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Strong Law of large number Law of the iterated logarithm for nonlinear probabilities

ZENGJING CHEN

SHANDONG UNIVERSITY

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Outline

- ◇ ***History of LLN and LIL for probabilities***
- ◇ ***Why to study LLN and LIL for capacities***
- ◇ ***Nonlinear probabilities and nonlinear expectations***
- ◇ ***Main results***
- ◇ ***Applications***



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0.1. History of LLN and LIL for probability

★ Law of large number(LLN):

(1) Brahmagupta (598-668), Cardano (1501-1576)

(2) Jakob Bernoulli(1713), Poisson (1835)

(3) Chebyshev, Markov, Borel(1909), Cantelli and Kolmogorov(IID).

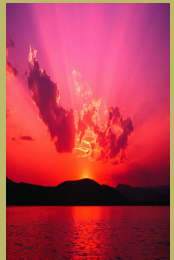
★ Law of iterated logarithm(LIL):

(1) Khintchine(1924) for Bernoulli model

Kolmogorov(1929), Hartman–Wintner(1941) (IID)

(2) Levy(1937) for Martingale

(3) Strassen(1964) for functional random variables.



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0.2. Strong LLN and LIL for probabilities

Assumption: $\{X_i\}$ IID , $S_n/n := \sum_{i=1}^n X_i$, $EX_1 = \mu$, Then

Theorem 1:Kolmogorov:

$$P\left(\lim_{n \rightarrow \infty} S_n/n = \mu\right) = 1$$

Theorem 2: Hartman–Wintner(1941): If $EX_1 = 0$, $EX_1^2 = \sigma^2$, Then

(a)

$$P\left(\limsup_{n \rightarrow \infty} \frac{S_n}{\sqrt{2n \log \log n}} = \sigma\right) = 1$$

(b)

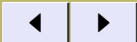
$$P\left(\liminf_{n \rightarrow \infty} \frac{S_n}{\sqrt{2n \log \log n}} = -\sigma\right) = 1$$

(c) Suppose that $C(\{x_n\})$ is the cluster set of a sequence of $\{x_n\}$ in R , then

$$P\left(C(\{\omega : S_n(\omega)/\sqrt{2n \log \log n}\}) = [-\sigma, \sigma]\right) = 1.$$

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0.3. Why to study LLN and LIL in Finance

THEOREM 1 (Black-Scholes, 1973:) *In complete markets, there exists a unique probability measure Q , such that the pricing of option ξ at strike date T is given by $E_Q[\xi e^{-rT}]$. Where $r = 0$ is interest rate of bond.*

Monte Carlo, $\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n X_i = E_Q[\xi]$.

- ★ (Linear) expectation \leftarrow **Black-Scholes** \rightarrow Complete Markets
- ★ $\inf_{Q \in \mathcal{P}} E_Q[\xi], \sup_{Q \in \mathcal{P}} E_Q[\xi] \iff$ Incomplete Markets, Q is not unique, SET \mathcal{P} .
- ★ **Super-pricing:** $\inf_{Q \in \mathcal{P}} E_Q[\xi], \sup_{Q \in \mathcal{P}} E_Q[\xi]$. Nonlinear expectation!
 $\lim_{n \rightarrow \infty} S_n/n = ?$



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0.4. Bernoulli Trials with ambiguity

★ Bernoulli Trials:

Repeated **independent** trials are called Bernoulli trials if there are only two possible outcomes for each trial and their probabilities **REMAIN** (are no longer) the same throughout the trials.

★ Let $X_i = 1$ if head occurs and $X_i = 0$ if tail occurs.

$$P_\theta(X_i = 1) = \theta, \quad P_\theta(X_i = 0) = 1 - \theta, \quad S_n := \sum_{i=1}^n X_i$$

★ If $\theta = 1/2$ (Unbalance), LLN stats

$$P_\theta(\lim_{n \rightarrow \infty} S_n/n = 1/2) = 1$$

Or

$$\lim_{n \rightarrow \infty} S_n/n = 1/2 \quad a.s \quad (P_\theta)$$



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★ If a coin is balance. $P_\theta(X_i = 1) = \theta \in [1/3, 1/2]$.

