

Notions of asymptotic theory *

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1. Stochastic convergence

1.1 Definition Let $\{X_n = X_n(\omega) : n = 1, 2, \dots\}$ a sequence of real r.v.'s defined on a probability space (Ω, \mathcal{A}, P) and $X = X(\omega)$ another real r.v. defined on the same space.

(a) X_n converges in probability to X as $n \rightarrow \infty$ (denoted $X_n \xrightarrow{p} X$) iff

$$\lim_{n \rightarrow \infty} P[|X_n - X| > \varepsilon] = 0, \forall \varepsilon > 0. \quad (1.1)$$

(b) X_n converges almost surely to X as $n \rightarrow \infty$ (denoted $X_n \xrightarrow{a.s.} X$) iff

$$P\left[\lim_{n \rightarrow \infty} X_n = X\right] = 1. \quad (1.2)$$

(c) X_n converges completely to X as $n \rightarrow \infty$ (denoted $X_n \xrightarrow{c} X$) iff

$$\sum_{n=1}^{\infty} P[|X_n - X| > \varepsilon] < \infty, \forall \varepsilon > 0. \quad (1.3)$$

(d) Suppose $E|X_n|^r < \infty, \forall n$, where $r > 0$. X_n converges in mean of order r to X (denoted $X_n \xrightarrow{r} X$) iff

$$\lim_{n \rightarrow \infty} E[|X_n - X|^r] = 0. \quad (1.4)$$

In this case, we also say that X_n converges to X in L_r . If $r = 2$, we say X_n converges to X in quadratic mean (q.m.).

(e) Let $F_n(x)$ and $F(x)$ be the distribution functions of X_n and X respectively. X_n converges in law (or in distribution) to X as $n \rightarrow \infty$ (denoted $X_n \xrightarrow{L} X$) iff

$$\lim_{n \rightarrow \infty} F_n(x) = F(x) \text{ at all continuity points of } F(x). \quad (1.5)$$

1.2 Proposition UNICITY OF PROBABILITY LIMIT. Let $\{X_n : n = 1, 2, \dots\}$ be a sequence of real r.v.'s defined on a probability space (Ω, \mathcal{A}, P) , and let X and Y be two real r.v.'s defined on the same probability space. Then

$$X_n \xrightarrow{p} X \text{ and } X_n \xrightarrow{p} Y \Rightarrow P[X \neq Y] = 0. \quad (1.6)$$

2. Relations between convergence concepts

2.1 Assumption Let $\{X_n\} \equiv \{X_n : n = 1, 2, \dots\}$ be a sequence of real r.v.'s defined on a probability space (Ω, \mathcal{A}, P) and X another real r.v. defined on the same space.

Unless stated otherwise, this assumption will hold for all the definitions, propositions and theorems in this section.

2.2 Proposition RELATIONS BETWEEN CONVERGENCE CONCEPTS.

$$(a) \quad X_n \xrightarrow{a.s.} X \Leftrightarrow \sup_{k \geq n} |X_k - X| \xrightarrow{p} 0 \text{ as } n \rightarrow \infty$$

$$\Leftrightarrow \lim_{n \rightarrow \infty} P \left[\sup_{k \geq n} |X_k - X| > \varepsilon \right] = 0, \forall \varepsilon > 0.$$

$$(b) \quad X_n \xrightarrow{c} X \Rightarrow X_n \xrightarrow{a.s.} X \Rightarrow X_n \xrightarrow{p} X \Rightarrow X_n \xrightarrow{L} X.$$

$$(c) \quad X_n \xrightarrow{r} X \Rightarrow X_n \xrightarrow{s} X \text{ for all } s \text{ such that } 0 < s \leq r \Rightarrow X_n \xrightarrow{p} X \Rightarrow X_n \xrightarrow{L} X.$$

2.3 Remark In general, the implications in **2.2** (b) and (c) cannot be reversed.

2.4 Proposition MOMENT CONDITION FOR CONVERGENCE IN MEAN OF ORDER r .

$$\sum_{n=1}^{\infty} E[|X_n - X|^r] < \infty \text{ for some } r > 0$$

$$\Rightarrow X_n \xrightarrow{c} X \text{ and } X_n \xrightarrow{r} X.$$

2.5 Proposition SUFFICIENT CONDITION FOR COMPLETE CONVERGENCE. Let $g : \mathbb{R} \rightarrow \mathbb{R}$ be a nondecreasing monotonic function for $x \geq 0$, such that $g(x) \geq 0$ and $g(x) = g(-x)$. Then

$$\sum_{n=1}^{\infty} E[g(X_n - X)] < \infty \Rightarrow X_n \xrightarrow{c} X.$$

2.6 Proposition $X_n \xrightarrow{p} X \Rightarrow$ *there is a subsequence $\{X_{n_k} : k = 1, 2, \dots\} \subseteq \{X_n\}$ such that $X_{n_k} \xrightarrow[k \rightarrow \infty]{a.s.} X$ and*

$$\sum_{k=1}^{\infty} P \left[|X_{n_k} - X| \geq \frac{1}{2^k} \right] < \infty .$$

2.7 Proposition $X_n \xrightarrow{p} X \Leftrightarrow$ *each subsequence $\{X_{n_k}\}$ of $\{X_n\}$ contains a subsequence $\{X_{m_k}\} \subseteq \{X_{n_k}\}$ such that $X_{m_k} \xrightarrow{a.s.} X$.*

