Assignment-5

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https://github.com/PrasannaLanka/Assignment5/ blob/main/Assignment5/codes/Assignment5.tex

PROBLEM UGC/MATH (2018 Dec-Math Set-A) Q.104

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

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

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

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

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

SOLUTION

Theorem 0.1 (chi-square distribution). If X_1, X_2, \cdots are independent normally distributed random variables with mean 0 and variance 1. Then χ = $X_1^2 + X_2^2 + \cdots + X_n^2$ is chi-square distributed with n degrees of freedom.

$$E(\chi) = n \tag{0.0.1}$$

$$Var(\chi) = 2n \tag{0.0.2}$$

Theorem 0.2 (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 $n \to \infty$

$$\frac{S_n}{n} \xrightarrow{i.p} \mu, \tag{0.0.3}$$

in probability

Proof. Define a new variable

$$X \equiv \frac{X_1 + X_2 + \dots + X_n}{n} \tag{0.0.4}$$

Then, as $n \to \infty$

$$E(X) = E\left(\frac{X_1 + X_2 + \dots + X_n}{n}\right)$$
 (0.0.5)

$$=\frac{E(X_1)+\cdots E(X_n)}{n} \tag{0.0.6}$$

$$= \frac{E(X_1) + \dots + E(X_n)}{n}$$

$$= \frac{n\mu}{n}$$
(0.0.6)

$$= \mu \tag{0.0.8}$$

In addition,

$$Var(X) = Var\left(\frac{X_1 + X_2 + \dots + X_n}{n}\right) \tag{0.0.9}$$

$$= Var\left(\frac{X_1}{n}\right) + \cdots Var\left(\frac{X_n}{n}\right) \qquad (0.0.10)$$

$$=n\frac{\sigma^2}{n^2} \tag{0.0.11}$$

$$=\frac{\sigma^2}{n}\tag{0.0.12}$$

Therefore, by Chebyshev inequality, for all $\epsilon > 0$,

$$\Pr(|X - \mu| \ge \epsilon) \le \frac{Var(X)}{\epsilon^2} = \frac{\sigma^2}{n\epsilon^2}$$
 (0.0.13)

As $n \to \infty$, it follows that

$$\lim_{n \to \infty} \Pr(|X - \mu| \ge \epsilon) = 0 \tag{0.0.14}$$

Stated other way as $n \to \infty$

$$\frac{S_n}{n} \xrightarrow{i.p} \mu, \tag{0.0.15}$$

in probability

Theorem 0.3 (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 $n \to \infty$

$$\frac{S_n}{n} \xrightarrow{a.s} \mu, \tag{0.0.16}$$

almost surely.

Theorem 0.4 (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.

Definition 0.1 (Almost sure convergence). A sequence of random variables $\{X_n\}_{n\in\mathbb{N}}$ is said to converge almost surely or with probability 1 (denoted by a.s or w.p 1) to X if

$$\Pr(\omega|X_n(\omega) \to X(\omega)) = 1$$
 (0.0.17)

Lemma 0.5 (Properties of mean and variance). *If* X is a random variable with a probability density function of f(x). If a and b are constants.

$$E(X) = \int_{R} x f(x) dx \qquad (0.0.18)$$

$$E(X + Y) = E(X) + E(Y)$$
 (0.0.19)

$$E(aX + b) = aE(X) + b, (0.0.20)$$

$$Var(X) = E(X^2) - E(X)^2$$
 (0.0.21)

$$Var(X + Y) = Var(X) + Var(Y)$$
 (0.0.22)

$$Var(aX + b) = a^{2}Var(X)$$
 (0.0.23)

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

$$f_{X_i}(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}, i \in \{1, 2, \dots\}$$
 (0.0.24)

(A) From theorem (0.1) S_n is a chi-distributed function with n degrees of freedom. From (0.0.1)

$$E(S_n) = n \tag{0.0.25}$$

From (0.0.25) and (0.0.20)

$$E\left(\frac{S_n - n}{\sqrt{2}}\right) = \frac{E(S_n) - n}{\sqrt{2}} \tag{0.0.26}$$

$$=\frac{n-n}{\sqrt{2}}\tag{0.0.27}$$

$$= 0$$
 (0.0.28)

