

# UGC/MATH (2018 Dec-Math set-a ) Q.104

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# Topics covered

## Prerequisites:

- Chi-square distribution
- Weak law of large numbers with its proof
- Strong law of large numbers
- Almost sure convergence
- Central limit theorem
- Properties of mean and variance

## UGC NET problem

- Question
- Solution

## Chi-square distribution

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

$$E(\chi) = n \quad (1)$$

$$Var(\chi) = 2n \quad (2)$$

## Weak law of large numbers

Let  $X_1, X_2, \dots$  be i.i.d random variables with same expectation( $\mu$ ) and finite variance( $\sigma^2$ ). Let  $S_n = X_1 + X_2 + \dots + X_n$ , Then as  $n \rightarrow \infty$

$$\frac{S_n}{n} \xrightarrow{i.p} \mu, \quad (3)$$

in probability

### Proof:

Define a new variable

$$X \equiv \frac{X_1 + X_2 + \dots + X_n}{n} \quad (4)$$

Then, as  $n \rightarrow \infty$

$$E(X) = E\left(\frac{X_1 + X_2 + \cdots X_n}{n}\right) \quad (5)$$

$$= \frac{E(X_1) + \cdots E(X_n)}{n} \quad (6)$$

$$= \frac{n\mu}{n} \quad (7)$$

$$= \mu \quad (8)$$

In addition,

$$Var(X) = Var\left(\frac{X_1 + X_2 + \cdots X_n}{n}\right) \quad (9)$$

$$= Var\left(\frac{X_1}{n}\right) + \cdots Var\left(\frac{X_n}{n}\right) \quad (10)$$

$$= \frac{n\sigma^2}{n^2} \quad (11)$$

$$= \frac{\sigma^2}{n} \quad (12)$$

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

$$\Pr(|X - \mu| \geq \epsilon) \leq \frac{\text{Var}(X)}{\epsilon^2} = \frac{\sigma^2}{n\epsilon^2} \quad (13)$$

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

$$\lim_{n \rightarrow \infty} \Pr(|X - \mu| \geq \epsilon) = 0 \quad (14)$$

Stated other way as  $n \rightarrow \infty$

$$\frac{S_n}{n} \xrightarrow{i.p} \mu, \quad (15)$$

in probability

## Strong law of large numbers

Let  $X_1, X_2, \dots$  be i.i.d random variables with same expectation( $\mu$ ) and finite variance( $\sigma^2$ ). Let  $S_n = X_1 + X_2 + \dots + X_n$ , Then as  $n \rightarrow \infty$

$$\frac{S_n}{n} \xrightarrow{\text{a.s.}} \mu, \quad (16)$$

almost surely.

## Almost sure convergence

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

$$\Pr(\omega | X_n(\omega) \rightarrow X(\omega)) = 1 \quad (17)$$

## 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.



## 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.

$$E(X) = \int_R xf(x)dx \quad (18)$$

$$E(X + Y) = E(X) + E(Y) \quad (19)$$

$$E(aX + b) = aE(X) + b, \quad (20)$$

$$\text{Var}(X) = E(X^2) - E(X)^2 \quad (21)$$

$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y) \quad (22)$$

$$\text{Var}(aX + b) = a^2 \text{Var}(X) \quad (23)$$

## Question

### UGC/MATH (2018 Dec-Math set-a ) Q.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  $\Pr(S_n \leq n + \sqrt{nx}) \rightarrow \Pr(Y \leq x) \forall x \in R$ , where  $Y \sim N(0, 2)$

## Solution

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

$$f_{X_i}(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}, i \in 1, 2, \dots \quad (24)$$

## Option A

From definition of chi-square distribution  $S_n$  is a chi-distributed function with  $n$  degrees of freedom.

From (1)

$$E(S_n) = n \quad (25)$$

From (25) and (20)

$$E\left(\frac{S_n - n}{\sqrt{2}}\right) = \frac{E(S_n) - n}{\sqrt{2}} \quad (26)$$

$$= \frac{n - n}{\sqrt{2}} \quad (27)$$

$$= 0 \quad (28)$$

From (2)

$$\text{Var}(S_n) = 2n \quad (29)$$

## Option A contd.

From (29) and (23)

$$\text{Var} \left( \frac{S_n - n}{\sqrt{2}} \right) = \text{Var} \left( \frac{S_n}{\sqrt{2}} \right) \quad (30)$$

$$= \frac{\text{Var}(S_n)}{2} \quad (31)$$

$$= \frac{2n}{2} \quad (32)$$

$$= n \quad (33)$$

Hence, from central limit theorem

$$\left( \frac{S_n - n}{\sqrt{2}} \right) \sim N(0, n) \quad (34)$$

Hence **Option A is false.**

## Option B

Given

$$S_n = X_1^2 + X_2^2 + \cdots + X_n^2, \forall n \geq 1 \quad (35)$$

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

$$X \equiv \frac{X_1^2 + X_2^2 + \cdots + X_n^2}{n} \quad (36)$$

From (25) and (20)

$$E(X) = E\left(\frac{S_n}{n}\right) \quad (37)$$

$$= \frac{E(S_n)}{n} \quad (38)$$

$$= \frac{n}{n} \quad (39)$$

$$= 1 \quad (40)$$

## Option B contd.

From weak law of large numbers

$$\frac{S_n}{n} \xrightarrow{i.p.} E\left(\frac{S_n}{n}\right) \quad (41)$$

$$\lim_{n \rightarrow \infty} \Pr\left(\left|\frac{S_n}{n} - 1\right| > \epsilon\right) = 0 \quad (42)$$

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

But this is contradiction to our assumption.

Hence **Option B is false**.

## Option C

Given

$$S_n = X_1^2 + X_2^2 + \cdots + X_n^2, \forall n \geq 1 \quad (43)$$

Hence from Strong law of large numbers we can write

$$\frac{S_n}{n} \xrightarrow{a.s.} E\left(\frac{S_n}{n}\right) \quad (44)$$

From (40)

$$\frac{S_n}{n} \xrightarrow{a.s.} 1 \quad (45)$$

almost surely.

$$\Pr\left(\lim_{n \rightarrow \infty} \frac{S_n}{n} = 1\right) = 1 \quad (46)$$



## Option C contd.

From definition of Almost sure convergence and (46) we can write,

$$\frac{S_n}{n} \xrightarrow{w.p.1} 1 \quad (47)$$

with probability 1.

Hence **Option C is true.**

## Option D

Let

$$Y = \frac{S_n - n}{\sqrt{n}} \quad (48)$$

Using (25) and (20)

$$E\left(\frac{S_n - n}{\sqrt{n}}\right) = \frac{E(S_n) - n}{\sqrt{n}} \quad (49)$$

$$= \frac{n - n}{\sqrt{n}} \quad (50)$$

$$= 0 \quad (51)$$

using (25) and (20)

$$\text{Var}\left(\frac{S_n - n}{\sqrt{n}}\right) = \frac{\text{Var}(S_n)}{n} \quad (52)$$

$$= \frac{2n}{n} \quad (53)$$

$$= 2 \quad (54)$$

## Option D contd.

Hence, from central limit theorem

$$Y \sim N(0, 2) \quad (55)$$

$$\Pr\left(\frac{S_n - n}{\sqrt{n}} \leq x\right) = \Pr(S_n \leq n + \sqrt{nx}) \quad (56)$$

Therefore,

$$\Pr(S_n \leq n + \sqrt{nx}) \rightarrow \Pr(Y \leq x) \forall x \in R \quad (57)$$

Hence, **Option D is true.**