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Assignment 5

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Download all latex codes from

https://github.com/vaibhavchhabra25/AI1103-course/blob/main/Assignment-5/main.tex

1 Problem

(UGC/MATH 2018 (June set-a)-Q.106) Let $\{X_i\}_{i\geq 1}$ be a sequence of i.i.d. random variables with $E(X_i) = 0$ and $V(X_i) = 1$. Which of the following are true?

1)
$$\frac{1}{n} \sum_{i=1}^{n} X_i^2 \to 0$$
 in probability

2)
$$\frac{1}{n^{3/4}} \sum_{i=1}^{n} X_i \to 0$$
 in probability

3)
$$\frac{1}{n^{1/2}} \sum_{i=1}^{n} X_i \to 0$$
 in probability

4)
$$\frac{1}{n} \sum_{i=1}^{n} X_i^2 \rightarrow 1$$
 in probability

2 Solution

Definition 1. (Convergence in probability)

Let X1, X2, ... be an infinite sequence of random variables, and let Y be another random variable. Then the sequence $\{X_n\}$ converges in probability to Y, if for all $\epsilon > 0$,

$$\lim_{n \to \infty} \Pr\left(|X_n - Y| \ge \epsilon\right) = 0 \tag{2.0.1}$$

And we write as $n \to \infty$, $X_n \to Y$ in probability.

Theorem 2.1. (Strong Law of Large Numbers) Let X_1, X_2, \cdots be a sequence of i.i.d. random variables, each having finite mean $E(X_i)$. Then for any $\epsilon > 0$,

$$\lim_{n \to \infty} \Pr\left(\left| \frac{1}{n} \sum_{i=1}^{n} X_i - E(X_i) \right| \ge \epsilon \right) = 0$$
 (2.0.2)

Or, $\frac{1}{n}\sum_{i=1}^{n} X_i$ converges in probability to $E(X_i)$.

Given,

$$E(X_i) = 0 (2.0.3)$$

$$V(X_i) = 1 (2.0.4)$$

Also, we know that,

$$E(X_i^2) = V(X_i) + (E(X_i))^2$$
 (2.0.5)

Putting given values, we get,

$$E(X_i^2) = 1 + 0^2 = 1$$
 (2.0.6)

$$\implies E(X_i^2) = 1 \tag{2.0.7}$$

So, $E(X_i^2)$ is finite.

As $\{X_i\}$ is sequence of i.i.d. random variables, it follows the following conditions $\forall x, x_i \in \mathbb{R}$:

1)
$$Pr(X_1 \le x) = Pr(X_k \le x) \ \forall k \in \{1, 2, 3 \cdots n\}$$

2)
$$\Pr(X_1 \le x_1, X_2 \le x_2 \cdots X_n \le x_n) = \Pr(X_1 \le x_1) \cdot \Pr(X_2 \le x_2) \cdots \Pr(X_n \le x_n)$$

We can rewrite (1) as

$$\Pr(X_1^2 \le x^2) = \Pr(X_k^2 \le x^2)$$
 (2.0.8)

Putting $x^2 = y$, we get,

$$\Pr\left(X_1^2 \le y\right) = \Pr\left(X_k^2 \le y\right) \tag{2.0.9}$$

Also, we can rewrite (2) as

$$\Pr\left(X_{1}^{2} \leq x_{1}^{2}, X_{2}^{2} \leq x_{2}^{2} \cdots X_{n}^{2} \leq x_{n}^{2}\right) = \\ \Pr\left(X_{1}^{2} \leq x_{1}^{2}\right) \cdot \Pr\left(X_{2}^{2} \leq x_{2}^{2}\right) \cdot \cdot \cdot \Pr\left(X_{n}^{2} \leq x_{n}^{2}\right)$$
(2.0.10)

Putting $x_i^2 = y_i$, we get,

$$\Pr\left(X_1^2 \le y_1, X_2^2 \le y_2 \cdots X_n^2 \le y_n\right) =$$

$$\Pr\left(X_1^2 \le y_1\right) \cdot \Pr\left(X_2^2 \le y_2\right) \cdot \cdot \cdot \Pr\left(X_n^2 \le y_n\right)$$
(2.0.11)

By (2.0.9) and (2.0.11), $\{X_i^2\}$ must also be a sequence of i.i.d. random variables.

So, we can apply S.L.L.N. to this sequence.

Then, $\frac{1}{n}\sum_{i=1}^{n}X_{i}^{2}$ converges in probability to $E(X_{i}^{2})$.

Or, $\frac{1}{n} \sum_{i=1}^{n} X_i^2 \to 1$ in probability.

Thus, option 1 is wrong and option 4 is correct.

Now, we define

$$Y_n = \frac{1}{n^{3/4}} \sum_{i=1}^n X_i$$
 (2.0.12)

Then,

$$E(Y_n) = \frac{1}{n^{3/4}} E\left(\sum_{i=1}^n X_i\right)$$
 (2.0.13)

$$\implies E(Y_n) = \frac{1}{n^{3/4}} \left(E(X_1) + E(X_2) + \dots + E(X_n) \right)$$
(2.0.14)

Using (2.0.3)

$$E(Y_n) = \frac{1}{n^{3/4}}(0) = 0 (2.0.15)$$

Now,

$$V(Y_n) = V\left(\frac{1}{n^{3/4}} \sum_{i=1}^n X_i\right)$$
 (2.0.16)

$$\implies V(Y_n) = \frac{1}{n^{3/2}}V(X_1 + X_2 + \dots + X_n) \quad (2.0.17)$$

As $X_1, X_2, \dots X_n$ are independent of each other,

$$V(Y_n) = \frac{1}{n^{3/2}} \left(V(X_1) + V(X_2) + \dots + V(X_n) \right)$$
(2.0.18)

Using (2.0.4)

$$V(Y_n) = \frac{1}{n^{3/2}} (1 + 1 + \dots + 1) = \frac{1}{n^{3/2}} \times n = \frac{1}{n^{1/2}}$$
(2.0.19)

Now for any $\epsilon > 0$, consider the probability

$$\Pr(|Y_n - 0| \ge \epsilon) = \Pr(|Y_n - E(Y_n)| \ge \epsilon)$$
 (2.0.20)

Applying Chebyschev's inequality here, we get,

$$\Pr(|Y_n - 0| \ge \epsilon) \le \frac{V(Y_n)}{\epsilon^2} = \frac{1}{n^{1/2} \epsilon^2}$$
 (2.0.21)

So,

$$\lim_{n \to \infty} \Pr(|Y_n - 0| \ge \epsilon) \le \lim_{n \to \infty} \frac{1}{n^{1/2} \epsilon^2} = 0 \quad (2.0.22)$$

$$\implies \lim_{n \to \infty} \Pr\left(\left|\frac{1}{n^{3/4}} \sum_{i=1}^{n} X_i - 0\right| \ge \epsilon\right) = 0 \quad (2.0.23)$$

So, $\frac{1}{n^{3/4}} \sum_{i=1}^{n} X_i \to 0$ in probability.

Thus, option 2 is also correct. So the answer must be options 2,4.