

CHAPTER IS :

L^p MARTINGALES

- Goals:
- maximal inequalities
 - convergence Thms for L^p -martingales.

- Setup:
- (Ω, \mathcal{F}, P) fixed probability space.
 - $(\mathcal{F}_n)_{n \geq 0}$ filtration.
 - $L^p = \{X : \Omega \rightarrow \mathbb{R} \text{ n.v. s.t. } E(|X|^p) < \infty\}, p > 1.$

1 AN OPTIONAL STOPPING RESULT.

Lemma (an optional stopping result for submart.).

Let $(X_n)_{n \geq 0}$ be a submartingale, T a stopping time.

For every $n \in \mathbb{N}$, we have

$$E(X_0) \leq E(X_{n \wedge T}) \leq E(X_n).$$

Proof. $E(X_0) \leq E(X_{n \wedge T})$ holds because $(X_{n \wedge T})_{n \geq 0}$ submart.

For $n \geq 0$, we have

$$(X_n - X_{n \wedge T}) = \sum_{k=1}^n 1_{k > T} \cdot (X_k - X_{k-1})$$

$$= (H \cdot X)_n,$$

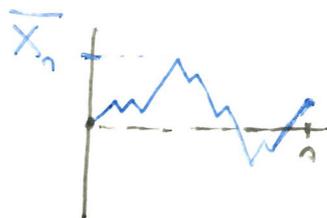
where $H_n = 1_{T < n}$ \mathcal{F}_{n-1} -meas., bounded.

Hence $(X_n - X_{n \wedge T})$ submartingale, and we have

$$E(X_n - X_{n \wedge T}) \geq E(X_0 - X_{0 \wedge T}) = 0 \quad \blacksquare$$

2 DOOB MAXIMAL INEQUALITY.

Notation: $\overline{X}_n = \max_{0 \leq k \leq n} X_k^+$



Thm (Doob maximal inequality).

Let $(X_n)_{n \geq 0}$ be a submartingale. For every $n \geq 0$

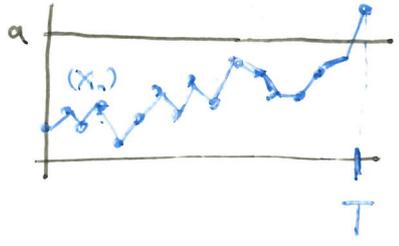
for every $a > 0$, we have

$$P(\overline{X}_n \geq a) \leq \frac{1}{a} E(X_n 1_{\overline{X}_n \geq a}) \leq \frac{1}{a} E(X_n^+).$$

" X_n^+ controls the max \overline{X}_n "

Rk: . Markov inequality would give $P(\bar{X}_n \geq a) \leq \frac{1}{a} E(\bar{X}_n)$.

Proof: Let $T = \min \{n \in \mathbb{N} : X_n \geq a\}$
($a > 0$ fixed). Let $n \geq 0$.



Rk: $T \leq n \Leftrightarrow \bar{X}_n \geq a$

We have $E(X_{T \wedge n}) = E(X_T 1_{T \leq n}) + E(X_n 1_{T > n})$

$E(X_n) = E(X_n 1_{T \leq n}) + E(X_n 1_{T > n})$

Since $E(X_{T \wedge n}) \leq E(X_n)$ (lemma), we get

$E(\underbrace{X_T}_{\geq a} 1_{T \leq n}) \leq E(X_n 1_{T \leq n})$

Hence, a $P(\bar{X}_n \geq a) \leq E(X_n 1_{\bar{X}_n \geq a})$.

For the second inequality, use $X_n 1_A \leq X_n^+ 1_A \leq X_n^+$ for every A .

3 APPLICATIONS TO MARTINGALES

Reminder: φ convex
 (X_n) mart } $\Rightarrow (\varphi(X_n))$ submart.

\hookrightarrow we can apply Doob max inequality to $(\varphi(X_n))$.

Not.: For $n \geq 0$, write $X_n^* = \max_{0 \leq k \leq n} |X_k|$.

