

Probability Theory

chapter 4. Solution

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$$\text{III. } N: \Omega \rightarrow \mathbb{N} \cup \{\infty\}$$

(1) When $\{\omega \in \Omega \mid N(\omega) = n\} \in \mathcal{F}_n$ for all $n=1,2,3,\dots$

then N is called a stopping time

Here $\{\mathcal{F}_n\}_{n \geq 1}$ are sub- σ -algebras of $\mathcal{F}$ defined

(($\Omega, \mathcal{F}, P$) ... Probability Space)

$$\text{as } \mathcal{F}_n \stackrel{\text{def}}{=} \sigma[X_1, X_2, \dots, X_n] (= \sigma[\bigcup_{j=1}^n \sigma[X_j]])$$

$$(\sigma[X_j] = \{X_j^{-1}(B) \mid B \in \mathcal{B}(\mathbb{R}^1)\})$$

(2) We define $N(\omega) = \inf \{n \mid S_n \in A\}$

then $N(\omega)$ is a stopping time

$$\begin{aligned} (\text{proof}) \quad \{N(\omega) = n\} &= \underbrace{\{S_1 \notin A\}}_{\in \sigma[X_1]} \cap \underbrace{\{S_2 \notin A\}}_{\in \sigma[X_1, X_2]} \cdots \cap \underbrace{\{S_n \in A\}}_{\in \sigma[X_1, \dots, X_n]} \\ &\in \sigma[X_1, X_2, \dots, X_n] = \mathcal{F}_n \end{aligned}$$

So $N(\omega)$ is a stopping time

2 We examine if N_t is a stopping time or not.

$$\{N_t = n\} = \underbrace{\{S_1 \leq t\}}_{\mathcal{G}[X_1]} \cap \dots \cap \underbrace{\{S_n \leq t\}}_{\mathcal{G}[X_1, \dots, X_n]} \cap \underbrace{\{S_{n+1} > t\}}_{\mathcal{G}[X_1, \dots, X_{n+1}]}$$

$$\text{So } \{N_{t+1} = n\} = \{N_t = n-1\} \in \mathcal{G}[X_1, \dots, X_n]$$

Thus N_{t+1} is a stopping time

[3] First we confirm the definition of $\mathcal{F}_N$ (N : stopping time)

$$\mathcal{F}_N = \{A \in \mathcal{F} \mid \forall n=1,2,3,\dots A \cap \{N=n\} \in \mathcal{F}_n\}.$$

Let k be a positive integer.

Let $A \in \mathcal{F}_N$.

Let $\{B_1, B_2, \dots, B_k\}$ be Borel sets.

$$\begin{aligned} & P\left(\bigcap_{j=1}^k \{B_j \in X_{N+j}\} \cap A \cap \{N < \infty\}\right), \quad \{N < \infty\} \\ &= \sum_{n=1}^{\infty} P\left(\bigcap_{j=1}^k \{B_j \in X_{n+j}\} \cap A \cap \{N=n\}\right) = \sum_{n=1}^{\infty} P(\underbrace{\bigcap_{j=1}^k \{B_j \in X_{n+j}\}}_{\sigma[X_{n+1}, \dots, X_{n+k}]} \cap \underbrace{A \cap \{N=n\}}_{\in \mathcal{F}_n}) \end{aligned}$$

Since $A \in \mathcal{F}_N$,
hence $A \cap \{N=n\} \in \mathcal{F}_n$

$$= \sum_{n=1}^{\infty} P\left(\underbrace{\bigcap_{j=1}^k \{B_j \in X_{n+j}\}}_{\sigma[X_{n+1}, \dots, X_{n+k}]} \cap \underbrace{A \cap \{N=n\}}_{\in \mathcal{F}_n}\right)$$

These two events are independent

$$= \sum_{n=1}^{\infty} P\left(\bigcap_{j=1}^k \{B_j \in X_{n+j}\}\right) P(A \cap \{N=n\})$$

$$= \sum_{n=1}^{\infty} \left(\prod_{j=1}^k F(B_j)\right) P(A \cap \{N=n\}) = \prod_{j=1}^k F(B_j) \cdot P(A \cap \{N < \infty\})$$

