

APA lecture 12

Martingales

Def: Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space

A filtration is a sequence of sub- σ -fields of \mathcal{F}
 $(\mathcal{F}_n, n \in \mathbb{N})$

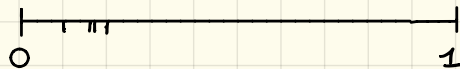
such that $\mathcal{F}_n \subset \mathcal{F}_{n+1}$, $\forall n \in \mathbb{N}$.

Ex: $\Omega = [0, 1]$, $\mathcal{F} = \mathcal{B}([0, 1])$, $\mathbb{P} =$ Lebesgue measure on $[0, 1]$

$X_n(\omega) = n^{\text{th}}$ decimal of $\omega \in [0, 1]$ $n \geq 1$

$\mathcal{F}_0 = \{\emptyset, \Omega\}$, $\mathcal{F}_1 = \sigma(X_1)$, $\mathcal{F}_2 = \sigma(X_1, X_2)$... $\mathcal{F}_n = \sigma(X_1, \dots, X_n)$
 $n \geq 1$

$\omega = 0,17392 \dots$



Def: Let $(\mathcal{F}_n, n \in \mathbb{N})$ be a filtration on $(\Omega, \mathcal{F}, \mathbb{P})$

A square-int. martingale w.r.t $(\mathcal{F}_n, n \in \mathbb{N})$ is a discrete-time stochastic process $(M_n, n \in \mathbb{N})$ such that

(i) $\mathbb{E}(M_n^2) < +\infty \quad \forall n \in \mathbb{N}$

(ii) M_n is \mathcal{F}_n -measurable $\forall n \in \mathbb{N}$ (adapted process)

\longrightarrow (iii) $\mathbb{E}(M_{n+1} | \mathcal{F}_n) = M_n$ a.s. $\forall n \in \mathbb{N}$

Remark: Because $\mathbb{E}(M_{n+1} | \mathcal{F}_n)$ is \mathcal{F}_n -measurable by definition, condition (ii) is actually redundant.

Remark: martingale \neq Markov process

Example: The simple symmetric random walk.

Let $(X_n, n \geq 1)$ be iid r.v.'s st. $P(\{X_1 = +1\}) = P(\{X_1 = -1\}) = \frac{1}{2}$

$$S_0 = 0, S_n = X_1 + \dots + X_n \quad n \geq 1$$

$$\mathcal{F}_0 = \{\emptyset, \Omega\}, \mathcal{F}_n = \sigma(X_1, \dots, X_n) \quad n \geq 1$$

Then $(S_n, n \in \mathbb{N})$ is a sq-int. martingale w.r.t $(\mathcal{F}_n, n \in \mathbb{N})$

Proof:

(i) $\mathbb{E}(S_n^2) = n \cdot \mathbb{E}(X_1^2) = n < +\infty \quad \forall n \geq 1$

(ii) S_n is \mathcal{F}_n -measurable: yes, as $S_n = X_1 + \dots + X_n$ ✓

(iii) $\mathbb{E}(S_{n+1} | \mathcal{F}_n) = \mathbb{E}(S_n + X_{n+1} | \mathcal{F}_n) = \mathbb{E}(S_n | \mathcal{F}_n) + \mathbb{E}(X_{n+1} | \mathcal{F}_n)$

$$= S_n + \mathbb{E}(X_{n+1}) = S_n \text{ a.s.}$$

\downarrow
 \mathcal{F}_n -measurable

\searrow
 $X_{n+1} \perp \mathcal{F}_n$

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Remark: If $(M_n, n \in \mathbb{N})$ is a sq-int. martingale, then

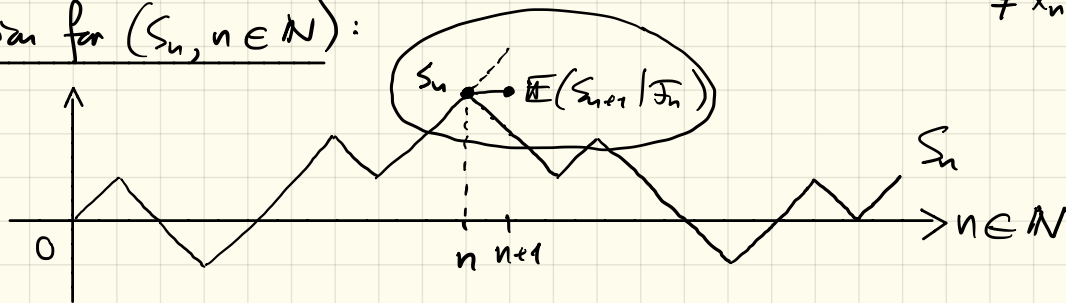
$$\mathbb{E}(M_{n+1}) = \mathbb{E}(\mathbb{E}(M_{n+1} | \mathcal{F}_n)) = \mathbb{E}(M_n) \quad \forall n \in \mathbb{N}$$