2.8 Proposition *Let $\{X_n\}$ a sequence of non-negative r.v.'s such that $X_n \leq X_{n+1}, \forall n$, or $X_n \geq X_{n+1}, \forall n$. Then*

$$X_n \xrightarrow{p} X \Rightarrow X_n \xrightarrow{a.s.} X .$$

2.9 Definition SEQUENCE OF UNIFORMLY CONTINUOUS INTEGRALS. *The integrals of the r.v.'s $\{X_n : n = 1, 2, \dots\}$ are uniformly continuous iff*

$$\sup_{n \geq 1} \int_A X_n dP \rightarrow 0 \text{ as } P(A) \rightarrow 0 .$$

2.10 Definition UNIFORMLY INTEGRABLE SEQUENCE. *Let $B_n = \{\omega \in \Omega : |X_n| \geq a\}$. The sequence of r.v.'s $\{X_n : n = 1, 2, \dots\}$ is uniformly integrable iff*

$$\sup_{n \geq 1} \int_{B_n} X_n dP \rightarrow 0 \text{ as } a \rightarrow \infty .$$

2.11 Proposition $X_n \xrightarrow{r} X \Leftrightarrow X_n \xrightarrow{p} X$ *and at least one of the three following conditions holds:*

- (a) $E(|X_n|^r) \rightarrow E[|X|^r]$;
- (b) *the sequence of r.v.'s $\{|X_n|^r : n = 1, 2, \dots\}$ is uniformly integrable;*
- (c) *the integrals of the r.v.'s $\{|X_n|^r : n = 1, 2, \dots\}$ or those of $\{|X_n - X|^r : n = 1, 2, \dots\}$ are uniformly continuous.*

2.12 Proposition *If $E|X_n|^r \leq c < \infty$, $\forall n$, for $r > 0$, then*

$$X_n \xrightarrow{p} X \Rightarrow X_n \xrightarrow{r'} X \text{ for all } 0 < r' < r.$$

2.13 Proposition *If $|X_n| \leq Y$, $\forall n$, and $E(|Y|^r) < \infty$, then*

$$X_n \xrightarrow{p} X \Rightarrow X_n \xrightarrow{r} X \text{ and } E(|X|^r) < \infty.$$

2.14 Proposition *If $|X_n| \leq c < \infty$, $\forall n$, then*

$$X_n \xrightarrow{p} X \Rightarrow X_n \xrightarrow{r} X \text{ and } E(|X|^r) < \infty \text{ for all } r > 0.$$

2.15 Proposition *If the r.v.'s X_n are independent, then*

$$X_n \xrightarrow{a.s.} X \Rightarrow X_n \xrightarrow{c} X.$$

2.16 Proposition *If $P[X = d] = 1$, then*

$$X_n \xrightarrow{p} d \Leftrightarrow X_n \xrightarrow{L} d.$$

2.17 Proposition CAUCHY CRITERION.

(a) *There is a r.v. X such that $X_n \xrightarrow{p} X$*

$$\Leftrightarrow \lim_{m, n \rightarrow \infty} P[|X_m - X_n| > \varepsilon] = 0, \forall \varepsilon > 0$$

$$\Leftrightarrow \lim_{n \rightarrow \infty} \left\{ \sup_{k \geq 1} P[|X_{n+k} - X_n| > \varepsilon] \right\} = 0, \forall \varepsilon > 0.$$

(b) *There is a r.v. X such that $X_n \xrightarrow{a.s.} X$*

$$\begin{aligned} &\Leftrightarrow \lim_{n \rightarrow \infty} P \left[\sup_{k, \ell \geq n} |X_k - X_\ell| > \varepsilon \right] = 0, \forall \varepsilon > 0 \\ &\Leftrightarrow \lim_{n \rightarrow \infty} P \left[\sup_{k \geq 1} |X_{n+k} - X_n| > \varepsilon \right] = 0, \forall \varepsilon > 0 \\ &\Leftrightarrow \sup_{k \geq 1} |X_{n+k} - X_n| \xrightarrow{p} 0 \\ &\Leftrightarrow \sup_{k \geq 1} |X_{n+k} - X_n| \xrightarrow{a.s.} 0. \end{aligned}$$

(c) *There is a r.v. $X \in L_r$ such that $X_n \xrightarrow{r} X$*

$$\begin{aligned} &\Leftrightarrow \lim_{m, n \rightarrow \infty} E[|X_m - X_n|^r] = 0 \\ &\Leftrightarrow \sup_{k \geq 1} E|X_{n+k} - X_n|^r \xrightarrow{n \rightarrow \infty} 0. \end{aligned}$$

2.18 Proposition *Let $\{\varepsilon_n\}_{n=1}^\infty$ be a sequence of positive constants such that $\lim_{n \rightarrow \infty} \varepsilon_n = 0$.*

Then

$$\sum_{n=1}^{\infty} P[|X_n - X| > \varepsilon_n] < \infty \Leftrightarrow X_n \xrightarrow{a.s.} X.$$

2.19 Proposition $X_n \xrightarrow{a.s.} X \Leftrightarrow \exists$ *a sequence ε_n such that $\lim_{n \rightarrow \infty} \varepsilon_n = 0$ and*

$$\lim_{n \rightarrow \infty} P[|X_k - X| \geq \varepsilon_k, \text{ for at least one } k \geq n] = 0.$$

3. Convergence of expectations and functions of random variables

3.1 Assumption *Let $\{X_n : n = 1, 2, \dots\}$ be a sequence of real r.v.'s, X a real r.v. and $g : \mathbb{R} \rightarrow \mathbb{R}$ a function such that $g(X)$ and $g(X_n)$, $n = 1, 2, \dots$, are real r.v.'s.*

Unless stated otherwise, this assumption will hold for all the definitions, propositions and theorems in this section.

