# Stochastic Processes

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## 目录

2 Markov Chains		

1 Review of Martingales

### 1 Review of Martingales

- $(X_n)_{n>0}$  is  $L^2$ -bounded martingale  $\Rightarrow X_n$  converges in  $L^2$ .
- $(X_n)_{n>0}$  is  $L^1$ -bounded martingale  $\Rightarrow X_n$  converges a.s.
- (1) + (2): If  $(X_n)_{n\geq 0}$  is  $L^p$ -bounded martingale for p>1, then  $X_n$  converges in  $L^{p'}$  for  $p'\in [1,p)$ .
- Statement is false when p=1. Example:  $\Omega=[0,1), \mathscr{F}_n=\sigma\{[\frac{i}{2^n},\frac{i+1}{2^n})\}_{i=0}^{2^n-1}, X_n(\omega):=\begin{cases} 2^n & \omega\in[0,\frac{1}{2^n})\\ 0 & \text{otherwise} \end{cases}$ .
- Let p > 1 and  $(X_n)_{n \ge 0}$  be  $L^p$  bounded martingale w.r.t.  $\mathscr{F}_n$ . Then  $\exists X \in L^p(\Omega, \mathscr{F}_\infty, P)$  s.t.  $X_n \to X$  in  $L^p$  and a.s. and  $X_n = \mathbb{E}(X|\mathscr{F}_n)$ .
- Doob's maximal inequality: Let p > 1,  $\exists C = C_p$  s.t.  $\forall$  martingale  $(X_n)_{n \geq 0}$ , we have  $\mathbb{E}|X_n^*|^p \leq C_p \mathbb{E}|X_n|^p$  where  $|X_n^*| = \sup_{0 \leq k \leq n} \sup |X_k|$ .
- Let  $(Z_n)_{n\geq 0}$  be a nonnegative sub-martingale and  $Z_n^* = \sup_{0\leq k\leq n} Z_k$ , then  $P(Z_n^* > \lambda) \leq \frac{1}{\lambda} \mathbb{E}(Z_n 1_{\{Z_n^* > \lambda\}}) \leq \frac{1}{\lambda} \mathbb{E}Z_n$ . Corollary:  $P(Z_n^* > \lambda) \leq \frac{1}{\lambda^p} \mathbb{E}(Z_n^p 1_{\{Z_n^* > \lambda\}}) \leq \frac{1}{\lambda^p} \mathbb{E}(Z_n^p)$ .
- If  $(X_n)_{n\geq 0}$  is a martingale with  $\sup_n \mathbb{E}(|X_n|\log(1+|X_n|)) < +\infty$ , then  $X_n$  converges in  $L^1$ .
- Two prob measures P and Q on  $(\Omega, \mathscr{F})$ , Q << P on  $\mathscr{F}_n$  for every n and  $M_n = \frac{dQ}{dP}|_{\mathscr{F}_n}$ .  $(M_n)_{n\geq 0}$  is a P-martingale w.r.t.  $(\mathscr{F}_n)_{n\geq 0}$ . Q << P on  $\mathscr{F}_\infty$  if and only if  $M_n \to M$  in  $L^1$ .  $Q(A) = \int_A M dP + Q(A \cap \{M = +\infty\})$ .
- Statement is false if  $M_n \not\to M$  in  $L^1$ . Example:  $\Omega = \{\omega = (\omega_1, \dots, \omega_n, \dots) \in \{\pm 1\}^{\mathbb{N}}\}, X_n(\omega) = \omega_n$ .  $X_n$ 's are i.i.d. under P and Q, but  $P(X_n = 1) = \frac{1}{2}, P(X_n = -1) = \frac{1}{2}, Q(X_n = 1) = \frac{1}{3}, Q(X_n = -1) = \frac{2}{3}$ .  $\mathscr{F}_n = \sigma(X_1, \dots, X_n)$ .  $P(\lim_{n \to \infty} \frac{1}{n} \sum_{k=1}^n X_k = 0) = 1, Q(\lim_{n \to \infty} \frac{1}{n} \sum_{k=1}^n X_k = -\frac{1}{3}) = 1$ .
- Monotone class theorem for functions: Suppose  $\mathcal{A}$  us a  $\pi$ -system and  $\mathcal{H}$  be a class of functions from  $\Omega$  to  $\mathbb{R}$  s.t. (1)  $1_A \in \mathcal{H}$  for every  $A \in \mathcal{A}$ , (2) if  $f, g \in \mathcal{H}$  then  $af + bg \in \mathcal{H}$ , (3) if  $f_n \in \mathcal{H}$  and  $f_n \uparrow f$  then  $f \in \mathcal{H}$ . Then all nonnegative  $\sigma(\mathcal{A})$ -measurable functions are in  $\mathcal{H}$ .
- Let  $(Y_n)_{n\geq 0}$  be i.i.d., nonnegative r.v.'s with  $\mathbb{E}Y_k=1$ . Then  $M_n=\prod_{k=1}^n Y_k$  converges in  $L^1$  iff  $Y_n\equiv 1$ . Otherwise  $M_n\to 0$  a.s.
- Kakutani's theorem:  $M_n = \prod_{k=1}^n Y_k$ ,  $Y_k \ge 0$  are independent,  $\mathbb{E}Y_k = 1$ ,  $\lambda_k = \mathbb{E}\sqrt{Y_k}$ . (1) If  $\prod_k \lambda_k > 0$ , then  $M_n \to M$  in  $L^1$ ; (2) If  $\prod_k \lambda_k = 0$ , then  $M_n \to 0$  a.s.