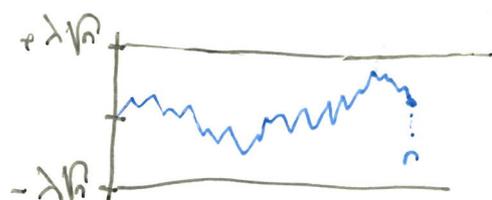
Prop. Let $(X_n)_{n \geq 0}$ be a martingale. For every $n \in \mathbb{N}$, for every $a > 0$, we have

$$P(X_n^* \geq a) \leq \frac{1}{a} E(|X_n|).$$

Pf. $(|X_n|)_{n \geq 0}$ is a non negative submartingale. Hence we can apply Doob maximal inequality.

Application to SRW

Z_1, Z_2, \dots iid $\mathcal{U}(\{-1, +1\})$. $X_n = Z_1 + \dots + Z_n$



(X_n^2) submartingale. For all $\lambda > 0$

$$P\left(\max_{i \leq n} |X_i| \geq \lambda \sqrt{n}\right) = P\left(\max_{i \leq n} X_i^2 \geq \lambda^2 n\right)$$

$$\stackrel{\text{Doob}}{\leq} \frac{E(X_n^2)}{\lambda^2 n}$$

$$= \frac{1}{\lambda^2}$$

4 DOOB L^p INEQUALITY.

Thm: Let $(X_n)_{n \geq 0}$ be a submartingale. For every $p \in (1, +\infty)$,

$$E\left(\bar{X}_n^p\right) \leq \left(\frac{p}{p-1}\right)^p E\left((X_n^+)^p\right)$$

Rk.: $E((X_n^+)^p) \leq E(\bar{X}_n^p) \leq C(p) E((X_n^+)^p)$

\uparrow trivial \uparrow Doob L^p

• in the deterministic case, if $(u_n) \uparrow$ we have

$$u_n = \max_{i \leq n} u_i$$

Exercise. (generalization of Tailsum).

Let Z be a non negative n.v. ($Z \geq 0$ a.s.). $p \in [1, \infty)$

$$E(Z^p) = p \int_0^\infty t^{p-1} P(Z \geq t) dt$$

Reminder (Hölder). Let $p, q \in (1, \infty)$ s.t. $\frac{1}{p} + \frac{1}{q} = 1$

For every $Y, Z \geq 0$ n.v.s

$$E(YZ) \leq E(Y^p)^{\frac{1}{p}} E(Z^q)^{\frac{1}{q}}$$

Proof of Thm: Let $n \geq 0$. set $Y = X_n^+$, $Z = \overline{X}_n$.

$$E(Z^p) = p \int_0^\infty t^{p-1} P(Z \geq t) dt$$

Doob
$$\leq p \int_0^\infty t^{p-2} E(Y 1_{Z \geq t}) dt$$

Fubini
$$= p E\left(Y \cdot \underbrace{\left(\int_0^\infty t^{p-2} 1_{Z \geq t} dt\right)}_{= \int_0^Z t^{p-2} dt} \right)$$

$$= \frac{1}{p-1} Z^{p-1}$$

$$= \frac{p}{p-1} E(Y Z^{p-1})$$

Let $q \in (1, \infty)$ s.t. $\frac{1}{p} + \frac{1}{q} = 1$.

NB: $\frac{1}{p} + \frac{1}{q} = 1 \Leftrightarrow p+q = pq \Leftrightarrow (p-1)q = p$

By Hölder inequality, we get

$$E(Z^p) \leq \left(\frac{p}{p-1}\right) E(Y^p)^{\frac{1}{p}} E(Z^q)^{\frac{1}{q}}$$

Dividing by $E(Z^q)^{\frac{1}{q}}$ we obtain

$$E(Z^p)^{\frac{1}{p}} \leq \left(\frac{p}{p-1}\right) E(Y^p)^{\frac{1}{p}}$$



$$\text{Rk: } \left(\frac{p}{1-p}\right)^p \xrightarrow{p \rightarrow 1} \infty.$$

A L^1 -version requires a log:

$$E(\bar{X}_n) \leq \left(1 - \frac{1}{e}\right)^{-1} E(X_n^+ \log X_n^+) \quad (\text{see Exercises}).$$

Corollary. Let $(X_n)_{n \geq 0}$ be a martingale, $p \in (1, \infty)$.

