Janko Gravner

MAT 235B

February 13, 2024

Recall that a family $\{X_n : n\}$ of random variables is *uniformly integrable* if

$$\lim_{M\to\infty}\sup_n E[|X_n|1_{\{|X_n|\geq M\}}]=0$$

Here, the index set is arbitrary (but we assume natural numbers in convergence statements). The random variables do not need to be defined on the same probability space.

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Theorem

- (0) If $|X_n| \leq Y_n$, and Y_n are u.i., then X_n are u.i.
- (1) If X_n are u.i., then $\sup_n E|X_n| < \infty$.
- (2) If $\sup_n E|X_n|^{\alpha} < \infty$ for some $\alpha > 1$, then X_n are u.i.
- (3) If $X_n \stackrel{d}{=} X$ and $E|X| < \infty$, then X_n are u.i.
- (4) If X_n and Y_n are u.i., so is $X_n + Y_n$.
- (5) If X_n are u.i., $X_n \stackrel{d}{\longrightarrow} X$, then $EX_n \to EX$.
- (6) If $E|X_n| \to 0$, $E|X_n| < \infty$ for all n, then X_n are u.i.
- (7) If $X_n \to X$ a.s., $E|X_n| \to E|X|$, $E|X_n| < \infty$ for all n, and $E|X| < \infty$, then X_n are u.i.

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Proof.

(0)–(5) are routine, except perhaps (4):

$$|X_n + Y_n| \mathbf{1}_{\{|X_n + Y_n| \geq M\}} \leq 2|X_n| \mathbf{1}_{\{|X_n| \geq M/2\}} + 2|Y_n| \mathbf{1}_{\{|Y_n| \geq M/2\}}$$

For example, (5) is proved by assuming a.s. convergence. By (3) and (4), $Y_n = |X_n - X|$ are u.i. For every M, $E(Y_n 1_{\{Y_n \leq M\}}) \to 0$ by DCT, and then $\limsup EY_n \leq \sup_n E(Y_n 1_{\{Y_n \geq M\}}) \to 0$ as $M \to \infty$. So, $\limsup EY_n = 0$.

To prove (6), fix $\epsilon > 0$. Choose $N = N_{\epsilon}$ so that $E|X_n| < \epsilon$ for $n \ge N_{\epsilon}$. Then choose $M = M_{\epsilon}$ so that $E[|X_n| 1_{\{|X_n| \ge M\}}] \le \epsilon$ for $n = 1, \dots, N$.

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Proof, continued.

To prove (7), pick $\epsilon > 0$. Let $\varphi = \varphi_M$ be the function on [0, M] that linearly connects (0,0) and (M-1,M-1), and (M-1,M-1) and (M,0). Choose M so that $E\varphi(|X|) \geq E|X| - \epsilon/2$ (MCT). As $E\varphi(|X_n|) \rightarrow E\varphi(|X|)$ (DCT), $E\varphi(|X_n|) \geq E|X| - \epsilon$ for $n \geq N_\epsilon$. Then for $n \geq N_\epsilon$,

$$E[|X_n|1_{\{|X_n|\geq M\}}] = E|X_n| - E[|X_n|1_{\{|X_n|< M\}}]$$

$$\leq E|X_n| - E\varphi(|X_n|)$$

$$\leq E|X_n| - E|X| + \epsilon < 2\epsilon,$$

for $n \ge N_{\epsilon}$ (possibly enlarged). The remaining r.v.'s $X_1, \dots X_{N_{\epsilon}-1}$ are dealt with as before.

Uniform integrability of conditional expectation

Theorem

Assume that X is a r.v. on (Ω, \mathcal{F}, P) with $E|X| < \infty$. Then

$$\{E[X \mid \mathcal{G}] : \mathcal{G} \subset \mathcal{F} \text{ is a } \sigma\text{-algebra}\}$$

is a u.i. family.

Proof.

Step 1. If $P(A_n) \to 0$, then $E(|X|1_{A_n}) \to 0$. If not, there exists a sequence A_n with $E(|X|1_{A_n}) \ge \delta > 0$ and $P(A_n) \le 1/2^n$. But then by BC, $1_{A_n} \to 0$ a.s., then $|X|1_{A_n} \to 0$ a.s., and then $E[|X|1_{A_n}] \to 0$ by DCT, contradiction.

Step 2. $P(E[|X| \mid \mathcal{G}] \ge M) \le \frac{1}{M}E|X|$. This is Markov's inequality.

Uniform integrability of conditional expectation

Proof, continued.

Step 3.

$$E[|E[X \mid \mathcal{G}]|1_{\{|E[X|\mathcal{G}]| \geq M\}}]$$

$$\leq E[E[|X| \mid \mathcal{G}]1_{\{E[|X||\mathcal{G}] \geq M\}}]$$

$$= E[|X|1_{\{E[|X||\mathcal{G}] \geq M\}}]$$

Pick $\epsilon > 0$. Choose a $\delta > 0$ so that $P(A) < \delta$ implies $E(|X|1_A] < \epsilon$ (Step 1). Then choose M so that $E|X|/M < \delta$ and use Step 2.