3.2 Theorem HELLY-BRAY. $X_n \xrightarrow{L} X \Rightarrow E[g(X_n)] \rightarrow E[g(X)]$ *for any bounded continuous function $g : \mathbb{R} \rightarrow \mathbb{R}$.*

3.3 Definition UNIFORMLY INTEGRABLE FUNCTION. $|g|$ is uniformly integrable with respect to the distributions F_n iff

$\forall \varepsilon > 0$, $\exists a(\varepsilon)$ and $b(\varepsilon)$ which do not depend on n such that

$a \leq a(\varepsilon)$ and $b \geq b(\varepsilon)$ implies

$$\int |g(x)| dF_n(x) - \int_a^b |g(x)| dF_n(x) < \varepsilon, \forall n.$$

3.4 Proposition If $g(x)$ is a continuous function and $X_n \xrightarrow{L} X$, then

(a) $\liminf_{n \rightarrow \infty} E[|g(X_n)|] \geq E[|g(X)|]$;

(b) $|g(x)|$ is uniformly integrable with respect to $F_n \Rightarrow E[g(X_n)] \rightarrow E[g(X)]$;

(c) $E[|g(X_n)|] \rightarrow E[|g(X)|] < \infty \Leftrightarrow |g(x)|$ is uniformly integrable with respect to F_n .

3.5 Proposition If $E[|X_n|^{r_0}] \leq c < \infty$, $\forall n$, for $r_0 > 0$, then

(a) $X_n \xrightarrow{L} X \Rightarrow E[|X_n|^r] \rightarrow E[|X|^r] < \infty$ for all $0 < r < r_0$,

and

(b) $E[X_n^k] \rightarrow E[X^k] \neq \pm\infty$ for any integer $0 < k < r_0$, $k \in \mathbb{N}$.

3.6 Proposition If $E(X_n^k)$ exists and is finite for all k ($k = 1, 2, \dots$) and n , then

(a) $X_n \xrightarrow{L} X$ and $E(X_n^k) \rightarrow \mu_k \neq \pm\infty \Rightarrow E(X^k) = \mu_k$;

(b) $E(X_n^k) \rightarrow \mu_k \neq \pm\infty$, where the sequence $\{\mu_k : k = 1, 2, \dots\}$ defines a unique distribution function $F(x) \Rightarrow X_n \xrightarrow{L} X$, where X is a r.v. whose distribution function is $F(x)$.

3.7 Proposition *If $E|X_n|^r < \infty, \forall n$, for $r > 0$, then*

$$\begin{aligned} X_n \xrightarrow{r} X &\Rightarrow E|X_n|^r \rightarrow E|X|^r \text{ and } E|X|^r < \infty \\ &\Rightarrow E(X_n^k) \rightarrow E(X^k) \text{ for } 0 < k < r. \end{aligned}$$

3.8 Proposition *If $E(X_n^2) < \infty, \forall n$, then $X_n \xrightarrow{2} X$ implies*

$$E(X_n^2) \rightarrow E(X^2) < \infty \text{ and } E(X_n) \rightarrow E(X) \neq \pm\infty.$$

3.9 Proposition *If $E|X_n|^r < \infty, \forall n$, and $E|X|^r < \infty$, for $r > 0$, then*

$$X_n \xrightarrow{r} X \Rightarrow E(X_n^k) \rightarrow E(X^k), \text{ for any integer } 0 < k \leq r.$$

3.10 Proposition *Let $g : \mathbb{R} \rightarrow \mathbb{R}$ a continuous function everywhere on \mathbb{R} , except possibly in a set $A \subseteq \mathbb{R}$, and let X a r.v. such that $P[X \in A] = 0$. Then*

- (a) $X_n \xrightarrow{p} X \Rightarrow g(X_n) \xrightarrow{p} g(X)$;
- (b) $X_n \xrightarrow{a.s.} X \Rightarrow g(X_n) \xrightarrow{a.s.} g(X)$;
- (c) $X_n \xrightarrow{L} X \Rightarrow g(X_n) \xrightarrow{L} g(X)$.

3.11 Proposition *Let $\{X_n\}$ and $\{Y_n\}$ two sequences of random variables. Then*

- (a) $X_n \xrightarrow{p} X$ and $Y_n \xrightarrow{p} Y \Rightarrow X_n + Y_n \xrightarrow{p} X + Y$;
- (b) $X_n \xrightarrow{a.s.} X$ and $Y_n \xrightarrow{a.s.} Y \Rightarrow X_n + Y_n \xrightarrow{a.s.} X + Y$;
- (c) $X_n \xrightarrow{p} X$ and $Y_n \xrightarrow{p} Y \Rightarrow g(X_n, Y_n) \xrightarrow{p} g(X, Y)$ for any continuous function $g(x, y)$.

