

Agenda: Problems on Convergence of sequences of random variables

1.  $\{x_n : n \geq 1\}$  is an independent sequence of random variables, with marginal pmfs given by

$$P(x_n = \frac{1}{2}(1 - \frac{1}{n})) = P(x_n = \frac{1}{2}(1 + \frac{1}{n})) = \frac{1}{2}.$$

- a) Show whether the sequence converges in mean-squared sense.  
 b) Show whether the sequence converges almost surely.

Ans: a) Intuitively, as  $n \rightarrow \infty$ ,  $x_n$ 's tend to take the value  $\frac{1}{2}$ .

We now demonstrate that this is indeed the case.

We have

$$E[(x_n - \frac{1}{2})^2] = \frac{1}{2} \cdot \frac{1}{4n^2} + \frac{1}{2} \cdot \frac{1}{4n^2} = \frac{1}{4n^2} \xrightarrow{n \rightarrow \infty} 0.$$

Thus,  $x_n \xrightarrow[n \rightarrow \infty]{\text{m.s.}} \frac{1}{2}$ .

b) By Chebyshev's inequality, for every  $\epsilon > 0$

$$P(|x_n - \frac{1}{2}| > \epsilon) \leq \frac{E[(x_n - \frac{1}{2})^2]}{\epsilon^2} = \frac{1}{4n^2 \epsilon^2}$$

$$\Rightarrow \sum_{n=1}^{\infty} P(|x_n - \frac{1}{2}| > \epsilon) \leq \frac{1}{4\epsilon^2} \sum_{n=1}^{\infty} \frac{1}{n^2} < \infty.$$

Thus, by the first part of Borel-Cantelli lemma, we conclude that  $P(\{|x_n - \frac{1}{2}| > \epsilon\} \text{ i.o.}) = 0 \quad \forall \epsilon > 0$ . This in turn

implies that  $X_n \xrightarrow[n \rightarrow \infty]{\text{a.s.}} Y_2$ .

Alternative solution for B :

$$\text{Let } \Omega_1^{(n)} = \{\omega \in \Omega : X_n(\omega) = \frac{1}{2} \left(1 - \frac{1}{n}\right)\}$$

$$\Omega_2^{(n)} = \{\omega \in \Omega : X_n(\omega) = \frac{1}{2} \left(1 + \frac{1}{n}\right)\}$$

$$\text{Clearly, } \Omega_1^{(n)} \cap \Omega_2^{(n)} = \emptyset, \text{ and } P(\Omega_1^{(n)} \cup \Omega_2^{(n)}) = 1.$$

Further, let

$$A_n = \Omega_1^{(n)} \cup \Omega_2^{(n)}.$$

$$\text{Then, } P(A_n) = 1 \quad \forall n \geq 1$$

$$\Rightarrow P\left(\bigcap_{n=1}^{\infty} A_n\right) = 1.$$

$$\text{Let } A = \bigcap_{n=1}^{\infty} A_n. \text{ Then,}$$

$$\omega \in A \Rightarrow \omega \in A_n \quad \forall n \geq 1$$

$$\Rightarrow \omega \in \Omega_1^{(n)} \cup \Omega_2^{(n)} \quad \forall n \geq 1$$

$$\Rightarrow \omega \in \Omega_1^{(n)} \quad \text{OR} \quad \omega \in \Omega_2^{(n)}, \quad \forall n \geq 1$$

$$\Rightarrow X_n(\omega) \in \frac{1}{2} \left(1 - \frac{1}{n}\right) \quad \text{OR} \quad X_n(\omega) = \frac{1}{2} \left(1 + \frac{1}{n}\right), \quad \forall n \geq 1$$

$$\Rightarrow \lim_{n \rightarrow \infty} X_n(\omega) = Y_2.$$

Thus, we have shown that

$$A \subseteq \{\omega \in \Omega : \lim_{n \rightarrow \infty} X_n(\omega) = Y_2\}.$$

$\Rightarrow$

$$P\left(\lim_{n \rightarrow \infty} X_n = Y_2\right) = 1 \quad (\text{since } P(A) = 1).$$

$\Rightarrow X_n \rightarrow Y_2$  almost surely.

2.  $\{X_n : n \geq 2\}$  is an independent sequence of random variables, with marginal pmfs given by

$$P\left(X_n = \frac{1}{2}(1 - Y_n)\right) = P\left(X_n = \frac{1}{2}(1 + Y_n)\right) = \frac{1}{2}\left(1 - \frac{1}{n}\right),$$

$$P(X_n = 1) = \frac{1}{n}.$$

a) Sketch the CDF of  $X_n$ , and show whether the sequence converges in distribution.