For every $n \geq 1$, we have

$$E((X_n^*)^p) \leq \left(\frac{p}{p-1}\right)^p E(|X_n|^p).$$

Proof. $(|X_n|)$ is a submartingale.

Application z_1, z_2, \dots iid $\mathcal{U}(\{-1, 1\})$, $X_n = z_1 + \dots + z_n$

$$E\left(\max_{i \leq n} S_i^2\right) \leq 4 E(S_n^2) = 4n.$$

5 MARTINGALES BOUNDED IN L^p .

Rk. If a martingale $(X_n)_{n \geq 0}$ is bounded in $L^p, p > 1$
(ie $\sup_{n \geq 0} E(|X_n|^p) < \infty$), then in particular

it is UI: $X_n \xrightarrow{a.s., L^1} X_\infty$ and $\forall n, X_n = E(X_\infty | \mathcal{F}_n)$

Not. $X_\infty^* = \sup_{n \geq 0} |X_n| = \lim_{n \rightarrow \infty} \uparrow X_n^*$

Thm: Let $p \in (1, \infty)$. Let $(X_n)_{n \geq 0}$ be a martingale s.t.
 $\sup E(|X_n|^p) < \infty$.

Then $(X_n)_{n \geq 0}$ cv a.s. and in L^p to a n.v. $X_\infty \in L^p$.
Furthermore

$$E(|X_\infty^*|^p) \leq \left(\frac{p}{p-1}\right)^p E(|X_0|^p)$$

"Doob L^p at time $n = +\infty$ "

Rk: The thm does not hold in general for $p = 1$.

$$\sup E(|X_n|) < \infty \Rightarrow \text{a.s cv but not } L^1 \text{ cv in general.}$$

• $E(|X_\infty^*|^p) \underset{\substack{\uparrow \\ L^p\text{-cv.}}}{=} \lim_{n \rightarrow \infty} E(|X_n|^p) = \sup_{n \geq 0} E(|X_n|^p)$.
 $|X_n|^p$ submart.

Proof. Since (X_n) bounded in L^1 , we have

$$X_n \xrightarrow{\text{a.s.}} X_\infty \text{ where } X_\infty \in L^1$$

By Doob L^p -ineq., for every $n \geq 0$ we have

$$\begin{aligned}
 E(|X_n^*|^p) &\leq \left(\frac{p}{p-1}\right)^p E(|X_n|^p) \quad (*) \\
 \text{Monotone cv.} \downarrow & \\
 E(|X_\infty^*|^p) &\leq C := \sup_{n \geq 0} E(|X_n|^p)
 \end{aligned}$$

Hence $(X_\infty^*)^p \in L^p$.

For every $n \geq 0$ $|X_n - X_\infty| \leq |X_n| + |X_\infty| \leq 2|X_\infty^*|$ a.s.

Hence, by dominated cv. $X_n \xrightarrow{L^p} X_\infty$.

In particular, we can take the limit in the inequality (*) to get

$$E(|X_\infty^*|^p) \leq \left(\frac{p}{p-1}\right)^p E(|X_n|^p)$$

CHAPTER 16.
 BACKWARD MARTINGALES

Setup: (Ω, \mathcal{F}, P) proba. space.

1 BACKWARD FILTRATION.

Def. A backward filtration is a sequence $(\mathcal{F}_n)_{n \in \mathbb{Z}_-}$ of σ -algebras $(\mathcal{F}_n \subset \mathcal{F})$ s.t.

$$\forall n \in \mathbb{Z}_- \quad \mathcal{F}_{n-1} \subset \mathcal{F}_n$$

($\mathbb{Z}_- = \{0, -1, -2, \dots\}$)

"gain of information whentime increases"
 \rightarrow

Filtration :



Backward filtration



"loss of information when we go back in time"
 \leftarrow

Notation:

$$\mathcal{F}_{-\infty} = \bigcap_{n \in \mathbb{Z}_-} \mathcal{F}_n$$

Example. $(X_n)_{n \leq 0}$ n.v. $\mathcal{F}_n = \sigma(X_n, X_{n-1}, X_{n-2}, \dots)$ back. filt..

