

# Notions of asymptotic theory \*

Jean-Marie Dufour †

First version: December 1995

Revised: January 2002

This version: January 21, 2003

Compiled: January 21, 2003, 11:13pm

---

\* This work was supported by the Canada Research Chair Program (Chair in Econometrics, Université de Montréal), the Alexander-von-Humboldt Foundation (Germany), the Canadian Network of Centres of Excellence [program on *Mathematics of Information Technology and Complex Systems* (MITACS)], the Canada Council for the Arts (Killam Fellowship), the Natural Sciences and Engineering Research Council of Canada, the Social Sciences and Humanities Research Council of Canada, and the Fonds FCAR (Government of Québec).

† Canada Research Chair Holder (Econometrics). Centre de recherche et développement en économie (C.R.D.E.), Centre interuniversitaire de recherche en analyse des organisations (CIRANO), and Département de sciences économiques, Université de Montréal. Mailing address: Département de sciences économiques, Université de Montréal, C.P. 6128 succursale Centre-ville, Montréal, Québec, Canada H3C 3J7. TEL: 1 514 343 2400; FAX: 1 514 343 5831; e-mail: jean.marie.dufour@umontreal.ca. Web page: <http://www.fas.umontreal.ca/SCECO/Dufour>.

# Contents

<b>List of Definitions, Propositions and Theorems</b>	<b>ii</b>
<b>1. Stochastic convergence</b>	<b>1</b>
<b>2. Relations between convergence concepts</b>	<b>2</b>
<b>3. Convergence of expectations and functions of random variables</b>	<b>5</b>
<b>4. Random series</b>	<b>8</b>
4.1. Definitions . . . . .	8
4.2. General convergence criteria . . . . .	9
4.3. Series of orthogonal variables . . . . .	11
4.4. Series of independent variables . . . . .	12
<b>5. Laws of large numbers</b>	<b>13</b>
<b>6. Central limit theorems</b>	<b>15</b>
<b>7. Extension to random vectors</b>	<b>16</b>
<b>8. Proofs and additional references</b>	<b>18</b>

## List of Definitions, Propositions and Theorems

1.2	<b>Proposition :</b> Unicity of probability limit . . . . .	1
2.2	<b>Proposition :</b> Relations between convergence concepts . . . . .	2
2.4	<b>Proposition :</b> Moment condition for convergence in mean of order $r$ . . . . .	2
2.5	<b>Proposition :</b> Sufficient condition for complete convergence . . . . .	2
2.9	<b>Definition :</b> Sequence of uniformly continuous integrals . . . . .	3
2.10	<b>Definition :</b> Uniformly integrable sequence . . . . .	3
2.17	<b>Proposition :</b> Cauchy criterion . . . . .	4
3.2	<b>Theorem :</b> Helly-Bray . . . . .	5
3.3	<b>Definition :</b> Uniformly integrable function . . . . .	6
4.4.2	<b>Theorem :</b> Three series theorem (Kolmogorov) . . . . .	13
5.2	<b>Theorem :</b> Khintchine weak law of large numbers . . . . .	14
5.3	<b>Theorem :</b> First Kolmogorov's strong law of large numbers . . . . .	14
5.4	<b>Theorem :</b> Second Kolmogorov's strong law of large numbers . . . . .	14
5.5	<b>Theorem :</b> Law of large numbers for correlated r.v.'s . . . . .	14
6.1	<b>Theorem :</b> Lindeberg-Lévy central limit theorem . . . . .	15
6.2	<b>Theorem :</b> Liapunov central limit theorem . . . . .	15
6.3	<b>Theorem :</b> Lindeberg-Feller central limit theorem . . . . .	15
7.1	<b>Definition :</b> Stochastic convergence for vectors . . . . .	16
7.2	<b>Theorem :</b> Univariate characterization of convergence in law for a sequence of vectors. . . . .	17

# 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  $(\Omega, \mathcal{A}, P)$  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  $(\Omega, \mathcal{A}, P)$ , 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  $(\Omega, \mathcal{A}, P)$  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  $X_n \xrightarrow{a.s.} X$

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

(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  $\varepsilon_n$  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)  $\sum_{t=1}^{\infty} \text{Var}(X_t^c)$  ;
- (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  $\text{Cov}(X_s, X_t) = 0$  for  $s \neq t$ , and let  $\mu_t = E(X_t)$ . Then



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

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

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

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  $(\Omega, \mathcal{Q}, P)$ , 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).

## References

- DAVIDSON, J. (1994): *Stochastic Limit Theory: An Introduction for Econometricians*. Oxford University Press, Oxford, U.K., second edn.
- GOURIÉROUX, C., AND A. MONFORT (1995): *Statistics and Econometric Models, Volumes One and Two*. Cambridge University Press, Cambridge, U.K.
- LEHMANN, E. L. (1999): *Elements of Large-Sample Theory*. Springer-Verlag, New York.
- LOÈVE, M. (1977): *Probability Theory, Volumes I and II*. Springer-Verlag, New York, 4th edn.
- LUKACS, E. (1975): *Stochastic Convergence*. Academic Press, New York, second edn.
- MCCABE, B., AND A. TREMAYNE (1993): *Elements of Modern Asymptotic Theory with Statistical Applications*. Manchester University Press, Manchester.
- PETERSEN, K. E. (1983): *Ergodic Theory*. Cambridge University Press, Cambridge, U.K.
- PRAKASA RAO, B. L. S. (1987): *Asymptotic Theory of Statistical Inference*. John Wiley & Sons, New York.
- RAO, C. R. (1973): *Linear Statistical Inference and its Applications*. John Wiley & Sons, New York, second edn.
- ROYDEN, H. L. (1968): *Real Analysis*. MacMillan, New York.
- SERFLING, R. J. (1980): *Approximation Theorems of Mathematical Statistics*. John Wiley & Sons, New York.
- SPANOS, A. (1999): *Probability Theory and Statistical Inference: Econometric Modelling with Observational Data*. Cambridge University Press, Cambridge, UK.
- STOUT, W. F. (1974): *Almost Sure Convergence*. Academic Press, New York.
- STOYANOV, J. M. (1997): *Counterexamples in Probability*. John Wiley & Sons, Chichester, U.K., second edn.
- VAN DER VAART, A. W. (1998): *Asymptotic Statistics*. Cambridge University Press, Cambridge, U.K.
- WHITE, H. (1984): *Asymptotic Theory for Econometricians*. Academic Press, Orlando, Florida.