Ans:

$$F_{X_n}(x) = \begin{cases} 0, & \text{if } x < \frac{1}{2}\left(1 - \frac{1}{n}\right), \\ \frac{1}{2}\left(1 - \frac{1}{n}\right), & \text{if } \frac{1}{2}\left(1 - \frac{1}{n}\right) \leq x < \frac{1}{2}(1 + Y_n) \\ 1 - \frac{1}{n}, & \text{if } \frac{1}{2}(1 + Y_n) \leq x < 1, \\ 1, & \text{if } x \geq 1. \end{cases}$$

We see that  $X_n \xrightarrow{d} \frac{1}{2}$ .

b) Show whether the sequence converges almost surely.

Ans: We see that

$$\sum_{n=1}^{\infty} P(X_n = 1) = \sum_{n=1}^{\infty} \frac{1}{n} = \infty,$$

and since  $X_n$ 's are independent, by second part of Borel-Cantelli lemma, we have  $P(\{X_n = 1\} \text{ i.o.}) = 1$ . This clearly implies  $X_n \xrightarrow{\text{a.s.}} \frac{1}{2}$ .

3. Let  $U \sim \text{unif}[0,1]$ , and  $\forall n \geq 1$ , let  $X_n = (-1)^n \frac{U}{n}$ .

a) Show whether  $X_n$  converges almost surely.

Ans: We note that

$$|X_n| = \frac{U}{n} \leq \frac{1}{n} \quad \forall n \geq 1, \text{ and hence we get}$$

$$\lim_{n \rightarrow \infty} |X_n| = 0 \quad \text{a.s.}$$

because  $f(x) = |x|$  is a continuous function of  $x$   
 $\Rightarrow f(\lim_{n \rightarrow \infty} X_n) = \lim_{n \rightarrow \infty} f(X_n)$ .

$$\Rightarrow \left| \lim_{n \rightarrow \infty} X_n \right| = 0 \quad \text{a.s.} \Rightarrow \lim_{n \rightarrow \infty} X_n = 0 \quad \text{a.s.}$$

b) Show whether the sequence converges in mean squared sense.

Ans: Note that

$$|X_n| = \frac{U}{n} \leq \frac{1}{n} \leq 1 \quad \text{a.s. for all } n \geq 1$$

$$\Rightarrow P(|X_n| \leq 1) = 1.$$

Also,

$$\begin{aligned} P(|X_n - 0| > \varepsilon) &= P(U > n\varepsilon) \\ &= 0 \quad \forall n \text{ s.t. } \frac{1}{n} < \varepsilon. \end{aligned}$$

$$\Rightarrow P(|X_n - 0| > \varepsilon) \xrightarrow{n \rightarrow \infty} 0 \quad \forall \varepsilon > 0.$$

Thus,  $X_n \xrightarrow{P} 0$ ,  $P(|X_n| \leq 1) \forall n \geq 1 \Rightarrow X_n \xrightarrow{\text{m.s.}} 0$ .

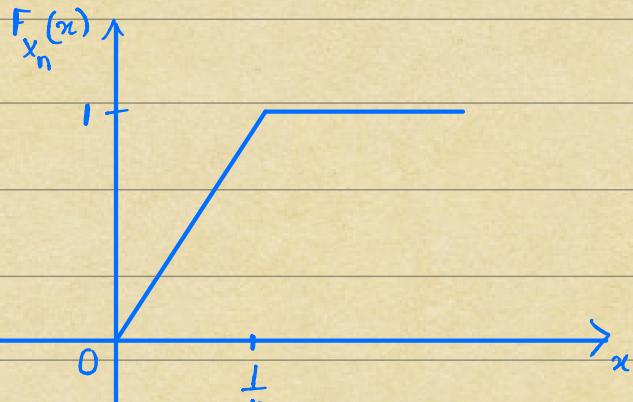
Alternatively,

$$E[X_n^2] = E\left[\frac{U^2}{n^2}\right] = \frac{1/2 + 1/4}{n^2} = \frac{1}{3n^2} \xrightarrow{n \rightarrow \infty} 0.$$

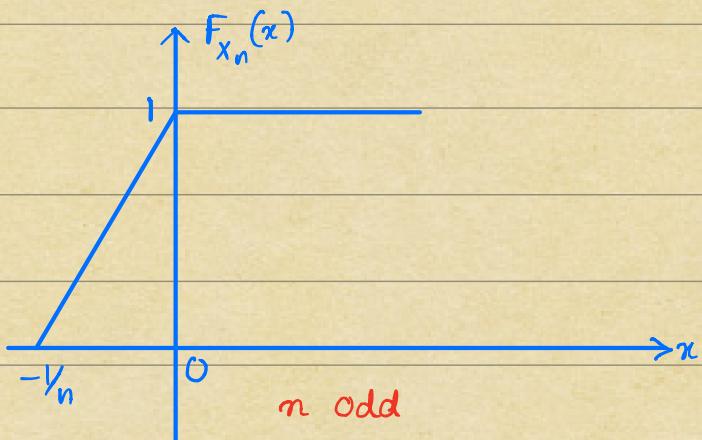
Thus,  $X_n \xrightarrow{\text{m.s.}} 0$ .

c) Demonstrate formally that  $X_n \xrightarrow{d} 0$ .

