AUA CS 108, Statistics, Fall 2019 Lecture 15

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Contents

► Convergence Types of R.V. Sequences

Last Lecture ReCap

▶ Well, since this will be a Recap of a Recap, we can skip it ¨

Additions to Important Distributions

▶ See many other Distributions at Wiki or in different textbooks.

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- ► Another page for the Relationship: L. Leemis Page

Convergence of a sequence of

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We toss a coin, infinitely many times, and let X_k be 0, it the k-th toss resulted in Heads, and $X_k = 1$ otherwise.

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I know, almost all examples are examples of *finite* sequences, but for theoretical (and practical) reasons we can assume they are infinite.

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Also, we will use a lot the CLT, say, to construct Confidence Intervals and design Tests for Hypotheses, so we need to know the CLT, and CLT is about the limit of a r.v. sequence.

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This is because, a sequence of r.v., besides being just a sequence of functions¹, also encloses randomness behind, and we need to deal with that randomness.

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Aha, that's the problem - it is not so easy to define the closedness

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Definition: We will say that $X_n \to X$ almost sure, and we will write $X_n \to X$ a.s. or $X_n \xrightarrow{a.s.} X$, if

$$\mathbb{P}\left(\omega \in \Omega : \lim_{n \to +\infty} X_n(\omega) = X(\omega)\right) = 1,$$

or, for short,

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Equivalently, we can write

$$X_n \xrightarrow{a.s.} X$$
 iff $\mathbb{P}(X_n \not\to X) = 0$.

Convergence in Probability

Definition: We will say that $X_n \to X$ in **Probability**, and we will write $X_n \stackrel{\mathbb{P}}{\longrightarrow} X$, if

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Equivalently, we can write

$$X_n \stackrel{\mathbb{P}}{\longrightarrow} X$$
 iff $\mathbb{P} \Big(|X_n - X| < \varepsilon \Big) \to 1$ for any $\varepsilon > 0$.

Convergence in the Mean Square Sence

Definition: We will say that $X_n \to X$ in the Quadratic Mean Sense or in L^2 (or in the Mean Square Sense), and we will write $X_n \xrightarrow{L^2} X$ or $X_n \xrightarrow{qm} X$, if

$$MSE(X_n, X) = \mathbb{E}((X_n - X)^2) \to 0, \quad \text{when} \quad n \to \infty.$$

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Here $MSE(X_n, X)$ is the Mean Square Error (of the approximation of X by X_n).

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 as $n \to \infty$ at any point of continuity x of $F_X(x)$.

Remark: This is equivalent to saying that for (almost) any subsets $A \subset \mathbb{R}$

$$\mathbb{P}(X_n \in A) \to \mathbb{P}(X \in A).$$

Example: Assume I am tossing a fair coin infinitely many times (independently), and let X_n be 1 if Head shows in the n-th trial, and 0 otherwise. So the Distribution of X_n is



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Solution: OTB. Not a good/correct example, impossible to answer to the questions except to the first one.

Example: Assume X_n is a Discrete r.v. with the following PMF, defined on the same Probability Space:

$$\frac{X_n \mid 3 + \frac{1}{n^2} \mid n}{\mathbb{P}(X_n = x) \mid 1 - \frac{1}{n} \mid \frac{1}{n}.}$$

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$$\begin{array}{c|c} X_n & 3 + \frac{1}{n^2} & n \\ \hline \mathbb{P}(X_n = x) & 1 - \frac{1}{n} & \frac{1}{n}. \end{array}$$

Which of the followings are true (use only the definitions):

- $X_n \stackrel{\mathbb{P}}{\longrightarrow} 3;$
- $ightharpoonup X_n \stackrel{qm}{\longrightarrow} 3;$
- $\longrightarrow X_n \xrightarrow{D} 3$?

Solution: OTB