$$\text{So } \mathbb{E}(M_n) = \mathbb{E}(M_{n-1}) = \dots = \mathbb{E}(M_0) \quad \forall n \in \mathbb{N}$$

So a martingale is a process with constant expectation; but it is also much more than that!

Ex: The process $(X_n, n \geq 1)$ with X_n iid r.v. has constant expectation, but it is not a martingale: $\mathbb{E}(X_{n+1} | \mathcal{F}_n) = \mathbb{E}(X_{n+1}) = 0 \neq X_n$

Illustration for $(S_n, n \in \mathbb{N})$:

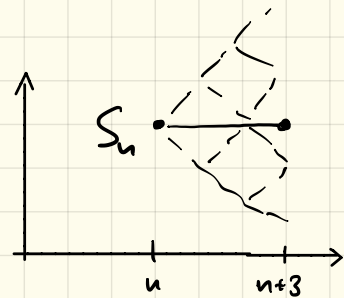


Property: Let $(M_n, n \geq 1)$ be a square-integrable martingale

- $$\mathbb{E}(M_{n+2} | \mathcal{F}_n) = \mathbb{E}(\underbrace{\mathbb{E}(M_{n+2} | \mathcal{F}_{n+1})}_{\mathcal{F}_n \subset \mathcal{F}_{n+1}} | \mathcal{F}_n) = \mathbb{E}(M_{n+1} | \mathcal{F}_n) = M_n$$
- $$\mathbb{E}(M_{n+m} | \mathcal{F}_n) = M_n \quad \forall n, m \in \mathbb{N}$$

Remark: martingale \neq Markov

- martingale property:
$$\mathbb{E}(M_{n+1} | \mathcal{F}_n) = M_n$$



- Markov property:
$$\mathbb{P}(\{M_{n+1} \in B\} | \mathcal{F}_n) = \mathbb{P}(\{M_{n+1} \in B\} | M_n) \quad \forall B \in \mathcal{B}(\mathbb{R})$$

equivalently:
$$\mathbb{E}(g(M_{n+1}) | \mathcal{F}_n) = \mathbb{E}(g(M_{n+1}) | M_n) \quad \forall g \in C_b(\mathbb{R})$$

Sub- and supermartingales (square-integrable)

Def: A process $(M_n, n \in \mathbb{N})$ is a $\left. \begin{array}{l} \text{sub-} \\ \text{super-} \end{array} \right\}$ martingale w.r.t. a filtration $(\mathcal{F}_n, n \in \mathbb{N})$ if:

$$(i) \mathbb{E}(M_n^2) < +\infty \quad \forall n \in \mathbb{N}$$

→ (ii) M_n is \mathcal{F}_n -measurable $\forall n \in \mathbb{N}$

$$(iii) \mathbb{E}(M_{n+1} | \mathcal{F}_n) \begin{array}{l} \geq \\ \leq \end{array} M_n \text{ a.s. } \forall n \in \mathbb{N}$$

Rem: 1) if M is a $\left. \begin{array}{l} \text{sub-} \\ \text{super-} \end{array} \right\}$ martingale, then $\mathbb{E}(M_{n+1}) \begin{array}{l} \geq \\ \leq \end{array} \mathbb{E}(M_n)$.

2) Not every process is either a submartingale, or a martingale, or a supermartingale.