3.12 Proposition *Let $\{X_n\}$ and $\{Y_n\}$ two sequences of r.v.'s such that $X_n \xrightarrow{L} X$ and $Y_n \xrightarrow{p} c$, where X is a r.v. and c is a real constant ($-\infty < c < +\infty$). Then*

- (a) $X_n + Y_n \xrightarrow{L} X + c$;

- (b) $X_n Y_n \xrightarrow{L} Xc$;
- (c) $X_n/Y_n \xrightarrow{L} X/c$ if $c \neq 0$;
- (d) $(X_n, Y_n) \xrightarrow{L} (X, c)$.

3.13 Proposition Let $\{X_n\}$ and $\{Y_n\}$ two sequences of r.v.'s such that $X_n - Y_n \xrightarrow{p} 0$ and $Y_n \xrightarrow{L} Y$, and let $g : \mathbb{R} \rightarrow \mathbb{R}$ be a continuous function. Then

- (a) $X_n \xrightarrow{L} Y$;
- (b) $g(X_n) - g(Y_n) \xrightarrow{p} 0$;
- (c) $g(X_n) \xrightarrow{L} g(Y)$.

3.14 Proposition $X_n \xrightarrow{2} X \Leftrightarrow \lim_{\substack{m \rightarrow \infty \\ n \rightarrow \infty}} E(X_m X_n)$ exists and is finite irrespective of the way m and $n \rightarrow \infty$.

3.15 Proposition $X_n \xrightarrow{2} X$ and $Y_n \xrightarrow{2} Y \Rightarrow E(X_n + Y_n) \rightarrow E(X + Y)$ and $E(X_n Y_n) \rightarrow E(XY)$, where $E(X + Y)$ and $E(XY)$ are finite real numbers.

4. Random series

4.1. Definitions

4.1.1 Definition Let $\{X_t : t \in \mathbb{N}\}$ be a real-valued stochastic process and consider the series $\sum_{t=1}^{\infty} X_t$.

4.1.2 Definition We say $\sum_{t=1}^{\infty} X_t$ converges (according to given mode of convergence) iff there exists a real r.v. Y such that

$$\sum_{t=1}^N X_t \xrightarrow[N \rightarrow \infty]{} Y \text{ (according to the same mode of convergence).}$$

4.1.3 Remark Mode of convergence : a.s., in probability or in mean of order r .

4.1.4 QUESTIONS :

(1) Under which conditions does the series $\sum_{t=1}^{\infty} X_t$ converge?

(2) Under which conditions can one write

$$E \left(\sum_{t=1}^{\infty} X_t \right) = \sum_{t=1}^{\infty} E(X_t) ?$$

4.1.5 Definition We say $\sum_{t=1}^{\infty} X_t$ converges absolutely (according to a given convergence mode) iff $\sum_{t=1}^{\infty} |X_t|$ converges (according to the same mode of convergence). If $\sum_{t=1}^{\infty} |X_t|$ converges with probability one (a.s.), we write $\sum_{t=1}^{\infty} |X_t| < \infty$ a.s.

4.2. General convergence criteria

4.2.1 Proposition $\sum_{t=1}^{\infty} |X_t| < \infty$ a.s. $\Rightarrow \sum_{t=1}^{\infty} X_t$ converges a.s.

4.2.2 Proposition $\sum_{t=1}^{\infty} X_t$ converges absolutely in probability $\Leftrightarrow \sum_{t=1}^{\infty} X_t$ converges absolutely a.s.

4.2.3 Proposition If there is a sequence of positive constants $\{\varepsilon_t\}_{t=1}^{\infty}$ such that

- (a) $\sum_{t=1}^{\infty} \varepsilon_t < \infty$,
- (b) $\sum_{t=1}^{\infty} P(|X_t| \geq \varepsilon_t) < \infty$,

then $\sum_{t=1}^{\infty} X_t$ converges a.s. (i.e. there exists a r.v. X such that $\sum_{t=1}^N X_t \xrightarrow[N \rightarrow \infty]{a.s.} X$).

4.2.4 Proposition

$$\sum_{t=1}^{\infty} |X_t| < \infty \text{ a.s.} \Leftrightarrow P \left(\sum_{t=1}^n |X_t| < x \right) \xrightarrow[n \rightarrow \infty]{} F(x), \forall x, \quad (4.1)$$

where $F(x)$ is the distribution function of a r.v.

4.2.5 Proposition $\sum_{t=1}^{\infty} E|X_t| < \infty \Rightarrow \sum_{t=1}^{\infty} |X_t| < \infty$ a.s. and $E \left[\sum_{t=1}^{\infty} |X_t| \right] = \sum_{t=1}^{\infty} E|X_t|$.

4.2.6 Proposition $\sum_{t=1}^{\infty} (E|X_t|^r)^{1/r} < \infty$, where $r \geq 1 \Rightarrow \sum_{t=1}^{\infty} |X_t|$ and $\sum_{t=1}^{\infty} X_t$ converge a.s. and in mean of order r .