(=: iid) ($X_1, X_2, \dots$ iid F)

$$\begin{aligned} \therefore & P\left(\bigcap_{j=1}^K \{B_j \in X_{N_j}\} \cap A \cap \{N < \infty\}\right) \\ &= \prod_{j=1}^K \underbrace{P(B_j \in X_j)}_{= F(B_j)} \cdot P(A \cap \{N < \infty\}) \\ &\quad \text{C. 114)} \end{aligned}$$

This is what the statement means. ■

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[Step] Consider $X_n \geq 0$ (non-negative)

Since $E[N] < \infty$, $N < \infty$ (a.s.).

$$E[S_N] = E[S_N \cdot \mathbb{I}_{\{N < \infty\}}]$$

$$= \int_{\Omega} S_N \cdot \mathbb{I}_{\{N < \infty\}} dP$$

$$= \int_{\Omega} \sum_{n=1}^{\infty} S_n \cdot \mathbb{I}_{\{N=n\}} dP \quad \downarrow \quad N=n$$

$$= \int_{\Omega} \sum_{n=1}^{\infty} S_n \cdot \mathbb{I}_{\{N=n\}} dP \quad \downarrow \quad \text{Monotone Convergence Theorem}$$

$$= \sum_{n=1}^{\infty} \int_{\Omega} S_n \cdot \mathbb{I}_{\{N=n\}} dP$$

$$= \sum_{n=1}^{\infty} \sum_{m=1}^n \int_{\Omega} \underbrace{X_m \cdot \mathbb{I}_{\{N=n\}}}_{\geq 0} dP$$

Since $\forall n, m \int_{\Omega} X_m \cdot \mathbb{I}_{\{N=n\}} dP$

we may swap $\sum_{n=1}^{\infty} \sum_{m=1}^n$

$$= \sum_{m=1}^{\infty} \sum_{n=m}^{\infty} \int_{\Omega} X_m \cdot \mathbb{I}_{\{N=n\}} dP$$

$$= \sum_{m=1}^{\infty} \int_{\Omega} \underbrace{X_m \cdot \mathbb{I}_{\{N \geq m\}}}_{\geq 0} dP$$

They are independent

$$\{N \geq m\} = \{N < m\}^c$$

$$= \sum_{m=1}^{\infty} \int_{\Omega} X_m \cdot \mathbb{I}_{\{N \geq m\}} dP \quad (\because \{N < m\}^c = \{N \geq m\})$$

$$= \sum_{m=1}^{\infty} \underbrace{E[X_m]}_{\in \sigma(X_1, \dots, X_{m-1})} \cdot P(N \geq m)$$

$$(\underbrace{= E[X_1]}_{\text{if } m=1})$$

$$= E[X_1] \sum_{m=1}^{\infty} P(N \geq m)$$

$$= E[X_1] \cdot E[N] \quad (\text{See chapter 2, [35]})$$

Now the proof is complete. $\blacksquare$

Step 2

$$S_N = X_1 + \dots + X_N = (X_1^+ - X_1^-) + \dots + (X_N^+ - X_N^-)$$

$$\text{Let } S_N^{(+)} \stackrel{\text{def}}{=} X_1^+ + \dots + X_N^+$$

$$S_N^{(-)} \stackrel{\text{def}}{=} X_1^- + \dots + X_N^-$$

$$\text{Then } E[S_N^{(+)}] = E[X_1^+] E[N] < \infty \quad (S_N^{(+)} < \infty \text{ (a.s.)})$$

$$E[S_N^{(-)}] = E[X_1^-] E[N] < \infty \quad (S_N^{(-)} < \infty \text{ (a.s.)})$$

$$\therefore \underbrace{E[S_N^{(+)} - S_N^{(-)}]}_{\text{This can be defined as.}} = E[X_1] E[N]$$

$\blacksquare$ This can be defined as.

$$= E[S_N]$$

[5]

① $\min\{S, T\}$ is a stopping time

$$\{\min\{S, T\} = n\} = \{S = n\} \cap \underbrace{\{T \geq n\}}_{\substack{\parallel \\ \{T < n\}^c}} \cup \{T = n\} \cap \underbrace{\{S \geq n\}}_{\substack{\parallel \\ \{S < n\}^c}}$$

$$(\in \mathcal{G}[X_1 \dots X_{n-1}] \subseteq \mathcal{G}[X_1 \dots X_n])$$

$$\in \mathcal{G}[X_1, X_2, \dots, X_n]$$

Hence $\min\{S, T\}$ is a stopping time.