Example (simple asymmetric random walk)

$S_0=0, S_n = X_1 + \dots + X_n$ with $(X_n, n \geq 1)$ iid r.v.'s with

$$\mathbb{P}(\{X_1=+1\}) = p = 1 - \mathbb{P}(\{X_1=-1\}) \quad 0 < p < 1$$

$(S_n, n \in \mathbb{N})$ is a $\left. \begin{array}{l} \text{sub-} \\ \text{super-} \end{array} \right\}$ martingale if $\left\{ \begin{array}{l} p \geq \frac{1}{2} \\ p \leq \frac{1}{2} \end{array} \right.$

Pf: $\mathbb{E}(S_{n+1} | \mathcal{F}_n) = \mathbb{E}(S_n | \mathcal{F}_n) + \mathbb{E}(X_{n+1} | \mathcal{F}_n) = S_n + \mathbb{E}(X_{n+1})$

Property: If M is a square-integrable martingale $\left\{ \begin{array}{l} \geq 0 \\ \leq 0 \end{array} \right. \checkmark$

& $\varphi: \mathbb{R} \rightarrow \mathbb{R}$ is a convex function st $\mathbb{E}(\varphi(M_n)^2) < +\infty$

Then $(\varphi(M_n), n \in \mathbb{N})$ is a submartingale. $\forall n \in \mathbb{N}$

Pf: $\mathbb{E}(\varphi(M_{n+1}) | \mathcal{F}_n) \underset{\text{Jensen}}{\geq} \varphi(\mathbb{E}(M_{n+1} | \mathcal{F}_n)) = \varphi(M_n) \checkmark$

Stopping times & optional stopping theorem

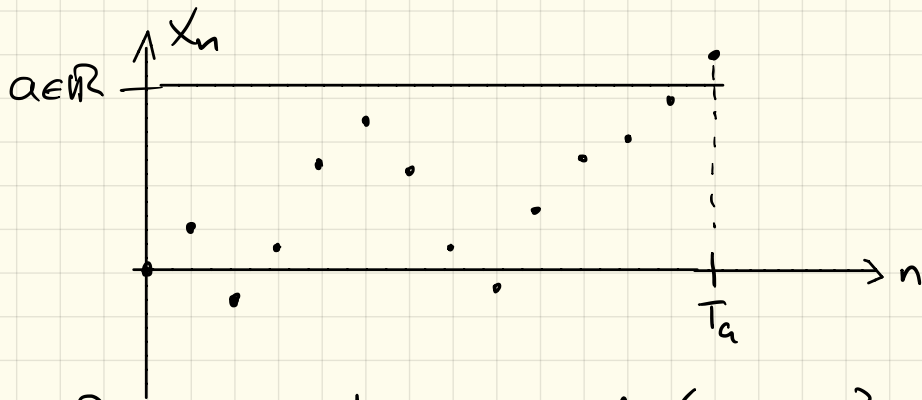
Def: A random time is a random variable $T: \Omega \rightarrow \mathbb{N}$

Let $(\mathcal{F}_n, n \in \mathbb{N})$ be a filtration on $(\Omega, \mathcal{F}, \mathbb{P})$.

Def: A stopping time w.r.t. $(\mathcal{F}_n, n \in \mathbb{N})$ is a random time such that $\{\omega \in \Omega : T(\omega) = n\} \in \mathcal{F}_n \quad \forall n \in \mathbb{N}$

Ex: Let $(X_n, n \in \mathbb{N})$ be a discrete-time stochastic process
 $\mathcal{F}_n = \sigma(X_0, X_1, \dots, X_n) \quad n \in \mathbb{N}$

Then $T_a = \inf \{n \in \mathbb{N} : X_n \geq a\}$ is a stopping time
 $a \in \mathbb{R}$ w.r.t. $(\mathcal{F}_n, n \in \mathbb{N})$



Proof: To be checked: $\{T_a = n\} \in \mathcal{F}_n \quad \forall n \in \mathbb{N}$

$$\{T_a = n\} = \{X_k < a \quad \forall 0 \leq k \leq n-1 \text{ \& } X_n \geq a\}$$

$$= \left(\bigcap_{0 \leq k \leq n-1} \underbrace{\{X_k < a\}}_{\in \mathcal{F}_k \subset \mathcal{F}_n} \right) \cap \underbrace{\{X_n \geq a\}}_{\in \mathcal{F}_n} \in \mathcal{F}_n \quad \checkmark$$

Def: If $(X_n, n \in \mathbb{N})$ is a discrete-time stochastic process and T is a random time, then we define

$$X_T(\omega) := X_{T(\omega)}(\omega) = \sum_{n \in \mathbb{N}} X_n(\omega) 1_{\{T=n\}}(\omega)$$