Ans: The CDF of  $X_n$  is graphed below:



n even



n odd

We see that  $F_{X_n}(x) \rightarrow 0 \quad \forall x < 0,$

$F_{X_n}(x) \rightarrow 1 \quad \forall x > 0.$

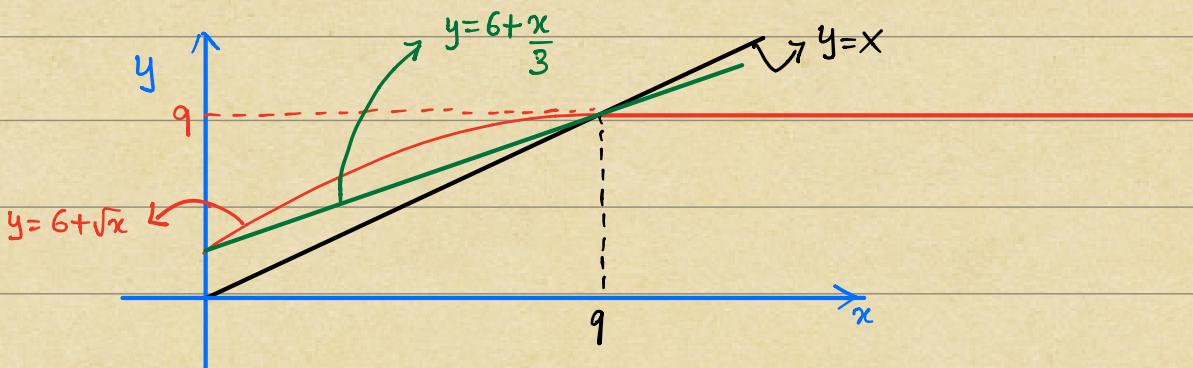
Thus,  $X_n \xrightarrow{d} 0$ , since 0 is the only discontinuity point of the CDF of  $X \equiv 0$ .

4. Let  $X_0$  be a non-negative random variable s.t.  $P(X_0 \geq 0) = 1$ .

$$X_n = 6 + \sqrt{X_{n-1}}, \quad n \geq 1.$$

Show that  $X_n \xrightarrow{\text{a.s.}} g$ .

Ans:



We first notice a couple of points:

a) If  $X_0 = g$ , then  $X_n = g \quad \forall n \geq 1$ .

b) If  $X_0 < g$ , then  $X_n \leq X_{n+1}, \quad \forall n \geq 1$ , and  $X_n$  seems to increase to  $g$ .

c) If  $x_0 > g$ , then  $x_n \geq x_{n+1} \forall n \geq 1$ , and  $x_n$  seems to decrease towards  $g$ .

Formally, we note that

$$6 + \sqrt{x} - g \geq 6 + \frac{x}{3} - g \quad \forall x \leq g,$$

$$6 + \sqrt{x} - g \leq 6 + \frac{x}{3} - g \quad \forall x \geq g,$$

and thus, for all  $x \geq 0$ ,

$$|6 + \sqrt{x} - g| \leq |6 + \frac{x}{3} - g|.$$

(for small  $x$ ,  $\sqrt{x}$  will be larger than an appropriately scaled version of  $x$ )

( $\sqrt{x}$  grows slower than  $x$  for large  $x$ . That is the idea)

Hence,

$$\begin{aligned} |x_n - g| &\leq |6 + \frac{x_{n-1}}{3} - g| \\ &= \frac{1}{3} |x_{n-1} - g| \end{aligned}$$

By induction, we get

$$|x_n - g| \leq \frac{1}{3^n} |x_0 - g|.$$

Hence,

$$\lim_{n \rightarrow \infty} |x_n(\omega) - g| = 0 \quad \forall \omega \text{ s.t. } x_0(\omega) \geq 0.$$

Hence,  $x_n \xrightarrow{\text{a.s.}} g$ .

5. Let  $\{U_n : n \geq 1\} \stackrel{\text{iid}}{\sim} \text{unif}[0, 1]$ . Define

$$X_n = \min\{U_1, \dots, U_n\}.$$

a) Show whether  $X_n$  converges almost surely.