From (0.0.2)

$$Var(S_n) = 2n \qquad (0.0.29)$$

From (0.0.29) and (0.0.23)

$$Var\left(\frac{S_n - n}{\sqrt{2}}\right) = Var\left(\frac{S_n}{\sqrt{2}}\right) \tag{0.0.30}$$

$$=\frac{Var(S_n)}{2} \tag{0.0.31}$$

$$=\frac{2n}{2}$$
 (0.0.32)

$$= n$$
 (0.0.33)

Hence,

$$\left(\frac{S_n - n}{\sqrt{2}}\right) \sim N(0, n) \tag{0.0.34}$$

Hence Option A is false.

(B) Given

$$S_n = X_1^2 + X_2^2 + \dots + X_n^2 . \forall n \ge 1$$
 (0.0.35)

Assume that For all $\epsilon > 0$, $\Pr\left(\left(\left|\frac{S_n}{n} - 2\right| > \epsilon\right)\right) \rightarrow$ 0 as $n \to \infty$ is true let

$$X \equiv \frac{X_1^2 + X_2^2 + \dots + X_n^2}{n} \tag{0.0.36}$$

From (0.0.25) and (0.0.20)

$$E(X) = E\left(\frac{S_n}{n}\right) \tag{0.0.37}$$

$$= \frac{E(S_n)}{n}$$
 (0.0.38)
= $\frac{n}{n}$ (0.0.39)

$$=\frac{n}{n}\tag{0.0.39}$$

$$= 1$$
 (0.0.40)

From theorem (0.2)

$$\frac{S_n}{n} \xrightarrow{i.p} E\left(\frac{S_n}{n}\right) \tag{0.0.41}$$

$$\lim_{n \to \infty} \Pr\left(\left| \frac{S_n}{n} - 1 \right| > \epsilon \right) = 0 \tag{0.0.42}$$

This means for all $\epsilon > 0$, $\Pr(\left|\frac{S_n}{n} - 1\right| > \epsilon) \to 0$ as $n \to \infty$

But this is contradiction to our assumption.

Hence Option B is false.

(C) Given

$$S_n = X_1^2 + X_2^2 + \dots + X_n^2, \forall n \ge 1$$
 (0.0.43)

Hence from theorem 0.3 we can write

$$\frac{S_n}{n} \xrightarrow{a.s} E\left(\frac{S_n}{n}\right) \tag{0.0.44}$$

From (0.0.40)

$$\frac{S_n}{n} \xrightarrow{a.s} 1 \tag{0.0.45}$$

almost surely.

$$\Pr\left(\lim_{n\to\infty}\frac{S_n}{n}=1\right)=1\tag{0.0.46}$$

From definition 0.1 and (0.0.46) we can write,

$$\frac{S_n}{n} \xrightarrow{w.p.1} 1 \tag{0.0.47}$$

with probability 1.

Hence Option C is true.

(D) Let

$$Y = \frac{S_n - n}{\sqrt{n}} \tag{0.0.48}$$

Using (0.0.25) and (0.0.20)

$$E\left(\frac{S_n - n}{\sqrt{n}}\right) = \frac{E(S_n) - n}{\sqrt{n}} \tag{0.0.49}$$

$$=\frac{n-n}{\sqrt{n}}\tag{0.0.50}$$

$$= 0 (0.0.51)$$

using (0.0.25) and (0.0.20)

$$Var\left(\frac{S_n - n}{\sqrt{n}}\right) = \frac{Var(S_n)}{n} \tag{0.0.52}$$

$$=\frac{2n}{n}\tag{0.0.53}$$

$$= 2 (0.0.54)$$

Hence, from theorem (0.4)

$$Y \sim N(0, 2)$$
 (0.0.55)

$$\Pr\left(\frac{S_n - n}{\sqrt{n}} \le x\right) = \Pr\left(S_n \le n + \sqrt{n}x\right)$$
(0.0.56)

Therefore,

$$\Pr(S_n \le n + \sqrt{n}x) \to \Pr(Y \le x) \, \forall x \in R$$
(0.0.57)

Hence, Option D is true.