② $\max\{S, T\}$ is a stopping time

$$\{\max\{S, T\} = n\} = \{S = n\} \cap \{T \leq n\} \cup \{T = n\} \cap \{S \leq n\}$$

$$\in \mathcal{G}[X_1 \dots X_n]$$

So it is a stopping time.

③. Let $T = n$. Then $\{T = k\} = \emptyset$, $\Omega \in \mathcal{G}[X_1 \dots X_k]$
 $(k \neq n) \quad (k < n) \quad (\because \mathcal{G}\text{-algebra})$

So T is a stopping time.

By ① and ② the proof is complete.

[6] The statement is true.

$$\begin{aligned}\{S+T=n\} &= \sum_{k=1}^n \{S=k \cap T=n-k\} \\ &= \sum_{k=1}^n \underbrace{\{S=k\}}_{\in \mathcal{G}[X_1 \dots X_k]} \cap \underbrace{\{T=n-k\}}_{\in \mathcal{G}[X_1 \dots X_{n-k}]} \\ &\in \mathcal{G}[X_1 \dots X_n]\end{aligned}$$

So $S+T$ is a stopping time.

□ $M \leq N$ is a stopping time then $\mathcal{F}_M \subseteq \mathcal{F}_N$.

We show that $\forall A \in \mathcal{F}_M \Rightarrow A \in \mathcal{F}_N$.

(proof) $A \in \mathcal{F}_M$

$$A \cap \{N=n\} = A \cap \{N=n\} \cap \{M \leq n\}$$

$$(\because M \leq N)$$

$$= \sum_{m=1}^n A \cap \{N=n\} \cap \{M=m\}$$

$$= \{N=n\} \cap \left(\sum_{m=1}^n A \cap \{M=m\} \right)$$

$$\in \mathcal{F}_M \quad (\because A \in \mathcal{F}_M)$$

$$= \sigma[X_1, \dots, X_n]$$

$$\in \sigma[X_1, \dots, X_n]$$

$$A \cap \{N=n\} \in \sigma[X_1, \dots, X_n]$$

Now the proof is complete. ■

8] $L \leq M$. $A \in \mathcal{F}_L$ (L.M. Stopping time)

$$N \stackrel{\text{def}}{=} \begin{cases} L & (w \in A) \\ M & (w \notin A) \end{cases}$$

Then N is a stopping time

(proof)

$$\{N=h\} = \underbrace{\{N=h\} \cap A}_{\in \mathcal{F}_h} \cup \underbrace{\{N=h\} \cap A^c}_{\in \mathcal{F}_h}$$

$$= \underbrace{\{L=h\} \cap A}_{\in \mathcal{F}_h} \cup \underbrace{\{M=h\} \cap A^c}_{\in \mathcal{F}_h}$$

($\because A \in \mathcal{F}_L$)

$$\underbrace{\{M=h\}}_{\in \mathcal{F}_h} \mid \underbrace{\{M=h\} \cap A}_{\in \mathcal{F}_h}$$

$$\otimes \{M=h\} \cap A = \{L \leq h\} \cap \{M=h\} \cap A$$

$$= \sum_{k=1}^h \{L=k\} \cap \{M=h\} \cap A \quad (\because L \leq M)$$

$$= \underbrace{\{M=h\}}_{\in \mathcal{F}_h} \cap \underbrace{\sum_{k=1}^h \{L=k\} \cap A}_{\in \mathcal{F}_h}$$

$\otimes \in \mathcal{F}_h$

[9]

(1) Definition of $E[X|\mathcal{F}]$.