4.2.7 Proposition If $X_t \in L_2$, $\mu_t \equiv E(X_t)$ and $\sigma_t^2 \equiv \text{Var}(X_t)$, $\forall t$, then $\sum_{t=1}^{\infty} [\sigma_t^2 + \mu_t^2]^{\frac{1}{2}} < \infty$ implies

$$\sum_{t=1}^{\infty} |X_t| \text{ and } \sum_{t=1}^{\infty} X_t \text{ converge a.s. and q.m.}$$

In particular, if $E(X_t) = 0$, $\forall t$,

$$\sum_{t=1}^{\infty} \sigma_t < \infty \Rightarrow \sum_{t=1}^{\infty} |X_t| \text{ and } \sum_{t=1}^{\infty} X_t \text{ converge a.s. and q.m.}$$

4.2.8 Proposition

$$\sum_{t=1}^{\infty} E|X_t| < \infty \Rightarrow E \left[\sum_{t=1}^{\infty} X_t \right] = \sum_{t=1}^{\infty} E(X_t).$$

4.2.9 Proposition If $X_t \in L_2, \forall t$, and if there are non-negative constants $\{\bar{\rho}_t : t = 1, 2, \dots\}$ such that $0 \leq \bar{\rho}_t \leq 1$,

$$E(X_s X_t) \leq \bar{\rho}_{t-s} [E(X_s^2) E(X_t^2)]^{\frac{1}{2}}, \text{ for } 0 < s \leq t,$$

and

$$\sum_{t=1}^{\infty} \bar{\rho}_t < \infty,$$

then

$$\sum_{t=1}^{\infty} (\log t)^2 E(X_t^2) < \infty \Rightarrow \sum_{t=1}^{\infty} X_t \text{ converges a.s.}$$

4.2.10 Remark When $E(X_t) = 0$, we have

$$\text{Corr}(X_s, X_t) \leq \bar{\rho}_{t-s}.$$

4.3. Series of orthogonal variables

4.3.1 Proposition Let $\{X_t\}_{t=1}^{\infty}$ a sequence of r.v.'s such that $X_t \in L_2, E(X_t) = 0$ and $E(X_s X_t) = 0$ for $s \neq t$. Then

- (a) $\sum_{t=1}^{\infty} X_t$ converges in quadratic mean $\Leftrightarrow \sum_{t=1}^{\infty} E|X_t|^2 < \infty$;
- (b) $\sum_{t=1}^{\infty} E|X_t|^2 < \infty \Rightarrow E|X|^2 = \sum_{t=1}^{\infty} E|X_t|^2 < \infty$, where $\sum_{t=1}^N X_t \xrightarrow[N \rightarrow \infty]{2} X$;
- (c) $\sum_{t=1}^{\infty} \frac{E|X_t|^2}{b_t^2} < \infty$ and $b_n \uparrow \infty \Rightarrow \frac{1}{b_n} \sum_{t=1}^n X_t \xrightarrow{2} 0$; (Kolmogorov)
- (d) $\sum_{t=1}^{\infty} (\log t)^2 E|X_t|^2 < \infty \Rightarrow \sum_{t=1}^{\infty} X_t$ converges a.s. and in quadratic mean;
- (e) $\sum_{t=1}^{\infty} \left(\frac{\log t}{b_t}\right)^2 E|X_t|^2 < \infty$ and $b_n \uparrow \infty \Rightarrow \frac{1}{b_n} \sum_{t=1}^n X_t \xrightarrow{a.s.} 0$.

4.3.2 Proposition Let $\{X_t\}_{t=1}^{\infty}$ a sequence of r.v.'s such that $X_t \in L_2$ and $\text{Cov}(X_s, X_t) = 0$ for $s \neq t$, and let $\mu_t = E(X_t)$. Then

- (a) $\sum_{t=1}^{\infty} (X_t - \mu_t)$ converges in quadratic mean to a r.v. $Y \Leftrightarrow \sum_{t=1}^{\infty} \text{Var}(X_t) < \infty$;
- (b) $\sum_{t=1}^{\infty} \text{Var}(X_t) < \infty \Rightarrow \text{Var}(Y) = \sum_{t=1}^{\infty} \text{Var}(X_t)$, where $\sum_{t=1}^N (X_t - \mu_t) \xrightarrow[N \rightarrow \infty]{2} Y$;
- (c) $\sum_{t=1}^{\infty} \frac{\text{Var}(X_t)}{b_t^2} < \infty$ and $b_n \uparrow \infty \Rightarrow \frac{1}{b_n} \sum_{t=1}^n (X_t - \mu_t) \xrightarrow{2} 0$;
- (d) $\sum_{t=1}^{\infty} (\log t)^2 \text{Var}(X_t) < \infty \Rightarrow \sum_{t=1}^{\infty} (X_t - \mu_t)$ converges in quadratic mean and a.s. ;
- (e) $\sum_{t=1}^{\infty} \left(\frac{\log t}{b_t} \right)^2 \text{Var}(X_t) < \infty$ and $b_n \uparrow \infty \Rightarrow \frac{1}{b_n} \sum_{t=1}^n (X_t - \mu_t) \xrightarrow{a.s.} 0$.

4.3.3 Proposition Let $\{X_t\}_{t=1}^{\infty}$ a sequence of r.v.'s such that $X_t \in L_2$ and $E(X_s X_t) = 0$ for $s \neq t$. Then

- (a) $\sum_{t=1}^{\infty} E(X_t^2) < \infty \Rightarrow \sum_{t=1}^{\infty} X_t$ converges in q.m.;
- (b) $\sum_{t=1}^{\infty} (\log t)^2 E(X_t^2) < \infty \Rightarrow \sum_{t=1}^{\infty} X_t$ converges a.s.