Ans: Intuitively, if the one of the  $U_n$ 's is 0, then

$X_n = 0$  for all  $n$  after some stage. Also,

We have  $X_n \geq X_{n+1} \forall n \geq 1$ . Thus, as a first guess,

We can take 0 to be a possible limit for convergence.

Let us first check for convergence in prob. to 0. We have

$$P(X_n > \varepsilon) = P(\min\{U_1, \dots, U_n\} > \varepsilon)$$

$$= \begin{cases} (1-\varepsilon)^n, & \text{if } 0 < \varepsilon < 1 \\ 0, & \text{o.w.} \end{cases}$$

$$\xrightarrow[n \rightarrow \infty]{} 0 \quad \forall \varepsilon > 0.$$

Thus, indeed,  $X_n \xrightarrow{P} 0$ . Furthermore,

$$\sum_{n=1}^{\infty} P(X_n > \varepsilon) < \infty \quad \forall \varepsilon > 0.$$

Thus, by the first part of Borel - Cantelli lemma, we get

$$P(\{X_n > \varepsilon\} \text{ i.o.}) = 0 \quad \forall \varepsilon > 0$$

$$\Rightarrow X_n \xrightarrow{\text{a.s.}} 0.$$

b) Let  $y_n = n \times \min\{U_1, \dots, U_n\}$ ,  $n \geq 1$ .

Show that  $y_n$  converges in distribution. What is the limit distribution?

Ans: Clearly,  $y_n$ 's are non-negative random variables.

Furthermore, the support of  $y_n$  is  $[0, n]$ , which grows to  $[0, \infty)$  as  $n \rightarrow \infty$ . Thus, the limit distribution must be a distribution on  $[0, \infty)$ .

Fix any arbitrary  $y \geq 0$ . Then, there exists  $n_0$  large

enough st.  $\frac{y}{n_0} \leq 1 \quad \forall n \geq n_0$  (this  $n_0$  may depend on the choice of  $y$ ).

$$\Rightarrow P\left(y_n > y\right) = \left(1 - \frac{y}{n}\right)^n \quad \forall n \geq n_0$$

$$\Rightarrow \lim_{n \rightarrow \infty} P\left(y_n > y\right) = e^{-y}, \quad \forall y \geq 0.$$

Thus,  $y_n \rightarrow y$ , where  $y \sim \text{Exp}(1)$ .

6. Let  $w_1, w_2, \dots$  be a sequence of iid  $N(0, \sigma^2)$  random variables. Let  $x_0 = 0$ , and define

$$x_{n+1} = \frac{x_n + w_{n+1}}{2}, \quad n \geq 0.$$

Argue that  $x_n$  converges in distribution. what is the limit distribution?

Ans: First, we note that for each  $n \geq 1$ , we have

$$x_n = \frac{w_1}{4^n} + \frac{w_2}{4^{n-1}} + \dots + \frac{w_n}{4}.$$

Thus, it follows that  $x_n \sim N\left(0, \sigma^2\left(\frac{1}{4^n} + \dots + \frac{1}{4}\right)\right)$ .

Therefore,

$$P(x_n \leq x) = P\left(\frac{x_n}{\sqrt{\text{Var } x_n}} \leq \frac{x}{\sqrt{\text{Var } x_n}}\right)$$

$$= \Phi\left(\frac{x}{\sqrt{\text{Var } x_n}}\right).$$

$$\text{Thus, } \lim_{n \rightarrow \infty} P(x_n \leq x) = \lim_{n \rightarrow \infty} \Phi\left(\frac{x}{\sqrt{\text{Var } x_n}}\right)$$

$$= \phi \left( \lim_{n \rightarrow \infty} \frac{x}{\sqrt{\text{Var } X_n}} \right) \quad (\because \phi \text{ is a continuous function})$$

Now,

$$\begin{aligned} \lim_{n \rightarrow \infty} \text{Var}(X_n) &= \lim_{n \rightarrow \infty} \sigma^2 \left( \frac{1}{4} + \frac{1}{4^2} + \dots + \frac{1}{4^n} \right) \\ &= \sigma^2 \cdot \sum_{n=1}^{\infty} \frac{1}{4^n} \\ &= \sigma^2 / 3. \end{aligned}$$

$$\text{Thus, } \lim_{n \rightarrow \infty} P(X_n \leq x) = \phi \left( \frac{x}{\sigma/3} \right).$$

$$\text{Thus, } X_n \xrightarrow{d} X, \text{ where } X \sim N(0, \frac{\sigma^2}{3}).$$


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