Let $\mathcal{P} := \{P_\theta, \theta \in [1/3, 1/2]\}$.

$E_{P_\theta}[X_i] = \theta$ Unknown,

But $\max_{P \in \mathcal{P}} E_P[X_i] = 1/2, \quad \min_{P \in \mathcal{P}} E_P[X_i] = 1/3.$

★ **Question:** what is the limit $S_n/n \rightarrow ?$

(a) Capacity: If $V(A) := \max_{P \in \mathcal{P}} P(A), \quad v(A) := \min_{P \in \mathcal{P}} P(A)$

Can S_n/n converge to $\max_{P \in \mathcal{P}} E_P[X_i]$ or $\min_{P \in \mathcal{P}} E_P[X_i]$ a.s. V or v ?

(b) The relation between the set of limit points of S_n/n and the interval of $\min_{P \in \mathcal{P}} E_P[X_i]$ and $\max_{P \in \mathcal{P}} E_P[X_i]$.



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0.5. Linear and Nonlinear Expectations

★ Kolmogorov: Linear expectation: $P : \mathcal{F} \rightarrow [0, 1], P(A) = E[I_A]$

$$P(A + B) = P(A) + P(B), \quad A \cap B = \emptyset \Leftrightarrow E[\xi + \eta] = E[\xi] + E[\eta]$$

Expectation is a linear functional of random variable.

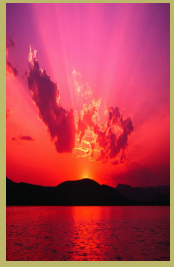
★ Nonlinear probability(capacity): $V(\cdot) : \mathcal{F} \rightarrow [0, 1]$ but

$$V(A + B) \neq V(A) + V(B), \text{ even } A \cap B = \emptyset.$$

★ Nonlinear expectation: $\mathbb{E}(\xi)$ is nonlinear functional in the sense of

$$\mathbb{E}[\xi + \eta] \neq \mathbb{E}[\xi] + \mathbb{E}[\eta].$$

Capacity $V(A) = \mathbb{E}[I_A]$ is nonlinear.



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Modes of nonlinear expectations and capacity

(1) Choquet expectations (Choquet 1953, physics)

$$C_V[X] := \int_0^\infty V(X \geq t)dt + \int_{-\infty}^0 [V(X \geq t) - 1]dt.$$

(2) g-expectation (Peng 1997)

(3) Sub-linear expectation (Peng 2007).

(a) Monotonicity: $X \geq Y$ implies $\mathbb{E}[X] \geq \mathbb{E}[Y]$.

(b) Constant preserving: $\mathbb{E}[c] = c, \forall c \in \mathbb{R}$.

(c) Sub-additivity: $\mathbb{E}[X + Y] \leq \mathbb{E}[X] + \mathbb{E}[Y]$.

(d) Positive homogeneity: $\mathbb{E}[\lambda X] = \lambda \mathbb{E}[X], \forall \lambda \geq 0$.

(1) Distorted probability measure: $V(A) = g(P(A)), g : [0, 1] \rightarrow [0, 1]$.

(2) 2-alternating capacity: $V(A \cup B) \leq V(A) + V(B) - V(A \cap B)$

(3) $V(A) = \max_{P \in \mathcal{P}} P(A), \mathcal{P}$ set of Probability.



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1. Independence w.r.t probability or capacity

★ **Linear:** A and B independent $P(AB) = P(A)P(B)$

$$\Leftrightarrow E[\phi(I_A + I_B)] = E[E[\phi(x + I_B)]|_{x=I_A}], \forall \phi(x)$$

★ **Nonlinear:** Epstein(2002), Marinacci(2005) $V(AB) = V(A)V(B)$

$$\Leftrightarrow \mathbb{E}[\phi(I_A + I_B)] = \mathbb{E}[\mathbb{E}[\phi(x + I_B)]|_{x=I_A}]$$



2. Definition of IID under expectation

DEFINITION 1 (Peng 2007)

Independence: A random variable $X \in \mathcal{H}$ is said to be independent under \mathbb{E} to Y , if for each φ such that $\varphi(X, Y) \in \mathcal{H}$ and $\varphi(X, y) \in \mathcal{H}$ for each $y \in \mathbb{R}$

$$\mathbb{E}[\varphi(X, Y)] = \mathbb{E}[\bar{\varphi}(Y)],$$

where $\bar{\varphi}(y) := \mathbb{E}[\varphi(X, y)]$.

