Convergence

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Question

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Let $\{X_j\}$ be a sequence of independent Bernoulli random variables with $\mathbb{P}(X_j=1)=\frac{1}{4}$ and let $Y_n=\frac{1}{n}\sum_{j=1}^n X_j^2$. Then Y_n converges, in probability, to _______.

Various Convergences

In probability theory, there exist several different notions of convergence of random variables.

Convergence in distribution or Converge weakly

$$\lim_{n \to \infty} F_{X_n}(x) = F_X(x) \quad \forall x \in \mathbb{R}$$
 (1)

2 Convergence in probability

$$\lim_{n \to \infty} \Pr\left(|X_n - X| \ge \epsilon\right) = 0 \tag{2}$$

3 Almost sure convergence

$$\Pr\left(\lim_{n\to\infty} X_n = X\right) = 1\tag{3}$$

3/13

More convergences

Sure convergence or pointwise convergence

$$\lim_{n \to \infty} X_n(\omega) = X(\omega) \quad \forall \omega \in \Omega$$
 (4)

where Ω is the sample space of the underlying probability space over which the random variables are defined.

5 Convergence in mean. Given a real number $r \geq 1$

$$\lim_{n \to \infty} E(|X_n - X|^r) = 0 \tag{5}$$

Properties

There are several properties involving different types of convergence of which few are listed below

- Almost sure convergence implies convergence in probability
- 2 Convergence in probability implies convergence in distribution
- **1** If X_n converges in distribution to a constant c, then X_n converges in probability to c
- Convergence in r-th order mean implies convergence in lower order mean, assuming that both orders are greater than or equal to one
- **o** Convergence in the r-th mean, for $r \ge 1$, implies convergence in probability

For more properties, refer to the properties section in https://en.wikipedia.org/wiki/Convergence_of_random_variables

Proof of property (5)

Proof

For any
$$\epsilon > 0$$
, (6)

$$\Pr(|Y_n - Y| \ge \epsilon) = \Pr(|Y_n - Y|^2 \ge \epsilon^2)$$
(7)

$$\Pr(|Y_n - Y| \ge \epsilon) \le \frac{E|Y_n - Y|^2}{\epsilon^2}$$
 (by Markov's Inequality) (8)

$$\lim_{n\to\infty} E(|Y_n - Y|^2) = 0 \tag{9}$$

$$0 \le \lim_{n \to \infty} \Pr(|Y_n - Y| \ge \epsilon) \le \frac{0}{\epsilon^2}$$
 (10)

$$\lim_{n \to \infty} \Pr(|Y_n - Y| \ge \epsilon) = 0 \quad \forall \epsilon > 0$$
(11)

Markov's Inequality

If X is a non-negative random variable and a > 0, then the probability that X is at least a is at most the expectation of X divided by a:

$$\Pr\left(X \ge a\right) \le \frac{E(X)}{a} \tag{12}$$

Solution

$$Pr(X_j = 1) = \frac{1}{4}$$

$$Pr(X_j = 0) = 1 - \frac{1}{4} = \frac{3}{4}$$
(13)

$$\Pr\left(X_{j}=0\right)=1-\frac{1}{4}=\frac{3}{4}\tag{14}$$

$$Y_n = \frac{1}{n} \sum_{i=1}^n X_i^2 \tag{15}$$

$$= \frac{1}{n} \sum_{j=1}^{n} X_j \tag{16}$$

$$\Pr\left(Y_n = y\right) = {^n}C_{ny} \left(\frac{1}{4}\right)^{ny} \left(\frac{3}{4}\right)^{n-ny} \tag{17}$$

$$\Pr\left(Y_n = \frac{k}{n}\right) = {^n}C_k \left(\frac{1}{4}\right)^k \left(\frac{3}{4}\right)^{n-k} \tag{18}$$

Solution contd.

$$E\left(|Y_n - \frac{1}{4}|^2\right) = E\left(Y_n^2 - \frac{1}{2}Y_n + \frac{1}{16}\right) \tag{19}$$

$$= E(Y_n^2) - \frac{1}{2}E(Y_n) + \frac{1}{16}$$
 (20)

$$E(Y_n^2) = \sum_{k=0}^n \left(\frac{k}{n}\right)^2 \Pr\left(Y_n = \frac{k}{n}\right)$$
 (21)

$$=\sum_{k=0}^{n} \left(\frac{k^2}{n^2}\right) {}^{n} C_k \left(\frac{1}{4}\right)^k \left(\frac{3}{4}\right)^{n-k} \tag{22}$$

$$=\frac{1}{16}+\frac{3}{16n}\tag{23}$$

9/13

Solution contd.

$$E(Y_n) = \sum_{k=0}^n \frac{k}{n} \Pr\left(Y_n = \frac{k}{n}\right)$$
 (24)

$$=\sum_{k=0}^{n} \left(\frac{k}{n}\right) {}^{n} C_{k} \left(\frac{1}{4}\right)^{k} \left(\frac{3}{4}\right)^{n-k}$$
 (25)

$$=\frac{1}{4}\tag{26}$$

Using equation (23),

$$E\left(|Y_n - \frac{1}{4}|^2\right) = \frac{1}{16} + \frac{3}{16n} - \frac{1}{2} \times \frac{1}{4} + \frac{1}{16}$$
 (27)

$$=\frac{3}{16n}\tag{28}$$

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Solution contd.

$$\lim_{n \to \infty} E\left(|Y_n - \frac{1}{4}|^2\right) = \lim_{n \to \infty} \frac{3}{16n}$$
 (29)

$$=\frac{3}{16}\lim_{n\to\infty}\frac{1}{n}\tag{30}$$

$$=0 (31)$$

Thus, Y_n converges, in mean square, to $\frac{1}{4}$ and hence Y_n converges, in probability, to $\frac{1}{4}$.

Figures

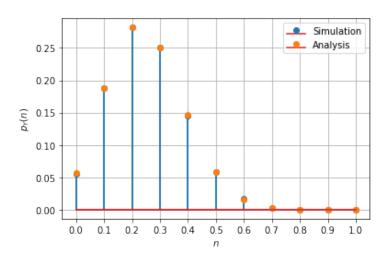


Figure: The PMF distribution of Y_n for n=10

Figures

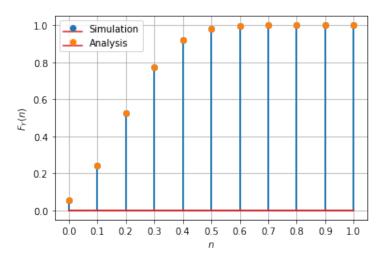


Figure: The CDF distribution of Y_n for n=10