In the rest of the chapter, we fix a backward filtration $(\mathcal{F}_n)_{n \leq 0}$

2 BACKWARD MARTINGALES

Def. A sequence of r.v.s $(X_n)_{n \leq 0}$ is a backward martingale if

- $\forall n \leq 0$ X_n is \mathcal{F}_n -measurable and $X_n \in L^1$
- $\forall n \leq 0$ $E(X_n | \mathcal{F}_{n-1}) = X_{n-1}$ a.s.

Ex1 Let $X \in L^1$. $X_n = E(X | \mathcal{F}_n)$ defines a backward martingale.

$$\begin{aligned}
 (\forall n \leq 0 \quad E(X_n | \mathcal{F}_{n-1}) &= E(E(X | \mathcal{F}_n) | \mathcal{F}_{n-1})) \\
 &= E(X | \mathcal{F}_{n-1})
 \end{aligned}$$

$\mathcal{F}_{n-1} \subset \mathcal{F}_n$

Ex2 Let z_{-1}, z_{-2}, \dots iid $\text{Ber}(\frac{1}{2})$.

$$\mathcal{F}_n = \sigma(z_{n-1}, z_{n-2}, \dots)$$

$$X_0 := \sum_{n \leq -1} z_n 2^n = z_{-1} \cdot \frac{1}{2} + z_{-2} \cdot \frac{1}{4} + z_{-3} \cdot \frac{1}{8} + \dots$$

$$X_{-1} = \frac{1}{4} + \sum_{k \leq -2} z_k 2^k = \frac{1}{4} + z_{-2} \cdot \frac{1}{4} + z_{-3} \cdot \frac{1}{8} + \dots$$

replace by average

$$X_{-2} = \frac{1}{4} + \frac{1}{8} + \sum_{k \leq -3} z_k 2^k = \frac{1}{4} + \frac{1}{8} + z_{-3} \cdot \frac{1}{8} + \dots$$

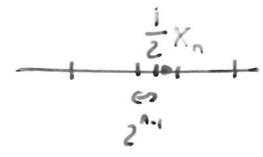
⋮

$$X_n = \sum_{n \leq k \leq -1} \frac{1}{2} \cdot 2^k + \sum_{k \leq n} z_k 2^k$$

Rk: $X_0 \sim U([0, 1])$

$X_1 \sim U([\frac{1}{4}, \frac{3}{4}])$

$X_n \sim U([\frac{1}{2} - 2^{n-1}, \frac{1}{2} + 2^{n-1}])$



Exercise: check that $(X_n)_{n \leq 0}$ is a backward martingale.

Rk: Let Z_1, Z_2, \dots iid $Ber(\frac{1}{2})$ $\mathcal{F}_n = \sigma(Z_1, \dots, Z_n)$ filtration

$X_n = Z_1 \cdot \frac{1}{2} + \dots + Z_n \cdot \frac{1}{2^n}$ (forward) - martingale.

Prop. Let (X_n) be a backward martingale. Then for every $m, n \leq 0$ s.t. $m \leq n$, we have

$E(X_n | \mathcal{F}_m) = X_m$

Proof. Fix $m \leq 0$. let $k \in \{1, \dots, m\}$

$E(X_{m+k} | \mathcal{F}_m) = E(E(X_{m+k} | \mathcal{F}_{m+k-1}) | \mathcal{F}_m)$
 $\mathcal{F}_m \subset \mathcal{F}_{m+k-1}$

$= E(X_{m+k-1} | \mathcal{F}_m)$ a.s.

And the result follows by finite induction.

Rk: In particular $\forall n \leq 0$ $X_n = E(X_0 | \mathcal{F}_n)$

and $E(X_n) = E(X_0)$

3 CONVERGENCE THM

Thm. Let $(X_n)_{n \leq 0}$ be a backward martingale.