Identical distribution: Random variables X and Y are said to be identically distributed, if for each φ such that $\varphi(X), \varphi(Y) \in \mathcal{H}$,

$$\mathbb{E}[\varphi(X)] = \mathbb{E}[\varphi(Y)].$$

Mutual independence: X and Y are mutually independent

$$\mathbb{E}[\phi(X + Y)] = \mathbb{E}[\mathbb{E}[\phi(X + y)]|_{y=Y}]$$

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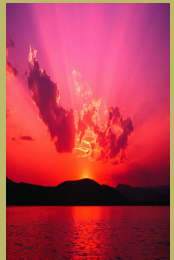
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3. Definition: capacity and nonlinear expectation

(1) **Probability space** : $(\Omega, \mathcal{F}, P) \Rightarrow (\Omega, \mathcal{F}, \mathcal{P})$. Where $\mathcal{P} := \{P_\theta : \theta \in \Theta\}$.

(2) **Capacity**: $P \Rightarrow (v, V)$, where

$$v(A) = \inf_{Q \in \mathcal{P}} Q(A), \quad V(A) = \sup_{Q \in \mathcal{P}} Q(A).$$

(3) **Property**:

$$V(A) + V(A^c) \geq 1, \quad v(A) + v(A^c) \leq 1$$

but

$$V(A) + v(A^c) = 1.$$

(4) **Nonlinear expectations**: Lower-upper expectation $\mathcal{E}[\xi]$ and $\mathbb{E}[\xi]$

$$\mathcal{E}[\xi] = \inf_{Q \in \mathcal{P}} E_Q[\xi], \quad \mathbb{E}[\xi] = \sup_{Q \in \mathcal{P}} E_Q[\xi]$$



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4. LLN for sub-linear expectations

★ Weak LLN:

THEOREM 2 (Peng 2007,2008) $\{X_i\}_{i=1}^{\infty}$ IID random variables, $\bar{\mu} := \mathbb{E}[X_1]$, $\underline{\mu} := \mathcal{E}[X_1]$. Then for any continuous and linear growth function ϕ ,

$$\mathbb{E} \left[\phi \left(\frac{1}{n} \sum_{i=1}^n X_i \right) \right] \rightarrow \sup_{\underline{\mu} \leq x \leq \bar{\mu}} \phi(x), \text{ as } n \rightarrow \infty.$$

★ Theorem (Peng, 2006,2007). CLT for IID

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★ $V(AB) = V(A)V(B), v(AB) = v(A)v(B)$

★ Theorem (Epstein, 02, Marinacci, 99, 05). ξ Bounded, Ω Polish, $C_v[X_i] = \underline{\mu}, C_V[X_i] = \bar{\mu}. \{X_i\}$ IID, then

$$v \left(\underline{\mu} \leq \liminf_{n \rightarrow \infty} S_n/n \leq \limsup_{n \rightarrow \infty} S_n/n \leq \bar{\mu} \right) = 1.$$

Where V is totally 2-alternating $V(A \cup B) \leq V(A) + V(B) - V(AB)$, here C_v and C_V is Choquet are integrals.

Note $C_v[X] \leq \mathcal{E}[X] \leq \mathbb{E}[X] \leq C_V[X], \forall X.$



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4.1. Limit theorem 1

Theorem: If $\{X_i\}$ is IID, then $\frac{S_n}{n}$ converges as $n \rightarrow \infty$ a.s. v if and only if

$$\mathcal{E}[X_1] = \mathbb{E}[X_1].$$

In this case,

$$\lim S_n/n = \mathcal{E}[X_1], \quad a.s. \quad v.$$



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5. Main results

THEOREM 3 $\{X_i\}_{i=1}^n$ IID under nonlinear expectation \mathbb{E} . Set $\bar{\mu} := \mathbb{E}[X_i]$, $\underline{\mu} := \mathcal{E}[X_i]$ and $S_n := \sum_{i=1}^n X_i$. If $\mathbb{E}[|X_i|^{1+\alpha}] < \infty$ for $\alpha > 0$. Then

