## 1

## 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  $\mu_i$  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 two parameters 'k' which specifies the number of degrees of freedom (i.e. the number of  $X_i$ ), and ' $\lambda$ ' which is called non-centrality parameter given by,

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

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 ' $\lambda$ ' given by

$$\lambda = \sum_{i=1}^{n} (\mu_i)^2$$
 (2.0.2)

$$\lambda = \sum_{i=1}^{n} 1 = n \tag{2.0.3}$$

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

$$M_X(t) = \frac{e^{\frac{\lambda t}{1-2t}}}{(1-2t)^{\frac{n}{2}}}$$
 (2.0.4)

From Lemma 2.1 Moment generating function of  $S_n$  is given by,

$$M_{S_n}(t) = \frac{e^{\frac{nt}{1-2t}}}{(1-2t)^{\frac{n}{2}}}$$
 (2.0.5)

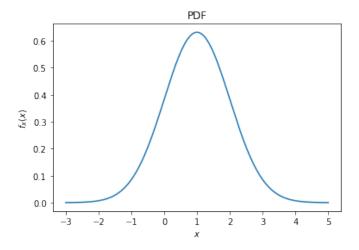


Fig. 4: PDF of  $X_1, X_2, ...$ 

**Lemma 2.2.** The nth moment of a random variable X whose MGF is  $M_X(t) \left( \equiv E(e^{tX}) = \sum_{k=0}^{\infty} \frac{(tX)^k}{k!} \right)$  is given by,

$$E(X^{n}) = \left(\frac{d^{n}}{dt^{n}}(M_{X}(t))\right)_{t=0}$$
 (2.0.6)

Proof.

$$M_X(t) \equiv E(e^{tX}) \tag{2.0.7}$$

Using taylor series expansion,

$$M_X(t) = E(1 + \frac{tX}{1!} + \frac{(tX)^2}{2!} + \frac{(tX)^3}{3!} + \dots + \frac{(tX)^n}{n!} + \dots)$$
(2.0.8)

Taking *n*th derivative on both sides with respect to 't' at t = 0 we get,

$$\left(\frac{d^n}{dt^n} (M_X(t))\right)_{t=0} = (E(X^n))_{t=0} + \left(tE\left(\frac{X^{n+1}}{n+1}\right)\right)_{t=0} + \left(t^2E\left(\frac{X^{n+2}}{(n+1)(n+2)}\right)\right)_{t=0} + \dots$$
 (2.0.9)

$$\left(\frac{d^n}{dt^n}(M_X(t))\right)_{t=0} = E(X^n)$$
 (2.0.10)

From lemma 2.2,

$$E(S_n) = \left(\frac{d}{dt} \left(M_{S_n}(t)\right)\right)_{t=0}$$
 (2.0.11)

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

$$\implies E(S_n) = 2n \tag{2.0.13}$$

From lemma 2.2,

$$E(S_n^2) = \left(\frac{d^2}{dt^2} \left(M_{S_n}(t)\right)\right)_{t=0}$$
 (2.0.14)

$$E(S_n^2) = \frac{2ne^{\frac{nt}{1-2t}}}{(1-2t)^{\frac{n}{2}+4}}(n(1-t) - (1-2t)^2 + (1-t)(1-2t)(n+4))|_{t=0} \quad (2.0.15)$$

$$E(S_n^2) = \frac{2n}{1n/2+2} \left( n - 1^2 + n + 4 \right)$$
 (2.0.16)

$$E(S_n^2) = 2n \times (2n+3) \tag{2.0.17}$$

$$\implies E(S_n^2) = 4n^2 + 6n \tag{2.0.18}$$

Variance of  $S_n$ ,

$$Var(S_n) = E(S_n^2) - (E(S_n))^2$$
 (2.0.19)

Substituting equations (2.0.13) and (2.0.18) in (2.0.6) (2.0.19),we get

$$Var(S_n) = 4n^2 + 6n - (2n)^2$$
 (2.0.20)

$$\implies Var(S_n) = 6n \tag{2.0.21}$$

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.