

# Notes 14 : Martingales in $L^p$

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Lecturer: Sebastien Roch

References: [Wil91, Section 12], [Dur10, Section 5.4].

## 1 Martingales in $L^2$

### 1.1 Preliminaries

**DEF 14.1** For  $1 \leq p < +\infty$ , we say that  $X \in \mathcal{L}^p$  if

$$\|X\|_p = \mathbb{E}[|X|^p]^{1/p} < +\infty.$$

By Jensen's inequality, for  $1 \leq p \leq r < +\infty$  we have  $\|X\|_p \leq \|X\|_r$  if  $X \in \mathcal{L}^r$ .

**Proof:** For  $n \geq 0$ , let

$$X_n = (|X| \wedge n)^p.$$

Take  $c(x) = x^{r/p}$  on  $(0, +\infty)$  which is convex. Then

$$(\mathbb{E}[X_n])^{r/p} \leq \mathbb{E}[(X_n)^{r/p}] = \mathbb{E}[ (|X| \wedge n)^r ] \leq \mathbb{E}[|X|^r].$$

Take  $n \rightarrow \infty$  and use (MON). ■

**DEF 14.2** We say that  $X_n$  converges to  $X_\infty$  in  $\mathcal{L}^p$  if  $\|X_n - X_\infty\|_p \rightarrow 0$ . By the previous result, convergence on  $\mathcal{L}^r$  implies convergence in  $\mathcal{L}^p$  for  $r \geq p \geq 1$ . (Moreover, by Chebyshev's inequality, convergence in  $\mathcal{L}^p$  implies convergence in probability.)

**LEM 14.3** Assume  $X_n, X_\infty \in \mathcal{L}^1$ . Then

$$\|X_n - X_\infty\|_1 \rightarrow 0,$$

implies

$$\mathbb{E}[X_n] \rightarrow \mathbb{E}[X_\infty].$$

**Proof:** Note that

$$|\mathbb{E}[X_n] - \mathbb{E}[X_\infty]| \leq \mathbb{E}|X_n - X_\infty| \rightarrow 0.$$

■

**DEF 14.4** We say that  $\{X_n\}_n$  is bounded in  $\mathcal{L}^p$  if

$$\sup_n \|X_n\|_p < +\infty.$$

## 1.2 $L^2$ convergence

**THM 14.5** Let  $M$  be a MG with  $M_n \in \mathcal{L}^2$ . Then  $M$  is bounded in  $\mathcal{L}^2$  if and only if

$$\sum_{k \geq 1} \mathbb{E}[(M_k - M_{k-1})^2] < +\infty.$$

When this is the case,  $M_n$  converges a.s. and in  $\mathcal{L}^2$ . (In particular, it converges in  $\mathcal{L}^1$ .)

**Proof:**

**LEM 14.6 (Orthogonality of increments)** Let  $s \leq t \leq u \leq v$ . Then,

$$\langle M_t - M_s, M_v - M_u \rangle = 0.$$

**Proof:** Use  $M_u = \mathbb{E}[M_v | \mathcal{F}_u]$ ,  $M_t - M_s \in \mathcal{F}_u$  and apply the  $L^2$  characterization of conditional expectations. ■

That implies

$$\mathbb{E}[M_n^2] = \mathbb{E}[M_0^2] + \sum_{1 \leq i \leq n} \mathbb{E}[(M_i - M_{i-1})^2],$$

proving the first claim.

By monotonicity of norms,  $M$  is bounded in  $L^2$  implies  $M$  bounded in  $L^1$  which, in turn, implies  $M$  converges a.s. Then using (FATOU) in

$$\mathbb{E}[(M_{n+k} - M_n)^2] = \sum_{n+1 \leq i \leq n+k} \mathbb{E}[(M_i - M_{i-1})^2],$$

gives

$$\mathbb{E}[(M_\infty - M_n)^2] \leq \sum_{n+1 \leq i} \mathbb{E}[(M_i - M_{i-1})^2].$$

The RHS goes to 0 which proves the second claim. ■

## 2 $L^p$ convergence theorem

Recall:

**LEM 14.7 (Markov's inequality)** Let  $Z \geq 0$  be a RV. Then for  $c > 0$

$$c\mathbb{P}[Z \geq c] \leq \mathbb{E}[Z; Z \geq c] \leq \mathbb{E}[Z].$$

MGs provide a useful generalization.