Def: We say that a random time is bounded if there exists $N \in \mathbb{N}$ such that $T(\omega) \leq N \quad \forall \omega \in \Omega$
(or $\mathbb{P}(\{T \leq N\}) = 1$)

Optional stopping theorem

- Let M be a (square-integrable) martingale w.r.t. a filtration $(\mathcal{F}_n, n \in \mathbb{N})$

Let T be a stopping time w.r.t. $(\mathcal{F}_n, n \in \mathbb{N})$

such that $\exists N \in \mathbb{N}$ with $0 \leq T(\omega) \leq N \quad \forall \omega \in \Omega$

$$\text{Then } \mathbb{E}(M_0) = \mathbb{E}(M_T) = \mathbb{E}(M_N)$$

$\uparrow \qquad \qquad \qquad \uparrow$

- If M is a (square-integrable) submartingale,

$$\text{then } \left(\mathbb{E}(M_0) \leq \right) \mathbb{E}(M_T) \leq \mathbb{E}(M_N)$$

not proven here \uparrow

Proof:

As $0 \leq T(\omega) \leq N \quad \forall \omega \in \Omega$, we have

$$M_T(\omega) = M_{T(\omega)}(\omega) = \sum_{n=0}^N M_n(\omega) \cdot 1_{\{T=n\}}(\omega)$$

$$\mathbb{E}(M_T) = \sum_{n=0}^N \mathbb{E}(M_n \cdot 1_{\{T=n\}}) = \sum_{n=0}^N \mathbb{E}(\mathbb{E}(M_n | \mathcal{F}_n) \cdot 1_{\{T=n\}})$$

(in case M is a submartingale)

$$= \sum_{n=0}^N \mathbb{E}(\mathbb{E}(M_n \cdot 1_{\{T=n\}} | \mathcal{F}_n)) = \sum_{n=0}^N \mathbb{E}(M_n \cdot 1_{\{T=n\}})$$

$$= \mathbb{E}(M_N \cdot \underbrace{\sum_{n=0}^N 1_{\{T=n\}}}_{=1}) = \mathbb{E}(M_N) = \mathbb{E}(M_0)$$

(M = martingale)

\neq

$\in \mathcal{F}_n$

When does the optional stopping theorem fail?

- If T is not a stopping time:

Ex: $T = \text{value of } n \in \{0, \dots, N\}$ such that $M_n \geq M_N \forall n \in \{0, \dots, N\}$

Then both $\mathbb{E}(M_T) > \mathbb{E}(M_0)$ & $\mathbb{E}(M_T) > \mathbb{E}(M_N)$

- If T is not bounded:

Ex: Consider $(X_n, n \geq 1)$ be a sequence of iid r.v.'s

s.t. $\mathbb{P}(\xi X_1 = +1) = \mathbb{P}(\xi X_1 = -1) = \frac{1}{2}$; $M_0 = 0$, $M_n = X_1 + \dots + X_n, n \geq 1$

$M = \text{martingale}$, $T_a = \inf \{n \in \mathbb{N} : M_n \geq a\}$ $a \in \mathbb{N}^*$

Then $\mathbb{E}(M_{T_a}) = a > 0 = \mathbb{E}(M_0)$.

$\underbrace{\hspace{1.5cm}}_{= a}$

Why is it an interesting theorem?

Consider again the previous example with $\Pi =$ simple symmetric random walk: $\Pi_n = X_1 + \dots + X_n$.

$X_i =$ can losses $\begin{matrix} \rightarrow +1 \text{ fr.} \\ \rightarrow -1 \text{ fr.} \end{matrix} \Rightarrow \Pi_n =$ accumulated gain at time n

$$\forall N \geq 1, \mathbb{E}(\Pi_N) = \mathbb{E}(\Pi_0) = 0$$

Let $T = \min\{n \in \mathbb{N} : \Pi_n \geq 10\}, N\}$

$$\mathbb{E}(\Pi_T) = \mathbb{E}(\Pi_0)$$

$\begin{matrix} \dots \\ \hookrightarrow \end{matrix} \begin{cases} = +10 & \text{with high probability} \\ = -x & \text{with tiny probability (when } T=N) \end{cases}$