#### 2 Markov Chains

- Let  $(X_n)_{n\geq 0}$  be a time-homogeneous Markov chain on a discrete space S.  $P^x$ : probability measure of  $(X_n)_{n\geq 0}$  conditioned on  $X_0 = x$ .  $P(X_{n+1} \in A | \mathscr{F}_n) = P^{X_n}(X_1 \in A) = P(X_1 \in A | X_0 = X_n)$ .  $\mathbb{E}^x$ : expectation under  $P^x$ .  $P^x(X_1 = y) = p(x, y)$ .
- For every  $f: S \to \mathbb{R}$  bounded, define  $(Pf)(x) = \sum_{y \in S} p(x, y) f(y) = \mathbb{E}^x (f(X_1)), (Lf)(x) = \sum_{y \in S} p(x, y) f(y) f(x)$ .  $L = P \mathrm{id}$ , the generator.
- Let  $(X_n)_{n\geq 0}$  be a homogeneous Markov chain with generator L. Then for every bounded  $f: S \to \mathbb{R}$ ,  $M_n = f(X_n) f(X_0) \sum_{k=0}^{n-1} (Lf)(X_k)$  is a martingale. Conversely, let  $(X_n)_{n\geq 0}$  be a process and L be an operator on  $\mathcal{B}(S)$  s.t.  $M_n^g$  is a martingale for every f, then  $(X_n)_{n\geq 0}$  is a Markov chain with generator L.

#### MARKOV CHAINS

- Given operator L on  $\mathcal{B}(\mathcal{S})$ , we say  $f: S \to \mathbb{R}$  is (1) harmonic for L if Lf = 0; (2) sub-harmonic for L if  $Lf \geq 0$ ; (3) super-harmonic for L if  $Lf \leq 0$ .
- Let f be the generator of a Markov chain  $(X_n)_{n\geq 0}$ . Then f is (sub-/super-)harmonic  $\Leftrightarrow f(X_n)_{n\geq 0}$  is a (sub-/super-) martingale.
- f is (sub-/super-)harmonic on  $D \subset S$  if  $Lf \leq / \geq / = 0$  on D. Let  $\tau = \inf\{k \geq 0 : X_k \in D^c\}$ , then  $(f(X_{n \wedge \tau}))_{n \geq 0}$  is a (sub-/super)martingale.
- Maximum principle: Let  $(X_n)_{n\geq 0}$  be a Markov chain and  $D\subset S$  s.t. the stopping time  $\tau=\inf\{k\geq 0, X_k\in D^c\}$  is a.s. finite. If f is bounded and sub-harmonic on D, then  $\sup_{x\in D}f(x)\leq \sup_{x\in D^c}f(x)$ .
- $A \subset S$ ,  $\tau_A = \sup\{k \ge 0 : X_k \in A\}$ ,  $u(x) = P^x(\tau_A < +\infty)$ . Then u(x) = 1 for  $x \in A$ ,  $u(x) = \sum_{y \in S} p(x, y)u(y) = (Pu)(x)$ ,  $\Rightarrow \begin{cases} Lu = 0 & \text{on } A^c \\ u = 1 & \text{on } A \end{cases}$ .
- Any nonnegative solution v to  $\begin{cases} Lv=0 & \text{on } A^c \\ v=1 & \text{on } A \end{cases}$  satisfies  $v\geq u$ . Furthermore, if  $u\equiv 1$ , then  $\exists 1$  bounded solution to  $\begin{cases} Lv=0 & \text{on } A^c \\ v=f & \text{on } A \end{cases}$  with  $v(x)=\mathbb{E}^x(f(X_{\tau_A}))$ .
- Doob's h-transform: Let h be positive, harmonic with  $h(x_0) = 1$  for some  $x_0 \in S \Rightarrow (h(X_n))_{n \geq 0}$  is a martingale with  $\mathbb{E}^{P^{x_0}}(h(X_n)) = 1$ .  $\exists 1$  measure  $Q^h$  on  $\mathscr{F}_{\infty}$  s.t.  $\frac{dQ^h}{dP^{x_0}}|_{\mathscr{F}_n} = h(X_n), \forall n \geq 0$ .  $Q(X_0 = x_0) = 1, (X_n)_{n \geq 0}$  never visits the set  $\{x : h(x) = 0\}$ . Under Q,  $(X_n)_{n \geq 0}$  is again a Markov chain on  $S \setminus D$  with transition probability  $q(x,y) = \frac{p(x,y)h(y)}{h(x)}$  (or equivalently,  $(L^h f)(x) = \frac{1}{h(x)}(L(hf))(x)$ ).