

# Probability Theory and Random Processes (MA225)

LECTURE SLIDES  
Lecture 18



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# Modes of Convergence

- In probability and statistics, it is often necessary to consider the **distribution of a random variable that is itself a function of several random variables**, for example,  $Y = g(X_1, \dots, X_n)$ .
- For example, the **sample mean** of random variables  $X_1, \dots, X_n$ .
- Unfortunately, **finding the distribution exactly is often very difficult** or very time-consuming even if the joint distribution of the random variables is known exactly.
- What is the **distribution of odds-ratio (OR)**? How to find it?
- In other cases, we may have only **partial information about the joint distribution** of  $X_1, \dots, X_n$  in which case it is impossible to determine the distribution of  $Y$ .
- However, when **n is large**, it may be **possible to obtain approximations to the distribution** of  $Y$  even when only partial information about  $X_1, \dots, X_n$  is available.
- In many cases, **these approximations can be remarkably accurate**.

# Modes of Convergence

Let  $\{X_n\}$  be a sequence of random variables defined on a probability space  $(\mathcal{S}, \mathcal{F}, P)$ . Let  $X$  be a random variable defined on the same probability space  $(\mathcal{S}, \mathcal{F}, P)$ .

**Def: (Almost sure convergence)** We say that  $X_n$  converges almost surely or with probability 1 to a random variable  $X$  if

$$P(\omega \in \mathcal{S} : X_n(\omega) \rightarrow X(\omega)) = 1.$$

**Example 1:** Let  $\mathcal{S} = [0, 1]$ ,  $\mathcal{F} = \mathcal{B}([0, 1])$  and  $P$  be the uniform measure. Define  $X_n = 1_{[0, \frac{1}{n}]}$ . Then  $X_n$  converges almost surely (w. p. 1) to the zero random variable.

is it true for  $x=0$ ?

i bethink not

actually  $P(w \text{ belongs to } (0,1]) = 1$  only

**Theorem:** Let  $\{X_n\}$  and  $\{Y_n\}$  be two sequences of random variables defined on a probability space  $(\mathcal{S}, \mathcal{F}, P)$ . Suppose  $X_n \rightarrow X$  w. p. 1 and  $Y_n \rightarrow Y$  w. p. 1. Then

- $X_n + Y_n \rightarrow X + Y$  w. p. 1.
- $X_n Y_n \rightarrow XY$  w. p. 1.
- $f(X_n) \rightarrow f(X)$  w. p. 1, for any  $f$  continuous.

**Def: (Convergence in probability)** We say that  $X_n$  converges in probability to a random variable  $X$  if for any  $\epsilon > 0$ ,

$$P(|X_n - X| > \epsilon) \rightarrow 0 \text{ as } n \rightarrow \infty.$$

**Example 2:** Let  $\mathcal{S} = [0, 1]$ ,  $\mathcal{F} = \mathcal{B}([0, 1])$  and  $P$  be the uniform measure. Define  $X_n = n1_{[0, \frac{1}{n}]}$ . Then  $X_n$  converges in probability to the zero random variable.

**Theorem:** Let  $\{X_n\}$  and  $\{Y_n\}$  be two sequences of random variables defined on a probability space  $(\mathcal{S}, \mathcal{F}, P)$ . Suppose  $X_n \rightarrow X$  in probability and  $Y_n \rightarrow Y$  in probability. Then

- $X_n + Y_n \rightarrow X + Y$  in probability.
- $X_n Y_n \rightarrow XY$  in probability.
- $f(X_n) \rightarrow f(X)$  in probability, for any  $f$  continuous.

**Def: (Convergence in  $r^{th}$  mean)** We say that  $X_n$  converges in  $r^{th}$  mean to a random variable  $X$  if

$$E|X_n - X|^r \rightarrow 0 \text{ as } n \rightarrow \infty.$$

**Example 3:** Let  $\mathcal{S} = [0, 1]$ ,  $\mathcal{F} = \mathcal{B}([0, 1])$  and  $P$  be the uniform measure. Define  $X_n = 1_{[0, \frac{1}{n}]}$ . Then  $X_n$  converges in  $r^{th}$  mean to the zero random variable.

**Theorem:** Let  $\{X_n\}$  and  $\{Y_n\}$  be two sequences of random variables defined on a probability space  $(\mathcal{S}, \mathcal{F}, P)$ .

- If  $X_n \rightarrow X$  in  $r^{th}$  mean and  $Y_n \rightarrow Y$  in  $r^{th}$  mean, then  $X_n + Y_n \rightarrow X + Y$  in  $r^{th}$  mean.
- If  $X_n \rightarrow X$  in  $r^{th}$  mean then  $f(X_n) \rightarrow f(X)$  in  $r^{th}$  mean, for any  $f$  bounded continuous.

**Def: (Convergence in distribution)** We say that  $X_n$  converges in distribution to a random variable  $X$  if

$$F_n(x) \rightarrow F(x) \text{ as } n \rightarrow \infty.$$

for all  $x$  where  $F$  is continuous. Here  $F_n$ s are the distribution functions of  $X_n$ s and  $F$  is the distribution function of  $X$ .

**Remark:** Unlike the first three modes of convergence, here  $X_n$ s can be defined on different probability spaces. We are only interested in the distribution functions. This flexibility makes this mode of convergence very useful.

**Example 4:** Suppose  $X_n$ s are random variables such that  $P(X_n = 1/n) = 1$ . Then  $X_n$  converges in distribution to the zero random variable.

**Theorem:** Let  $\{X_n\}$  and  $\{Y_n\}$  be two sequences of random variables defined on a probability space  $(\mathcal{S}, \mathcal{F}, P)$ . Suppose  $X_n \rightarrow X$  in distribution and  $Y_n \rightarrow c$  in probability for some constant  $c$ . Then

- $X_n + Y_n \rightarrow X + c$  in distribution.
- $X_n Y_n \rightarrow cX$  in distribution.
- $f(X_n) \rightarrow f(X)$  in distribution, for any  $f$  continuous.

**Important:** If  $X_n$  converges to  $X$  in distribution and  $Y_n$  converges to  $Y$  in distribution then  $X_n + Y_n$  may not converge to  $X + Y$  in distribution. Same for product.