(I)

$$V(\omega \in \Omega : \underline{\mu} \leq \liminf_{n \rightarrow \infty} S_n(\omega)/n \leq \limsup_{n \rightarrow \infty} S_n/n(\omega) \leq \bar{\mu}) = 1.$$

(II)

$$V(\omega \in \Omega : \limsup_{n \rightarrow \infty} S_n(\omega)/n = \bar{\mu}) = 1$$

$$V(\omega \in \Omega : \liminf_{n \rightarrow \infty} S_n(\omega)/n = \underline{\mu}) = 1.$$

(III) Suppose that $C(\{S_n(\omega)/n\})$ is the cluster set of a sequence of $\{S_n(\omega)/n\}$, then

$$V(\omega \in \Omega : C(\{S_n(\omega)/n\}) = [\underline{\mu}, \bar{\mu}]) = 1$$



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6. Law of iterated logarithm for sub-linear expectations

THEOREM 4 $\{X_n\}$ bounded IID. $\mathbb{E}[X_1] = \mathcal{E}[X_1] = 0$, $\bar{\sigma}^2 := \mathbb{E}[X_1^2]$, $\underline{\sigma}^2 := \mathcal{E}[X_1^2]$. Let $S_n := \sum_{i=1}^n X_i$, $a_n := \sqrt{2n \lg \lg n}$, then

(I)

$$v \left(\underline{\sigma} \leq \limsup_n \frac{S_n}{a_n} \leq \bar{\sigma} \right) = 1;$$

(II)

$$v \left(-\bar{\sigma} \leq \liminf_n \frac{S_n}{a_n} \leq -\underline{\sigma} \right) = 1.$$

(III) Suppose that $C(\{x_n\})$ is the cluster set of a sequence of $\{x_n\}$ in \mathbb{R} , then

$$v \left(C(\{S_n/\sqrt{2n \log \log n}\}) \supset (-\underline{\sigma}, \underline{\sigma}) \right) = 1.$$



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7. Key of proof

THEOREM 5 Suppose ξ is distributed to G normal $N(0; [\underline{\sigma}^2, \bar{\sigma}^2])$, where $0 < \underline{\sigma} \leq \bar{\sigma} < \infty$. Let ϕ be a bounded continuous function. Furthermore, if ϕ is a positively even function, then, for any $b \in R$,

$$e^{-\frac{b^2}{2\sigma^2}} \mathcal{E}[\phi(\xi)] \leq \mathcal{E}[\phi(\xi - b)].$$



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8. Application

Total 100 balls in box, Black + Red + Yellow = 100,

Black = Red, Yellow $\in [30, 40]$, then $P_Y \in [3/10, 4/10]$.

Take a ball from this box,

$X_i = 1$, if ball is black, $X_i = 0$, if ball is Yellow, $X_i = -1$ for red.

$S_n = \sum_{i=1}^n X_i$, is the excess frequency of black than Red

Then

(a) $\mathbb{E}[X_i] = \mathcal{E}[X_i] = 0$

(b)

$$\sqrt{6/10} \leq \limsup_{n \rightarrow \infty} \frac{S_n}{\sqrt{2n \lg \lg n}} \leq \sqrt{7/10}.$$



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9. Nonlinear expectation in Finance

In incomplete markets, there exists a set \mathcal{P} of probability measures, such that the super-sub-hedging price of option ξ at strike date T are given by

$$\underline{\mu} := \inf_{Q \in \mathcal{P}} E_Q[\xi], \bar{\mu} := \sup_{Q \in \mathcal{P}} E_Q[\xi].$$

then

(1)

$$\underline{\mu} \leq \liminf_{n \rightarrow \infty} S_n(\omega)/n \leq \limsup_{n \rightarrow \infty} S_n/n(\omega) \leq \bar{\mu}$$

(2)

$$\limsup_{n \rightarrow \infty} S_n(\omega)/n = \bar{\mu}, \quad V,$$

$$\liminf_{n \rightarrow \infty} S_n(\omega)/n = \underline{\mu}, \quad V$$



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