#### 1

# 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  $X_n \xrightarrow{d} X$ , then  $X_n \xrightarrow{p} X$ 

3) If 
$$0 and  $X_n \xrightarrow{d} X$ , then  $X_n \xrightarrow{p} X$   
4) If  $X_n \xrightarrow{p} X$ , then  $X_n \xrightarrow{a.s} X$$$

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, 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

1) Convergence in Distribution or Weak convergence:

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

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

2) Convergence in Probability:

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

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

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  $X_1, X_2, X_3, \dots, X_n$  converges to X in distribution.
- Let

$$B_c(\epsilon) \rightarrow$$
 open ball about c of radius  $\epsilon$  (2.0.4)

$$B_c(\epsilon)' \to \text{complement of } B_c(\epsilon) \quad (2.0.5)$$

$$\implies \Pr(|X_n - c| > \epsilon) = \Pr(X_n \in B_c(\epsilon)')$$
(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;

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

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

$$\Pr(c \in B_c(\epsilon)') = 0 \text{ (By defn)}$$
(2.0.10)

· Thus,

$$\lim_{n \to \infty} \Pr(|X_n - c| > \epsilon) = \lim_{n \to \infty} \Pr(X_n \in B_c(\epsilon)')$$

$$(2.0.11)$$

$$\leq \lim_{n \to \infty} \sup \Pr(X_n \in B_c(\epsilon)')$$

$$(2.0.12)$$

$$\leq \Pr(c \in B_c(\epsilon)')$$

$$(2.0.13)$$

$$\leq 0$$

$$(2.0.14)$$

$$= 0$$

$$(2.0.15)$$

(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.