$E[X|\mathcal{F}]$ is a $\mathcal{F}$ -measurable function

such that $\forall A \in \mathcal{F} (\mathcal{F} \subseteq \mathcal{F}_0)$

$$\int_A X dP = \int_A E[X|\mathcal{F}] dP$$

(In a sense of a.s. $E[X|\mathcal{F}]$ is unique)

$$(2) \int_{\Omega} E[X|\mathcal{F}] dP = \int_{\Omega} X dP$$

||

$$\int_{\Omega} Y dP$$

$$\therefore \left| \int_{\Omega} Y dP \right| = \left| \int_{\Omega} X dP \right| \leq \int_{\Omega} |X| dP < \infty$$

Here Y is integrable

(3) We show that $Y \geq Y'$ (a.s.) and $Y \leq Y'$ (a.s.)

① Let $A = \{Y < Y'\}$...

$A \in \mathcal{F}$ ($\because Y, Y'$ are $\mathcal{F}$ -measurable)

$$\int_A Y dP = \int_A X dP \in (-\infty, \infty) \quad (\because X: \text{integrable})$$

$$\int_A Y' dP = \int_A X dP \in (-\infty, \infty)$$

$$\text{Hence } \int_A Y dP - \int_A Y' dP = 0$$

$$\therefore \underbrace{\int_A (Y' - Y) dP}_{> 0} = 0$$

$$\text{So } P(A) = 0 \quad \therefore Y \geq Y' \text{ (a.s.)}$$

② Let $A = \{Y' < Y\}$

In the same way, we have $Y' \geq Y$ (a.s.)

[10] Let $B_1 = \{E[X_1|\mathcal{F}] > E[X_2|\mathcal{F}]\} \cap B$, $B_1 \in \mathcal{F}$

$$\text{Then } \int_{B_1} X_1 dP = \int_{B_1} E[X_1|\mathcal{F}] dP \in (-\infty, \infty)$$

$$\int_{B_1} X_2 dP = \int_{B_1} E[X_2|\mathcal{F}] dP \in (-\infty, \infty)$$

Since $X_1 = X_2$ (a.s.) on B .

(We suppose X_1, X_2 are integrable)

$$\int_{B_1} E[X_1|\mathcal{F}] dP = \int_{B_1} E[X_2|\mathcal{F}] dP$$

$$\Rightarrow \int_{B_1} \underbrace{(E[X_1|\mathcal{F}] - E[X_2|\mathcal{F}])}_{>0} dP = 0$$

$$\therefore P(B_1) = 0$$

$$\Rightarrow E[X_1|\mathcal{F}] \leq E[X_2|\mathcal{F}] \text{ (a.s.) on } B.$$

$$, B_2 = \{E[X_1|\mathcal{F}] < E[X_2|\mathcal{F}]\} \cap B.$$

Similarly we have $E[X_1|\mathcal{F}] \geq E[X_2|\mathcal{F}]$ (a.s.) on B .

(1) $(\Omega, \mathcal{F}_0)$: measurable space

$$(1) \forall A \in \mathcal{F}_0 \quad \mu(A) = 0 \Rightarrow \nu(A) = 0$$

Then $\nu \ll \mu$.

We call ν is absolutely continuous with regard to μ .

(2) Let μ, ν be σ -finite signed measures on $(\Omega, \mathcal{F}_0)$.

and suppose $\nu \ll \mu$.

Then there exists a $\mathcal{F}$ -measurable function f

such that $\nu(A) = \int_A f d\mu$ for all $A \in \mathcal{F}$.

