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AI1103 ASSIGNMENT 4

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Download the python code from

https://github.com/sarandeepmannam/

ASSIGNMENT4/blob/main/Assignment4.py

and latex-tikz code from

https://github.com/sarandeepmannam/

ASSIGNMENT4/blob/main/Assignment4.tex

1 QUESTION-CSIR UGC NET JUNE 2012,Q.50

Let $X_1, X_2,$ be i.i.d N(1,1) random variables.Let $S_n = X_1^2 + X_2^2 + ... + X_n^2$ for $n \ge 1$.Then

$$\lim_{n\to\infty}\frac{Var(S_n)}{n}=$$

- (A) 4
- (B) 6
- (C) 1
- (D) 0

2 SOLUTION-CSIR UGC NET JUNE 2012,Q.50

Definition 1 (NON-CENTRAL CHI SQUARE DISTRIBUTION). Let X_1, X_2, X_3, X_i, X_n be n independent, normally distributed random variables with means μ_i and unit variances. Then the random variable

$$\sum_{i=0}^{n} X_{i}$$

is distributed according to the non-central chi square distribution. It has It has two parameters 'k' which specifies the number of degrees of freedom (i.e. the number of X_i), and ' λ ' which is called non-centrality parameter given by,

$$\lambda = \sum_{i=0}^{n} \mu_i^2$$
 (2.0.1)

Lemma 2.1. Moment generating function of a non-central chi square distributed random variable X is given by,

$$M_X(t) = \frac{e^{\frac{\lambda t}{1-2t}}}{(1-2t)^{\frac{n}{2}}}$$
 (2.0.2)

We know that if two normal random variables X, Y are independent then X^2, Y^2 are also independent. Since $X_1, X_2, X_3, ...$ are mutually independent random variables therefore the random variables $X_1^2, X_2^2, X_3^2, ...$ are also mutually independent.

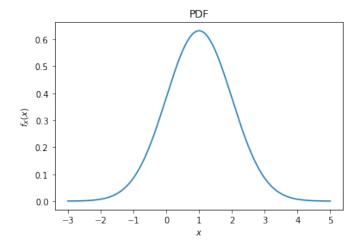


Fig. 4: PDF of $X_1, X_2, ...$

Given,

$$S_n = X_1^2 + X_2^2 + ... + X_n^2$$
 (2.0.3)

From definition 1, Since $X_1, X_2, ...$ are i.i.d N(1,1) normal random variables therefore the distribution of the random variable $S_n = X_1^2 + X_2^2 + ... X_n^2$ is a non-central chi square distribution with 'n' degrees of freedom and non centrality parameter ' λ ' given by

$$\lambda = \sum_{i=1}^{n} (\mu_i)^2 \tag{2.0.4}$$

$$\lambda = \sum_{i=1}^{n} 1 = n \tag{2.0.5}$$

From Lemma 2.1,

$$M_{S_n}(t) = \frac{e^{\frac{nt}{1-2t}}}{(1-2t)^{\frac{n}{2}}}$$
 (2.0.6)

and

$$Var(S_n) = E(S_n^2) - (E(S_n))^2$$
 (2.0.7)

From the properties of MGF,

$$E(S_n) = \frac{d}{dt} (M_{S_n}(t))|_{t=0}$$
 (2.0.8)

$$E(S_n) = \frac{2ne^{\frac{nt}{1-2t}}}{(1-2t)^{\frac{n}{2}+2}}(1-t)|_{t=0}$$
 (2.0.9)

$$\implies E(S_n) = 2n \tag{2.0.10}$$

From the properties of MGF,

$$E(S_n^2) = \frac{d^2}{dt^2} (M_{S_n}(t))|_{t=0}$$
 (2.0.11)

$$E(S_n^2) = \frac{2ne^{\frac{nt}{1-2t}}}{(1-2t)^{\frac{n}{2}+4}}(n(1-t) - (1-2t)^2 + (1-t)(1-2t)(n+4))|_{t=0}$$
 (2.0.12)

$$E(S_n^2) = \frac{2n}{1^{n/2+2}} \left(n - 1^2 + n + 4 \right)$$
 (2.0.13)

$$E(S_n^2) = 2n \times (2n+3) \tag{2.0.14}$$

$$\implies E(S_n^2) = 4n^2 + 6n$$
 (2.0.15)

Substituting equations (2.0.10) and (2.0.15) in (2.0.7), we get

$$Var(S_n) = 4n^2 + 6n - (2n)^2$$
 (2.0.16)

$$\implies Var(S_n) = 6n$$
 (2.0.17)

and

$$\lim_{n\to\infty} \frac{Var(S_n)}{n} = \lim_{n\to\infty} \frac{6n}{n} = \lim_{n\to\infty} 6 = 6$$

Hence, option(B) is correct.