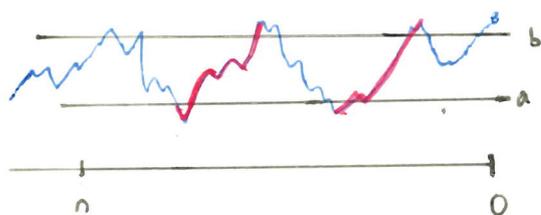
$$X_n \xrightarrow{n \rightarrow -\infty} E(X_0 | \mathcal{F}_{-\infty}) \text{ a.s. and in } L^1.$$

Rk: no hypothesis because a back. martingale is automatically "closed", In particular, since $X_n = E(X_0 | \mathcal{F}_n)$, the sequence is automatically U. I.

Pp. sketch: a.s. cv.

For $a < b$ $a, b \in \mathbb{Q}$, define

$$N_n(a, b) = \{ \text{upcrossings of } (X_n, \dots, X_0) \}$$



Doob upcrossing inequality:

$$\forall a < b \quad \underbrace{E(N_n(a, b))}_{\downarrow \text{ monotone cv}} \leq \frac{1}{b-a} E((X_0 - a)^-)$$

$$E(N_{-\infty}(a, b))$$

"total number of upcrossings"

Hence a.s. we have

$$\forall a < b \quad N_{-\infty}(a, b) < \infty, \\ a, b \in \mathbb{R}$$

which deterministically implies the existence of $X_{-\infty}$ s.t.

$$X_n \xrightarrow[n \rightarrow -\infty]{} X_{-\infty} \text{ a.s.}$$

L^1 cv and identification of the limit.

Since $X_n = E(X_0 | \mathcal{F}_n)$, the sequence (X_n) is UI. Since it cv a.s., it also cv in L^1 .

We have $X_n \xrightarrow[n \rightarrow -\infty]{L^1} X_{-\infty}$. (NB: $X_{-\infty}$ is $\mathcal{F}_{-\infty}$ -meas.)

Let $A \in \mathcal{F}_{-\infty}$. For every $n \leq 0$, $A \in \mathcal{F}_n$, hence

$$\begin{aligned} E(X_0 1_A) &= E(E(X_0 | \mathcal{F}_n) 1_A) \\ &= E(X_n 1_A) \\ &\xrightarrow[n \rightarrow -\infty]{} E(X_{-\infty} 1_A) \quad (\text{because } X_n \rightarrow X_{-\infty}) \end{aligned}$$

Therefore $X_{-\infty} = E(X_0 | \mathcal{F}_{-\infty})$ a.s. □

Application: Let $Z \in L^1$

$$E(Z | \mathcal{F}_n) \xrightarrow{L^1, \text{ a.s.}} E(Z | \mathcal{F}_{-\infty})$$

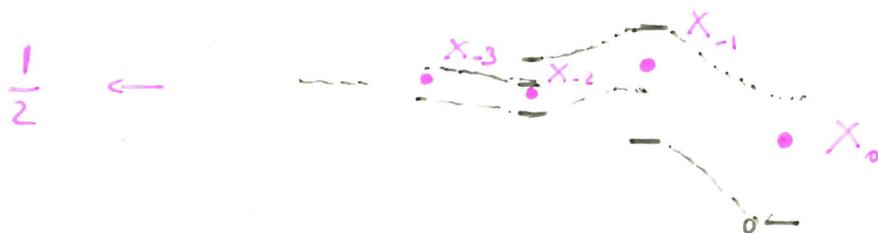
Appt. 2 : $X_n = \frac{1}{2} - 2^{-n-1} + \sum_{k \leq n-1} z_k 2^k$

$$X_n \xrightarrow{n \rightarrow -\infty} \frac{1}{2} \text{ a.s. and in } L^1$$

Pf. 1 : $|\sum_{k \leq n-1} z_k 2^k| \leq \sum_{k \leq n-1} 2^k = 2^n \rightarrow 0$ as $n \rightarrow -\infty$.

