

## AMS 241: Bayesian Nonparametric Methods – Spring 2018

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### Modes of convergence for sequences of random variables

Given a sequence of random variables  $\{X_n : n \geq 1\}$  and a limiting random variable  $X$ , there are several ways to formulate convergence “ $X_n \rightarrow X$  as  $n \rightarrow \infty$ ”. The following four definitions are commonly used to study limiting results for random variables and stochastic processes.

**Almost sure convergence** ( $X_n \rightarrow^{\text{a.s.}} X$ ).

Let  $\{X_n : n \geq 1\}$  and  $X$  be random variables defined on some probability space  $(\Omega, \mathcal{F}, P)$ .  $\{X_n : n \geq 1\}$  converges almost surely to  $X$  if

$$P\left(\left\{\omega \in \Omega : \lim_{n \rightarrow \infty} X_n(\omega) = X(\omega)\right\}\right) = 1.$$

**Convergence in  $r$ th mean** ( $X_n \rightarrow^{r\text{-mean}} X$ ).

Let  $\{X_n : n \geq 1\}$  and  $X$  be random variables defined on some probability space  $(\Omega, \mathcal{F}, P)$ .  $\{X_n : n \geq 1\}$  converges in mean of order  $r \geq 1$  (or in  $r$ th mean) to  $X$  if  $E(|X_n|^r) < \infty$  for all  $n$ , and

$$\lim_{n \rightarrow \infty} E(|X_n - X|^r) = 0.$$

**Convergence in probability** ( $X_n \rightarrow^P X$ ).

Let  $\{X_n : n \geq 1\}$  and  $X$  be random variables defined on some probability space  $(\Omega, \mathcal{F}, P)$ .  $\{X_n : n \geq 1\}$  converges in probability to  $X$  if for any  $\epsilon > 0$ ,

$$\lim_{n \rightarrow \infty} P(\{\omega \in \Omega : |X_n(\omega) - X(\omega)| > \epsilon\}) = 0.$$

**Convergence in distribution** ( $X_n \rightarrow^d X$ ).

Let  $\{X_n : n \geq 1\}$  and  $X$  be random variables with distribution functions  $\{F_n : n \geq 1\}$  and  $F$ , respectively.  $\{X_n : n \geq 1\}$  converges in distribution to  $X$  if

$$\lim_{n \rightarrow \infty} F_n(x) = F(x),$$

for all points  $x$  at which  $F$  is continuous.

Note that the first three types of convergence require that  $X_n$  and  $X$  are all defined on the same probability space, as they include statements involving the (common) probability measure  $P$ . However, convergence in distribution applies to random variables defined possibly on different probability spaces, as it only involves the corresponding distribution functions.

It can be shown that:

Almost sure convergence implies convergence in probability.

Convergence in  $r$ th mean implies convergence in probability, for any  $r \geq 1$ .

Convergence in probability implies convergence in distribution.

Convergence in  $r$ th mean implies convergence in  $s$ th mean, for  $r > s \geq 1$ .

No other implications hold without further assumptions on  $\{X_n : n \geq 1\}$  and/or  $X$ .

## Convergence theorems for expectations

**Monotone convergence theorem:** Consider a countable sequence  $\{X_n : n = 1, 2, \dots\}$  of  $\overline{\mathbb{R}}^+$ -valued random variables defined on the same probability space  $(\Omega, \mathcal{F}, P)$ . Assume that the sequence is pointwise (or almost surely) increasing, that is, for all  $n$ ,  $X_n(\omega) \leq X_{n+1}(\omega)$  for all  $\omega \in \Omega$  (or all  $\omega$  in an event of probability 1). Denote by  $X$  the pointwise (or almost sure) limit of the sequence  $\{X_n : n = 1, 2, \dots\}$ .

- Then,  $\lim_{n \rightarrow \infty} E(X_n) = E(X)$ .

**Dominated convergence theorem:** Consider a countable sequence  $\{X_n : n = 1, 2, \dots\}$  of  $\overline{\mathbb{R}}$ -valued random variables defined on the same probability space  $(\Omega, \mathcal{F}, P)$ . Assume there exists a random variable  $Y$  (also defined on  $(\Omega, \mathcal{F}, P)$ ) such that  $|X_n| \leq Y$ , almost surely for all  $n$ , and  $E(Y) < \infty$ .

- Then,

$$-\infty < E(\liminf_{n \rightarrow \infty} X_n) \leq \liminf_{n \rightarrow \infty} E(X_n) \leq \limsup_{n \rightarrow \infty} E(X_n) \leq E(\limsup_{n \rightarrow \infty} X_n) < \infty$$

In addition to the assumptions  $|X_n| \leq Y$ , almost surely for all  $n$ , and  $E(Y) < \infty$ , assume that the sequence  $\{X_n : n = 1, 2, \dots\}$  converges almost surely to random variable  $X$  (also defined on  $(\Omega, \mathcal{F}, P)$ ).

- Then,  $E(|X|) < \infty$ ,  $\lim_{n \rightarrow \infty} E(X_n) = E(X)$ , and  $\lim_{n \rightarrow \infty} E(|X_n - X|) = 0$ .

**Bounded convergence theorem:** Consider a countable sequence  $\{X_n : n = 1, 2, \dots\}$  of  $\overline{\mathbb{R}}$ -valued random variables defined on the same probability space  $(\Omega, \mathcal{F}, P)$ . Assume that the sequence converges almost surely to random variable  $X$  (also defined on  $(\Omega, \mathcal{F}, P)$ ) and that  $|X_n| \leq M$ , almost surely for all  $n$ , where  $M$  is a finite constant.

- Then,  $E(|X|) \leq M$ ,  $\lim_{n \rightarrow \infty} E(X_n) = E(X)$ , and  $\lim_{n \rightarrow \infty} E(|X_n - X|) = 0$ .