4.4. Series of independent variables

4.4.1 Proposition Let $\{X_t\}_{t=1}^{\infty}$ a sequence of independent r.v.'s such that $X_t \in L_2$. Then

- (a) $\sum_{t=1}^{\infty} \text{Var}(X_t) < \infty \Rightarrow \sum_{t=1}^{\infty} (X_t - EX_t)$ converges a.s. ;
- (b) if $|X_t| \leq c < \infty, \forall t$,

$$\sum_{t=1}^{\infty} (X_t - EX_t) \text{ converges a.s.} \Leftrightarrow \sum_{t=1}^{\infty} \text{Var}(X_t) < \infty ;$$

- (c) $\sum_{t=1}^{\infty} EX_t$ converges to a finite number and $\sum_{t=1}^{\infty} \text{Var}(X_t) < \infty \Rightarrow \sum_{t=1}^{\infty} X_t$ converges a.s. and in quadratic mean ;
- (d) $|X_t| \leq c < \infty, \forall t$, and $\sum_{t=1}^{\infty} X_t$ converges a.s. $\Rightarrow \sum_{t=1}^{\infty} X_t$ and $\sum_{t=1}^{\infty} \text{Var}(X_t)$ converge.

4.4.2 Proposition THREE SERIES THEOREM (KOLMOGOROV). Let $\{X_t\}_{t=1}^{\infty}$ a sequence of independent r.v.'s and let

$$\begin{aligned} X_t^c &= X_t, \text{ if } |X_t| < c \\ &= 0, \text{ if } |X_t| \geq c \end{aligned}$$

where $c > 0$. The series $\sum_{t=1}^{\infty} X_t$ converges a.s. \Leftrightarrow there exists a constant $c > 0$ such that the three following series converge in \mathbb{R} :

- (a) $\sum_{t=1}^{\infty} P[|X_t| \geq c]$;
- (b) $\sum_{t=1}^{\infty} Var(X_t^c)$;
- (c) $\sum_{t=1}^{\infty} E(X_t^c)$.

4.4.3 Proposition Let $\{X_t\}_{t=1}^{\infty}$ a sequence of independent r.v.'s. If the two series $\sum_{t=1}^{\infty} E(X_t)$ and $\sum_{t=1}^{\infty} E(|X_t|^p)$ where $1 < p \leq 2$ converge, then the series $\sum_{t=1}^{\infty} X_t$ converges a.s.

4.4.4 Proposition Let $\{X_t\}_{t=1}^{\infty}$ a sequence of independent r.v.'s. Then

- (a) $\sum_{t=1}^{\infty} X_t$ converges a.s. $\Leftrightarrow \sum_{t=1}^{\infty} X_t$ converges in probability $\Leftrightarrow \sum_{t=1}^{\infty} X_t$ converges in law;
- (b) if $|X_t| \leq c < \infty$ and $E(X_t) = 0, \forall t$, then $\sum_{t=1}^{\infty} X_t$ converges a.s. iff

$$\sum_{t=1}^{\infty} X_t \text{ converges in quadratic mean.}$$

5. Laws of large numbers

5.1 Proposition Let $\{X_t\}_{t=1}^{\infty}$ a sequence of r.v.'s such that $X_t \in L_2$ and $Cov(X_s, X_t) = 0$ for $s \neq t$, and let $\mu_t = E(X_t)$. Then

$$(a) \sum_{n=1}^{\infty} \frac{1}{n^2} \text{Var}(X_n) < \infty \Rightarrow \bar{X}_n - \bar{\mu}_n \xrightarrow[n \rightarrow \infty]{2} 0 \quad (\text{Chebychev law})$$

where $\bar{X}_n = \sum_{t=1}^n X_t/n$ and $\bar{\mu}_n = \sum_{t=1}^n \mu_t/n$, and

$$(b) \sum_{n=1}^{\infty} \left(\frac{\log n}{n}\right)^2 \text{Var}(X_n) < \infty \Rightarrow \bar{X}_n - \bar{\mu}_n \xrightarrow[n \rightarrow \infty]{a.s.} 0.$$

In particular, if $\text{Var}(X_t) = \sigma^2 < \infty$ and $E(X_t) = \mu$ for all t , then

$$\frac{1}{n} \sum_{t=1}^n X_t \xrightarrow[n \rightarrow \infty]{a.s.} \mu \text{ and } \frac{1}{n} \sum_{t=1}^n X_t \xrightarrow[n \rightarrow \infty]{2} \mu.$$

5.2 Theorem KHINTCHINE WEAK LAW OF LARGE NUMBERS. Let $\{X_t\}_{t=1}^{\infty}$ a sequence of independent and identically distributed r.v.'s whose mean $E(X_t)$ exists. Then

$$E(X_t) = \mu \Rightarrow \bar{X}_n \xrightarrow[n \rightarrow \infty]{p} \mu.$$

5.3 Theorem FIRST KOLMOGOROV'S STRONG LAW OF LARGE NUMBERS. Let $\{X_t\}_{t=1}^{\infty}$ a sequence of independent r.v.'s such that $E(X_t) = \mu_t$ and $\text{Var}(X_t) = \sigma_t^2$ exist for all t . Then

$$\sum_{n=1}^{\infty} (\sigma_n/n)^2 < \infty \Rightarrow \bar{X}_n - \bar{\mu}_n \xrightarrow[n \rightarrow \infty]{a.s.} 0.$$

5.4 Theorem SECOND KOLMOGOROV'S STRONG LAW OF LARGE NUMBERS. Let $\{X_t\}_{t=1}^{\infty}$ a sequence of independent and identically distributed r.v.'s. Then

$$E(X_t) \text{ exists and is equal to } \mu \Leftrightarrow \bar{X}_n - \bar{\mu}_n \xrightarrow[n \rightarrow \infty]{a.s.} 0.$$