**LEM 14.8 (Doob's submartingale inequality)** Let  $Z \geq 0$  a subMG. Then for  $c > 0$

$$c\mathbb{P}\left[\sup_{1 \leq k \leq n} Z_k \geq c\right] \leq \mathbb{E}[Z_n; \sup_{1 \leq k \leq n} Z_k \geq c] \leq \mathbb{E}[Z_n].$$

**Proof:** Divide  $F = \{\sup_{1 \leq k \leq n} Z_k \geq c\}$  according to the first time  $Z$  crosses  $c$ :

$$F = F_0 \cup \dots \cup F_n,$$

where

$$F_k = \{Z_0 < c\} \cap \dots \cap \{Z_{k-1} < c\} \cap \{Z_k \geq c\}.$$

Since  $F_k \in \mathcal{F}_k$  and  $\mathbb{E}[Z_n | \mathcal{F}_k] \geq Z_k$ ,

$$c\mathbb{P}[F_k] \leq \mathbb{E}[Z_k; F_k] \leq \mathbb{E}[Z_n; F_k].$$

Sum over  $k$ . ■

**EX 14.9 (Kolmogorov's inequality)** Let  $X_1, \dots$  be independent RVs with  $\mathbb{E}[X_k] = 0$  and  $\text{Var}[X_k] < +\infty$ . Define  $S_n = \sum_{k \leq n} X_k$ . Then for  $c > 0$

$$\mathbb{P}[\max_{k \leq n} |S_k| \geq c] \leq c^{-2} \text{Var}[S_n].$$

**THM 14.10 (Doob's  $L^p$  inequality)** Let  $p > 1$  and  $p^{-1} + q^{-1} = 1$ . Let  $Z \geq 0$  a subMG bounded in  $L^p$ . Define

$$Z^* = \sup_{k \geq 0} Z_k.$$

Then

$$\|Z^*\|_p \leq q \sup_k \|Z_k\|_p = q \uparrow \lim_k \|Z_k\|_p.$$

and  $Z^* \in L^p$ .

**Proof:** The last equality follows from (JENSEN). Let  $Z_n^* = \sup_{k \leq n} Z_k$ . By (MON) it suffices to prove:

**LEM 14.11**

$$\mathbb{E}[(Z_n^*)^p] \leq q^p \mathbb{E}[Z_n^p].$$

**Proof:** Recall the formula: for  $Y \geq 0$  and  $p > 0$

$$\mathbb{E}[Y^p] = \int_0^\infty p y^{p-1} \mathbb{P}[Y \geq y] dy.$$

Then for  $K > 0$  (note that  $\{Z_n^* \wedge K \geq c\}$  is either  $\{Z_n^* \geq c\}$  or empty (depending on whether  $K$  is smaller or bigger than  $c$ ) so Doob's inequality still applies)

$$\begin{aligned} \mathbb{E}[(Z_n^* \wedge K)^p] &= \int_0^\infty p c^{p-1} \mathbb{P}[Z_n^* \wedge K \geq c] dc \\ &\leq \int_0^\infty p c^{p-2} \mathbb{E}[Z_n; Z_n^* \wedge K \geq c] dc \\ &= \mathbb{E} \left[ Z_n \left( \frac{p}{p-1} \right) \int_0^\infty (p-1) c^{p-2} \mathbb{1}[Z_n^* \wedge K \geq c] dc \right] \\ &= \mathbb{E}[q Z_n (Z_n^* \wedge K)^{p-1}] \\ &\leq q \mathbb{E}[Z_n^p]^{1/p} \mathbb{E}[(Z_n^* \wedge K)^p]^{1/q}, \end{aligned}$$

where we used that  $(p-1)q = p$ . Rearranging and using (MON) gives the result. ■

**THM 14.12 ( $L^p$  convergence)** Let  $M$  be a MG bounded in  $L^p$  for  $p > 1$ . Then  $M_n \rightarrow M_\infty$  a.s. and in  $L^p$ .

**Proof:** Note that  $|M_n|$  is a subMG bounded in  $L^p$ . In particular, it is bounded in  $L^1$  and  $M_n \rightarrow M_\infty$  a.s. From the previous theorem,

$$|M_n - M_\infty|^p \leq (2 \sup_k |M_k|)^p \in L^1,$$

and by (DOM)

$$\mathbb{E}|M_n - M_\infty|^p \rightarrow 0.$$

■

## References

- [Dur10] Rick Durrett. *Probability: theory and examples*. Cambridge Series in Statistical and Probabilistic Mathematics. Cambridge University Press, Cambridge, 2010.
- [Wil91] David Williams. *Probability with martingales*. Cambridge Mathematical Textbooks. Cambridge University Press, Cambridge, 1991.