{Var}(X) \quad (2.0.6)$$

$$\implies \frac{S_n}{n} \rightarrow 1 \quad (2.0.7)$$

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

$$E\left(\frac{S_n - n}{\sqrt{2}}\right) = E\left(\frac{\sum_i (X_i^2 - 1)}{\sqrt{2}}\right) \quad (2.0.8)$$

$$= \frac{\sum_i E(X_i^2 - 1)}{\sqrt{2}} \quad (2.0.9)$$

From (2.0.4) we can write

$$E\left(\frac{S_n - n}{\sqrt{2}}\right) = 0 \quad (2.0.10)$$

$$\text{Var}\left(\frac{S_n - n}{\sqrt{2}}\right) = \text{Var}\left(\frac{\sum_i (X_i^2 - 1)}{\sqrt{2}}\right) \quad (2.0.11)$$

$$= \frac{\sum_i \text{Var}(X_i^2 - 1)}{\sqrt{2}} \quad (2.0.12)$$

$$\text{Var}(X_i^2 - 1) = \int_{-\infty}^{\infty} (X_i^2 - 1)^2 f_{X_i}(x) dx \quad (2.0.13)$$

$$= \int_{-\infty}^{\infty} (X_i^4 + 1 - 2X_i^2) f_{X_i}(x) dx \quad (2.0.14)$$

$$= 2 \quad (2.0.15)$$

$$\text{Var}\left(\frac{S_n - n}{\sqrt{2}}\right) = n \sqrt{2} \quad (2.0.16)$$

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

$$\left(\frac{S_n - n}{\sqrt{2}}\right) \sim N(0, n \sqrt{2}) \quad (2.0.17)$$

Hence **Option A is false.**

$$E\left(\frac{S_n - n}{\sqrt{n}}\right) = 0 \quad (2.0.18)$$

using (2.0.4) and (2.0.9).

$$\text{Var}\left(\frac{S_n - n}{\sqrt{n}}\right) = \frac{2n}{\sqrt{n}} \quad (2.0.19)$$

$$= 2 \sqrt{n}. \quad (2.0.20)$$

using (2.0.15). From 2.3 we can write,

$$\left(\frac{S_n - n}{\sqrt{n}}\right) \sim N(0, 2 \sqrt{n}) \quad (2.0.21)$$

$$\Pr\left(\frac{S_n - n}{\sqrt{n}}\right) \leq x = \Pr(S_n \leq n + \sqrt{n}x) \quad (2.0.22)$$

Hence using (2.0.21), **Option D is false.**