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Uniform integrability of conditional expectation

Corollary

If ξ_1, ξ_2, \ldots are i.i.d. with finite expectation μ , and $S_n = \xi_1 + \cdots + \xi_n$, then

$$E\left|\frac{S_n}{n}-\mu\right|\to 0.$$

Proof.

As $S_n/n = E[\xi_1|S_n]$, S_n/n are u.i.



Theorem

Assume X_n is a submartingale. TFAE:

- (i) X_n u.i.
- (ii) X_n converge a.s. and in L^1 . (There exist $X \in L^1$ so that $X_n \to X$ a.s. and $E|X_n X| \to 0$.)
- (iii) X_n converge in L^1 . (There exist $X \in L^1$ so that $E|X_n X| \to 0$.)

Proof.

- $((i)\Longrightarrow(ii))$ (i) also implies a.s. convergence by martingale CT and (1) of the previous theorem. Now (5) of the same theorem implies (ii).
- $((iii) \Longrightarrow (i))$ (iii) also implies a.s. convergence by martingale CT, and (7) of the previous theorem implies (i).

Theorem

Assume X_n is a martingale. TFAE:

- (i) X_n u.i.
- (ii) X_n converge a.s. and in L^1 .
- (iii) X_n converge in L^1 .
- (iv) There exists an $X \in L^1$ so that $X_n = E[X \mid \mathcal{F}_n]$.

Proof.

((ii) \Longrightarrow (iv)) Let $X = \lim X_n$. Then $E[X_m \mid \mathcal{F}_n] = X_n$ for $m \ge n$ means that, for all $A \in \mathcal{F}_n$ and $m \ge n$, $E[X_m 1_A] = E[X_n 1_A]$. But

$$|E[X_m 1_A] - E[X 1_A]| \le E|X_m - X| \to 0$$

as $m \to \infty$, and so $E[X1_A] = E[X_n1_A]$.



Theorem

Assume that $E|X| < \infty$ and $\mathcal{F}_1 \subset \mathcal{F}_2 \subset \cdots$ is a filtration. Let $\mathcal{F}_{\infty} = \sigma(\cup_{n \geq 1} \mathcal{F}_n)$. Then

$$E[X\mid \mathcal{F}_n] \to E[X\mid \mathcal{F}_\infty]$$

a.s. and in L^1 .

Proof.

The sequence $X_n = E[X \mid \mathcal{F}_n]$ is a u.i. martingale, hence $X_n \to Y = \limsup X_n$ a.s. and in L^1 . Note that Y is \mathcal{F}_{∞} -measurable. By previous proof, $E[X \mid \mathcal{F}_n] = E[Y \mid \mathcal{F}_n]$ for all n, and so $E[X1_A] = E[Y1_A]$ for all $A \in \bigcup_{n \geq 1} \mathcal{F}_n$, which is a π -system that generates \mathcal{F}_{∞} .

Corollary (Lévy 0-1 law)

Assume $\mathcal{F}_1 \subset \mathcal{F}_2 \subset \cdots$ is a filtration. For any $A \in \mathcal{F}_{\infty} = \sigma(\cup_{n \geq 1} \mathcal{F}_n)$,

$$E[1_A \mid \mathcal{F}_n] \rightarrow 1_A \text{ a.s.}$$

Observe that this is more general than Kolmogorov 0-1 law: in that context, if A is in the tail σ -algebra, it is independent of each \mathcal{F}_n and so $E[1_A \mid \mathcal{F}_n] = P(A)$.

Theorem

Assume that $E|X|<\infty$ and $\mathcal{F}_1\subset\mathcal{F}_2\subset\cdots$ is a filtration. Let $\mathcal{F}_\infty=\sigma(\cup_{n\geq 1}\mathcal{F}_n)$. Assume that a sequence of r.v.'s X_n converges to X a.s. and $Y=\sup_n|X_n|\in L^1$. Then

$$E[X_n\mid \mathcal{F}_n] \to E[X\mid \mathcal{F}_\infty]$$

a.s. and in L^1 .

Proof.

Let $W_N = \sup\{|X_m - X_n| : m, n \ge N\}$. Note that $W_N \le 2Y$. So, by the previous theorem, for all N,

$$\limsup_n E[|X_n - X| \mid \mathcal{F}_n] \leq \limsup_n E[W_N \mid \mathcal{F}_n] = E[W_N \mid \mathcal{F}_\infty].$$

As $N \to \infty$, $W_N \downarrow 0$ a.s. By conditional DCT, $E[W_N \mid \mathcal{F}_\infty] \downarrow 0$. Then

$$\limsup_n |E[X_n \mid \mathcal{F}_n] - E[X \mid \mathcal{F}_n]| \leq \limsup_n E[|X_n - X| \mid \mathcal{F}_n] = 0$$

a.s. The L^1 convergence follows from

$$\begin{aligned} & E|E[X_n \mid \mathcal{F}_n] - E[X \mid \mathcal{F}_\infty]| \\ & \leq E|E[X_n \mid \mathcal{F}_n] - E[X \mid \mathcal{F}_n]| + E|E[X \mid \mathcal{F}_n] - E[X \mid \mathcal{F}_\infty]| \\ & \leq E|X_n - X| + E|E[X \mid \mathcal{F}_n] - E[X \mid \mathcal{F}_\infty]| \to 0. \end{aligned}$$