Pf 2 : $X_{-\infty}$ is $\mathcal{F}_{-\infty}$ -meas. Since $\mathcal{F}_{-\infty}$ is P -trivial
 ($\forall A \in \mathcal{F}_{-\infty} \quad P(A \in \{0, 1\})$) $X_{-\infty} = C$ a.s.
 where $C = \text{cte.}$ and

$$C = E(X_{-\infty}) = E(X_0) = \frac{1}{2}.$$



4 LAW OF LARGE NUMBERS REVISITED.

Let z_1, z_2, \dots iid $z_i \in L'$. For $k \geq 0$, $S_k = z_1 + \dots + z_k$

We give a new proof of $\frac{S_k}{k} \xrightarrow[k \rightarrow \infty]{} E(z_1)$ a.s. and in L' .

Define for $k \geq 0$

$$\mathcal{F}_{-k}^1 = \sigma(S_{k+1}, z_{k+2}, z_{k+3}, \dots)$$

$$X_{-k} = \frac{S_{k+1}}{k+1}$$

$$\mathcal{F}_0 = \sigma(z_1, z_2, \dots) \quad X_0 = z_1$$

$$\mathcal{F}_{-1} = \sigma(S_2, z_2, \dots) \quad X_{-1} = \frac{z_1 + z_2}{2}$$

$$\mathcal{F}_{-2} = \sigma(S_3, z_3, \dots) \quad X_{-2} = \frac{z_1 + z_2 + z_3}{3}$$

⋮

Rk: $(\mathcal{F}_n)_{n \leq 0}$ is a backward filtration; Indeed, for all $k \geq 0$

$$\mathcal{F}_{-k-1}^1 = \sigma(S_{k+2}, z_{k+3}, \dots)$$

$$\subset \sigma(S_{k+1}, z_{k+2}, z_{k+3}, \dots) = \mathcal{F}_{-k}^1$$

because $S_{k+2} = S_{k+1} + z_{k+2}$

Prop. $(X_n)_{n \leq 0}$ is a $(\mathcal{F}_n)_{n \leq 0}$ -backward martingale.

Proof. We show that

$$\forall n \leq 0 \quad X_n = E(z_1 \mid \mathcal{F}_n).$$

Fix $k \geq 1$. We prove

$$\frac{S_k}{k} = E(z_1 \mid S_k, z_{k+1}, \dots)$$

Since z_{k+1}, z_{k+2}, \dots are indep of (z_1, S_k)

it suffices to show.

$$\frac{S_k}{k} = E(z_1 \mid S_k).$$

By symmetry, we have

$$\forall i \leq k \quad E(z_i \mid S_k) = E(z_1 \mid S_k)$$

(To prove this statement, consider $A \in \sigma(S_k)$

ie $1_A = \varphi(S_k) = \varphi(z_1 + \dots + z_k)$, for some φ meas. bounded.

Since $(z_1, S_k) \stackrel{(d)}{=} (z_i, S_k)$, we have

$$E(z_1 \varphi(z_1 + \dots + z_k)) = E(z_i \varphi(z_1 + \dots + z_k))$$

$$\text{ie } E(z_1 1_A) = E(z_i 1_A)$$

$$\begin{aligned}
 \text{Hence } k E(z_i | S_k) &= E(z_i | S_k) + \dots + E(z_k | S_k) \\
 &= E(S_k | S_k) \\
 &= S_k
 \end{aligned}$$

Hence $\exists X_{-\infty} \in L^1$ s.t. $X_n \xrightarrow{n \rightarrow \infty} X_{-\infty}$ a.s. and in L^1
 By Kolmogorov 0-1 Law,

$$X_{-\infty} = \lim_{k \rightarrow \infty} \frac{S_k}{k} \text{ is constant a.s.}$$

(because $\forall a \{ \lim_{k \rightarrow \infty} \frac{S_k}{k} \leq a \} \in \mathcal{G} = \bigcap_{k \geq 0} \sigma(z_k, z_{k+1}, \dots)$)

$\exists c$ s.t. $X_{-\infty} = c$ a.s.

But $X_{-\infty} = E(z_i | \mathcal{G}_{-\infty})$ so we must have

$$\boxed{E(z_i) = c}$$

This concludes

$$\frac{z_1 + \dots + z_k}{k} \xrightarrow{k \rightarrow \infty} E(z_i) \text{ a.s. and in } L^1$$