5.5 Theorem LAW OF LARGE NUMBERS FOR CORRELATED R.V.'S. Let $\{X_t\}_{t=1}^{\infty}$ a sequence of r.v.'s with means $E(X_t) = \mu_t$ and satisfying the following condition: there exists a sequence of real numbers $\{r_j\}_{j=0}^{\infty}$ such that

$$(a) 0 \leq r_j \leq 1, \text{ for all } j,$$

$$(b) \text{Cov}(X_s, X_t) \leq r_{t-s} [\text{Var}(X_s) \text{Var}(X_t)]^{1/2}, \text{ for } t \geq s,$$

$$(c) \sum_{j=1}^{\infty} r_j < \infty,$$

$$(d) \sum_{n=1}^{\infty} \left(\frac{\log(n)}{n} \right)^2 \text{Var}(X_n) < \infty.$$

Then

$$\bar{X}_n - \bar{\mu}_n \xrightarrow[n \rightarrow \infty]{a.s.} 0.$$

6. Central limit theorems

6.1 Theorem LINDBERG-LÉVY CENTRAL LIMIT THEOREM. *Let $\{X_t\}_{t=1}^{\infty}$ a sequence of independent and identically distributed r.v.'s in L_2 such that $E(X_t) = \mu$ and $\text{Var}(X_t) = \sigma^2 > 0$. Then*

$$\frac{1}{\sqrt{n}} \sum_{t=1}^n (X_t - \mu) / \sigma = \sqrt{n}(\bar{X}_t - \mu) / \sigma \xrightarrow[n \rightarrow \infty]{L} Z$$

where $Z \sim N(0, 1)$.

6.2 Theorem LIAPUNOV CENTRAL LIMIT THEOREM. *Let $\{X_t\}_{t=1}^{\infty}$ a sequence of independent r.v.'s in L_3 such that $E(X_t) = \mu_t$, $\text{Var}(X_t) = \sigma_t^2 \neq 0$, $E[|X_t - \mu_t|^3] = \beta_t$ for all t . Moreover,*

$$B_n = \left(\sum_{t=1}^n \beta_t \right)^{1/3}, C_n = \left(\sum_{t=1}^n \sigma_t^2 \right)^{1/2}.$$

If $\lim_{n \rightarrow \infty} (B_n / C_n) = 0$, then

$$\sum_{t=1}^n (X_t - \mu_t) / C_n = \sqrt{n}(\bar{X}_t - \bar{\mu}_n) / (C_n / \sqrt{n}) \sigma \xrightarrow[n \rightarrow \infty]{L} Z$$

where $Z \sim N(0, 1)$.

6.3 Theorem LINDBERG-FELLER CENTRAL LIMIT THEOREM. *Let $\{X_t\}_{t=1}^{\infty}$ a sequence of independent r.v.'s in L_2 such that*

$$P[X_t \leq x] = G_t(x), E(X_t) = \mu_t, \text{Var}(X_t) = \sigma_t^2 \neq 0,$$

for all t . Then

$$\sum_{t=1}^n (X_t - \mu_t) / C_n \xrightarrow[n \rightarrow \infty]{L} Z \text{ and } \lim_{n \rightarrow \infty} \max_{1 \leq t \leq n} (\sigma_t / C_n) = 0$$

iff

$$\lim_{n \rightarrow \infty} \frac{1}{C_n^2} \sum_{t=1}^n \int_{|x - \mu_t| > \varepsilon C_n} (x - \mu_t)^2 dG_t(x) = 0, \forall \varepsilon > 0.$$

7. Extension to random vectors

7.1 Definition STOCHASTIC CONVERGENCE FOR VECTORS. Let $\{X_n\}_{n=1}^{\infty}$ a sequence of vectors of dimension k ,

$$X_n = (X_{1n}, X_{2n}, \dots, X_{kn})', n = 1, 2, \dots$$

whose components are real random variables all defined on the same probability space (Ω, Q, P) , and

$$X = (X_1, X_2, \dots, X_k)'$$

another random vector of dimension k whose components are defined on the same space.

- (a) We say X_n converges to X in probability (almost surely, in mean of order r) as $n \rightarrow \infty$ if each component of X_n converges to the corresponding component of X in probability (almost surely, in mean of order r) as $n \rightarrow \infty$. Depending on the case considered, we then write $X_n \xrightarrow{P} X$, $X_n \xrightarrow{a.s.} X$ or $X_n \xrightarrow{r} X$.
- (b) We say X_n converges in law to X ($X_n \xrightarrow{L} X$) iff

$$\lim_{n \rightarrow \infty} F_{X_n}(x) = F_X(x) \text{ at all continuity points of } F_X(x).$$

where $x = (x_1, x_2, \dots, x_k)' \in \mathbb{R}^k$,

$$F_{X_n}(x) = P[X_{1n} \leq x_1, \dots, X_{kn} \leq x_k], n = 1, 2, \dots$$

and

$$F_X(x) = P[X_1 \leq x_1, \dots, X_k \leq x_k].$$

7.2 Theorem UNIVARIATE CHARACTERIZATION OF CONVERGENCE IN LAW FOR A SEQUENCE OF VECTORS.. Let $\{X_n\}_{n=1}^{\infty}$ be a sequence of random vectors of dimension $k \times 1$ and let X be another random vector of dimension $k \times 1$. Then

$$X_n \xrightarrow[n \rightarrow \infty]{L} X \Leftrightarrow \lambda' X_n \xrightarrow[n \rightarrow \infty]{L} \lambda' X, \forall \lambda \in \mathbb{R}^k.$$

In particular, if $X \sim N[\mu, \Sigma]$,

$$X_n \xrightarrow[n \rightarrow \infty]{L} N[\mu, \Sigma] \Leftrightarrow \lambda' X_n \xrightarrow[n \rightarrow \infty]{L} N[\lambda' \mu, \lambda' \Sigma \lambda], \forall \lambda \in \mathbb{R}^k.$$

8. Proofs and additional references

Proofs and further discussions of the results presented above may be found the following references.