(3) Let $\nu(A) \stackrel{\text{def}}{=} \int_A X d\mu$. (X : integrable)

$(\nu: \mathcal{F}_0 \rightarrow [0, \infty])$

Then $\nu(A)$ is a signed-measure on $(\Omega, \mathcal{F}_0)$ and finite, hence G -finite. ($\because X$ is integrable)

$$\text{And } \forall A \in \mathcal{F}_0 \quad P(A)=0, \quad \int_A X dP = 0$$

$$\Rightarrow \nu(A)=0 \quad \therefore \nu \ll P \text{ on } \mathcal{F}_0.$$

$$\Rightarrow \nu \ll P \text{ on } \mathcal{F}. \quad (\mathcal{F} \subseteq \mathcal{F}_0)$$

By 12) there exists a $\mathcal{F}$ -measurable function f

$$\text{st } \forall A \in \mathcal{F} \quad \nu(A) = \int_A f dP. \quad (\text{hence } = \int_A X dP)$$

Such a function f is the desired function.

$$\text{We denote } f \text{ as } E[X|\mathcal{F}]$$

[2] Let $A \in \mathcal{F}$. By the definition,

$$\int_A X dP = \int_A E[X|\mathcal{F}] dP \quad \text{for all } A \in \mathcal{F}.$$

$$\textcircled{1} A = \{X > E[X|\mathcal{F}]\} \in \mathcal{F}$$

$$\int_A X dP = \int_A E[X|\mathcal{F}] dP \in (-\infty, \infty)$$

$$\Rightarrow \underbrace{\int_A (X - E[X|\mathcal{F}]) dP}_{> 0} = 0$$

$$\Rightarrow P(A) = 0 \quad \therefore X \leq E[X|\mathcal{F}] \quad (a)$$

$$\textcircled{2} A = \{X < E[X|\mathcal{F}]\} \in \mathcal{F}$$

$$\text{Similarly we have } X \geq E[X|\mathcal{F}] \quad (a)$$

$$\therefore X = E[X|\mathcal{F}] \quad (a)$$

13 Let $A \in \mathcal{F}_1$

$$\begin{aligned}\int_A E[X|\mathcal{F}] dP &= \int_A X dP \\ &= \int_{\Omega} X \cdot \mathbb{I}_A dP \\ &= E[X \cdot \mathbb{I}_A] = E[X] P(A) \\ &= \int_A E[X] dP\end{aligned}$$

$E[X]$ is a constant $\Rightarrow \mathcal{F}$ -measurable

$$\text{And } \forall A \in \mathcal{F} \quad \int_A X dP = \int_A E[X] dP$$

$$\text{So } E[X|\mathcal{F}] = E[X] \quad (\text{a.s.})$$

by the definition of conditional expectation

($\phi(\cdot, \cdot) : \mathbb{R}^2 \rightarrow \mathbb{R}$ is a Borel measurable function)

$$\boxed{14} \quad E[\tilde{z}|X] \stackrel{\text{def}}{=} E[\tilde{z} | \sigma(X)]$$

$$(\sigma(X) = \{X^{-1}(B) \mid B \in \mathcal{B}(\mathbb{R})\})$$

(proof)

Let $A \in \sigma(X)$. ($\therefore \exists B \in \mathcal{B}(\mathbb{R})$ st $A = X^{-1}(B)$)

$$\int_A \phi(X, Y) dP = \int_A E[\phi(X, Y) | X] dP.$$

||

$$\int_{\Omega} \mathbb{I}_A \phi(X, Y) dP = \int_{\Omega} \mathbb{I}_{\{X(\omega) \in B\}} \phi(X(\omega), Y(\omega)) dP$$

Now by change variable formula,

$$\int_{\Omega} \mathbb{I}_{\{X \in B\}} \phi(X, Y) dP$$

$$\equiv \int_{\mathbb{R}^2} \mathbb{I}_B(x) \phi(x, y) d\mu dx d\nu$$

$$(X \sim \mu, Y \sim \nu)$$

Since X, Y are independent,

by Fubini's theorem

$$\begin{aligned}
& \int_{\mathbb{R}} \int_{\mathbb{R}} \mathbb{I}_B(x) \phi(x, y) \, d\mu \, d\nu \\
&= \int_B \int_{\mathbb{R}} \underbrace{\phi(x, y)}_{\parallel} \, d\nu \, d\mu \\
&\quad \parallel \quad \downarrow \text{change variable formula} \\
&\quad \mathbb{E}[\phi(X, Y)] \\
&= \int_B \mathbb{E}[\phi(X, Y)] \, d\mu \\
&= \int_B g(x) \, d\mu = \int_{\mathbb{R}} \mathbb{I}_B(x) g(x) \, d\mu \quad \downarrow \text{change variable formula} \\
&= \int_{\Omega} \mathbb{I}_A(\omega) g(X) \, dP \\
&= \int_A g(X) \, dP.
\end{aligned}$$

