# Assignment 5

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Download all python codes from

https://github.com/GouthamSai22/AI1103/blob/ main/Assignment5/Codes

and latex-tikz codes from

https://github.com/GouthamSai22/AI1103/blob/ main/Assignment5/main.tex

# 1 PROBLEM 103 FROM CSIR UGC NET - DEC 2012

Let  $\{X_n : n > 0\}$  and X be random variables defined on a common probability space. Further assume that  $\{X_n \ge 0 \ \forall n > 0\}$  and

$$\Pr(X = x) = \begin{cases} p, & x = 0\\ 1 - p, & x = 1\\ 0, & \text{otherwise} \end{cases}$$
 (1.0.1)

where,  $0 \le p \le 1$ . Which of the following statements are necessarily true?

1) If 
$$p = 0$$
 and  $X_n \xrightarrow{d} X$ ,  $X_n \xrightarrow{p} X$ 

2) If 
$$p = 1$$
 and  $X_n \xrightarrow{d} X$ , then  $X_n \xrightarrow{p} X$ 

2) If 
$$p = 1$$
 and  $X_n \longrightarrow X$ , then  $X_n \longrightarrow X$   
3) If  $0 and  $X_n \stackrel{d}{\longrightarrow} X$ , then  $X_n \stackrel{p}{\longrightarrow} X$   
4) If  $X_n \stackrel{p}{\longrightarrow} X$ , then  $X_n \stackrel{a.s}{\longrightarrow} X$$ 

4) If 
$$X_n \xrightarrow{p} X$$
, then  $X_n \xrightarrow{a.s} X$ 

### 2 Solution

For any given sequence of real-valued random variables  $X_1, X_2, X_3, \dots, X_n$  and a random variable X, let us look at some definitions

**Definition 1.** Convergence in Distribution or Weak convergence:

$$\lim_{n \to \infty} F_{X_n}(x) = F_X(x)$$
 (2.0.1)

where  $F_{X_n}$  and  $F_X$  are the cumulative probability distribution functions of  $X_n$  and X respectively.

**Definition 2.** Convergence in Probability:

$$\lim_{n \to \infty} \Pr(|X_n - X| > \epsilon) = 0 \,\forall \epsilon > 0 \tag{2.0.2}$$

This is stronger than the convergence in distribution but weaker than Almost sure convergence.

**Definition 3.** Almost sure Convergence:

$$\Pr\left(\lim_{n\to\infty} X_n = X\right) = 1\tag{2.0.3}$$

This type of convergence is stronger than both convergence in distribution and probability.

• In general, stronger statements imply weaker statements but not vice versa, i.e. Convergence in probability implies convergence in distribution and Almost sure convergence implies convergence in probability.

**Lemma 1.** Convergence in distribution implies convergence in probability if X is a constant.

#### **Proof**:

- Let  $\epsilon > 0$ . Let X = c and the sequence  $X_1, X_2, X_3, \ldots, X_n$  converges to X in distribu-
- Let

$$S \rightarrow$$
 sphere centered at c, of radius  $\epsilon$  (2.0.4)

$$S' \rightarrow \text{complement of } S$$
 (2.0.5)

$$\implies \Pr(|X_n - c| > \epsilon) = \Pr(X_n \in S') \quad (2.0.6)$$

• From the Portmanteaus lemma,

**Lemma 2.** The sequence  $X_1, X_2, X_3, \ldots, X_n$ converges in distribution to X if and only if

$$\limsup \Pr(X_n \in F) \le \Pr(X \in F) \qquad (2.0.7)$$

for every closed set F;

$$\lim_{n \to \infty} X_n = c \tag{2.0.8}$$

$$\implies \limsup_{n \to \infty} \Pr(X_n \in S') \le \Pr(c \in S')$$
(2.0.9)

$$Pr(c \in S') = 0$$
 (By defn) (2.0.10)

• Thus,

$$\lim_{n \to \infty} \Pr(|X_n - c| > \epsilon) = \lim_{n \to \infty} \Pr(X_n \in S')$$
(2.0.11)
$$\leq \limsup_{n \to \infty} \Pr(X_n \in S')$$
(2.0.12)
$$\leq \Pr(c \in S')$$
(2.0.13)
$$\leq 0$$
(2.0.14)
$$= 0$$
(2.0.15)

(Since probability cannot be negative)

• Thus, by definition,

Pr 
$$(|X_n - c| > \epsilon) = 0$$
 for any  $\epsilon > 0$  given,  

$$\lim_{n \to \infty} F_{X_n}(x) = F_X(x) \text{ and } X \text{ is constant}$$
(2.0.16)

Let us look at each option one after another.

1) Given,

$$p = 0 \implies X = 1$$

Since X is a constant, from Lemma 1, we can say that option 1 is true.

2) Given,

$$p = 1 \implies X = 0$$

Since X is a constant, from Lemma 1, we can say that option 2 is true.

3) Given,

$$0$$

Since X is not a constant, we can say that option 3 is false.

4) Since Convergence in probability is weaker than Almost sure convergence, we can say that option 4 is false as a weaker statement does not imply a stronger statement.