- 1.2 Lukacs (1975), Theorem 2.2.6, p. 39.
- 2.6 Loève (1977), p. 153.
- 2.7 Lukacs (1975), p. 49.
- 2.8 Lukacs (1975), pp. 37 and 49.
- 2.11 Loève (1977), vol. I, p. 165.
- 2.12 Loève (1977), vol. I, p. 166.
- 2.13 Loève (1977), vol. I, p. 166.
- 2.14 Lukacs (1975), p. 38.
- 2.15 Loève (1977), vol. I, p. 240.
- 2.17 (a) Loève (1977), vol. I, pp. 114, 118, 153, and Lukacs (1975), p. 48.
(b) Loève (1977), vol. I, p. 153, and Lukacs (1975), p. 45.
(c) Loève (1977), p. 163, and Lukacs (1975), p. 49.
- 2.18 Stout (1974), Theorem 2.1.2, p. 11.
- 2.19 Loève (1977), vol. I, p. 175.
- 3.2 Loève (1977), vol. I, p. 184.
- 3.4 Loève (1977), vol. I, p. 185.
- 3.5 Loève (1977), vol. I, p. 186.
- 3.6 Rao (1973), p. 121.
- 3.7 Loève (1977), vol. I, p. 165, and Prakasa Rao (1987), p. 10.
- 3.10 Prakasa Rao (1987), p. 12.
- 3.11 Prakasa Rao (1987), pp. 43 and 57-58.
- 3.12 Rao (1973), p. 122) and Prakasa Rao (1987), p. 10.
- 3.13 Rao (1973), p. 124.
- 3.14 Lukacs (1975), p. 50.
- 4.2.2 Implication of Proposition 2.8.
- 4.2.3 Lukacs (1975), Corollary 4.2.2, p. 82.
- 4.2.4 Loève (1977), vol. I, p. 175, Problem 10.
- 4.2.5 Loève (1977), vol. I, p. 175, Problem 10, Lukacs (1975), Theorem 4.2.1, p. 80, and the Monotone convergence theorem [Loève (1977), vol. I, p. 125].
- 4.2.6 Loève (1977), vol. I, p. 175, Problem 10, and Proposition 2.13 above.
- 4.2.8 Loève (1977), p. 111. This result is an implication of the Dominated convergence theorem: Loève (1977), vol. I, p. 126, and Royden (1968), pp. 88-89.
- 4.2.9 Stout (1974), Corollary 2.4.1, p. 28.
- 4.3.1 Loève (1977), vol. II, Theorem 30.1, p. 122, and Theorem 30.1B, p. 124.
- 4.3.3 (a) Loève (1977), Theorem 4.2.4A, p. 85, and Stout (1974), Lemma 2.2.2, p. 15.

- (b) Stout (1974), Theorem 2.3.2, p. 20.
- 4.4.1** (a) Lukacs (1975), Theorem 4.2.4, p. 84, and Loève (1977), vol. I, Theorem 17.3.Ia, p. 248.
- (b) Loève (1977), vol. I, Theorem 17.3.Ia, p. 248, and Lukacs (1975), Theorem 4.2.4, Corollary 3, p. 87.
- (c) Lukacs (1975), Theorem 4.2.4, Corollary 1, p. 84.
- (d) Loève (1977), Theorem 17.3.Ib, p. 248.
- 4.4.3** Lukacs (1975), Theorem 4.2.6, Corollary, p. 89.
- 4.4.4** (a) Lukacs (1975), Theorem 4.2.8, p. 92.
- 5.1** This result can be deduced on considering the case where $b_t = t$ in **4.3.2** (c) and (d).
- 5.2** Rao (1973), Section 2c.3, p. 112.
- 5.3** Rao (1973), Section 2c.3, p. 114.
- 5.4** Rao (1973), Section 2c.3, p. 115.
- 5.5** Gouriéroux and Monfort (1995), vol. 2, Appendix.
- 6.1** Rao (1973), Section 2c.5, p. 127.
- 6.2** Rao (1973), Section 2c.5, p. 127.
- 6.3** Rao (1973), Section 2c.5, p. 128.
- 7.2** Rao (1973), Section 2c.5, p. 128.

Several counterexamples relevant to the above results are presented in Stoyanov (1997). About laws of large of numbers, the literature on ergodic theorems is also relevant; for a review, see Petersen (1983). For general presentations of asymptotic theory in view of statistical applications, the reader may consult: Rao (1973, Chapter 2), Serfling (1980), White (1984), Prakasa Rao (1987), McCabe and Tremayne (1993), Davidson (1994), van der Vaart (1998), Spanos (1999, Chapter 9) and Lehmann (1999).

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