$$\begin{aligned}
\text{So } \int_A \phi(X, Y) \, dP &= \int_A g(X) \, dP \text{ for all } A \in \mathcal{G}(X) \\
& (= \int_A \mathbb{E}[\phi(X, Y) | Y] \, dP)
\end{aligned}$$

And $g(X) : \Omega \rightarrow \mathbb{R}$ is $\mathcal{G}(X)$ -measurable. $\dots \otimes$

$$\text{So } \mathbb{E}[\phi(X, Y) | Y] = g(X).$$

$$\begin{aligned}
& (\otimes) \dots g(X) \stackrel{\text{def}}{=} \int_{\Omega} \phi(X, Y) \, dP = \int_{\mathbb{R}} \phi(X, y) \, \nu(dy) \\
& \text{is a Borel-measurable function. See ch1, [35]} \\
& g : \mathbb{R} \rightarrow \mathbb{R} \Rightarrow g(X) \text{ is } \mathcal{G}(X)\text{-measurable.}
\end{aligned}$$

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(1) Let $a, b \in \mathbb{R}$ and let $A \in \mathcal{F}$.

$$\int_A (aX + bY) dP = \int_A E[aX + bY | \mathcal{F}] dP$$

(by definition)

Since X, Y are integrable,

$$\int_A (aX + bY) dP = a \int_A X dP + b \int_A Y dP$$

And by the definition of conditional expectation

$$= a \int_A E[X | \mathcal{F}] dP + b \int_A E[Y | \mathcal{F}] dP$$

$$= \int_A (a E[X | \mathcal{F}] + b E[Y | \mathcal{F}]) dP$$

$$\therefore \int_A E[aX + bY | \mathcal{F}] dP = \int_A (a E[X | \mathcal{F}] + b E[Y | \mathcal{F}]) dP$$

And $a E[X | \mathcal{F}] + b E[Y | \mathcal{F}]$ is also $\mathcal{F}$ -measurable

Therefore the statement is true

(2) Suppose $X \leq Y$ and let $A \in \mathcal{F}$.

$$\text{Then } \int_A X \, dP \leq \int_A Y \, dP.$$

By definition we have

$$\int_A E[X|\mathcal{F}] \, dP \leq \int_A E[Y|\mathcal{F}] \, dP$$

for all $A \in \mathcal{F}$.

Hence we have $E[X|\mathcal{F}] \leq E[Y|\mathcal{F}]$ (a.s.)

(You may put $A = \{E[Y|\mathcal{F}] < E[X|\mathcal{F}]\}$)

$$(3) \quad \int_A X_n \, dP = \int_A E[X_n|\mathcal{F}] \, dP \quad (\text{by definition})$$

($A \in \mathcal{F}$)

By monotone convergence theorem ($\rightarrow$)

$$\lim_{n \rightarrow \infty} \int_A X_n \, dP \xrightarrow{\text{MCT}} \int_A X \, dP = \int_A E[X|\mathcal{F}] \, dP \quad (\text{definition})$$

$\circledast$ II

$$\lim_{n \rightarrow \infty} \int_A E[X_n|\mathcal{F}] \, dP \xrightarrow{\text{MCT}} \int_A \lim_{n \rightarrow \infty} E[X_n|\mathcal{F}] \, dP$$

$$\text{So we have } \int_A \lim_{n \rightarrow \infty} E[X_n|\mathcal{F}] \, dP = \int_A E[X|\mathcal{F}] \, dP.$$

$$\text{By (2)} \quad E[X_n | \mathcal{F}] \leq E[X_{n+1} | \mathcal{F}] \quad (\text{a.s.})$$

Thus $\lim_{n \rightarrow \infty} E[X_n | \mathcal{F}]$ is defined almost surely.

($\because$ monotone-increasing)

$\lim_{n \rightarrow \infty} E[X_n | \mathcal{F}]$ is also $\mathcal{F}$ -measurable. (See ch1)

$$(\equiv \sup_{n \in \mathbb{N}} E[X_n | \mathcal{F}])$$

$$\text{Therefore} \quad \lim_{n \rightarrow \infty} E[X_n | \mathcal{F}] = E[X | \mathcal{F}] \quad (\text{a.s.})$$

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$$\left(\sup_{(a,b) \in S} \{ax+by\} = \phi(x) \right)$$

Let $S = \{(a,b) \in \mathbb{Q}^2 \mid \phi(x) > ax+by \text{ for all } x\}$

$$\begin{aligned} E[\phi(X) | \mathcal{F}] &\geq E[ax+by | \mathcal{F}] \quad (\forall (a,b) \in S) \\ &= aE[X | \mathcal{F}] + b \quad (as) \end{aligned}$$

$$\sup_{(a,b) \in S} \{aE[X | \mathcal{F}] + b\} = \phi \circ E[X | \mathcal{F}]$$

Since S is a countable set.

$$\text{if } \Omega^{(a,b)} \in \mathcal{F} : P(\Omega^{(a,b)}) = 1$$

$$\text{then } P\left(\bigcap_{(a,b) \in S} \Omega^{(a,b)}\right) = 1 \quad \dots \quad \textcircled{*}$$

$$\text{So } E[\phi(X) | \mathcal{F}] \geq \phi \circ E[X | \mathcal{F}] \quad (as)$$

↓
by $\textcircled{*}$, (as) still holds.

$$17 \quad \text{Let } \phi(x) = |x|^p$$

Then $\phi(x)$ is a convex function.

$$\text{By 1b, } E[|X|^p | \mathcal{F}] \geq |E[X | \mathcal{F}]|^p$$

$$\text{So } E \circ E[|X|^p | \mathcal{F}] \geq E[|E[X | \mathcal{F}]|^p]$$

$$\parallel \\ E|X|^p$$

$$\therefore \{E|X|^p\}^{\frac{1}{p}} \geq E[|E[X | \mathcal{F}]|^p]^{\frac{1}{p}}$$

$$\|X\|_p \stackrel{\text{def}}{=} (E|X|^p)^{\frac{1}{p}}$$

$$T(\cdot) \stackrel{\text{def}}{=} E[\cdot | \mathcal{F}]$$

$$\text{Therefore } \underline{\|X\|_p \geq \|T(X)\|_p}$$

So this is the desired result.

18 Let $A \in \mathcal{F}$.

$$\int_A X dP = \int_A E[X|\mathcal{F}] dP \quad (\text{by definition})$$

Since $A \in \mathcal{G} (\because \mathcal{F} \subset \mathcal{G})$

$$\int_A X dP = \int_A E[X|\mathcal{G}] dP.$$

$$\begin{aligned} \text{So } \forall A \in \mathcal{F} \quad \int_A E[X|\mathcal{F}] dP \\ = \int_A E[X|\mathcal{G}] dP \end{aligned}$$

And $E[X|\mathcal{G}]$ is also $\mathcal{F}$ -measurable

Hence $E[X|\mathcal{F}] = E[X|\mathcal{G}]$ (as) by uniqueness.

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(1) Let $X_1 = E[X | \mathcal{F}_1]$

X_1 is $\mathcal{F}_1$ -measurable $\Rightarrow \mathcal{F}_2$ -measurable

By [2], $E[X_1 | \mathcal{F}_2] = X_1$ (a.s.)

$\therefore E[E[X | \mathcal{F}_1] | \mathcal{F}_2] = E[X | \mathcal{F}_1]$

(2) Let $A \in \mathcal{F}_1$

Let $X_2 = E[X | \mathcal{F}_2]$

Then $\int_A X_2 dP = \int_A E[X_2 | \mathcal{F}_1] dP$

LHS = $\int_A E[X | \mathcal{F}_2] dP$
 $\downarrow (\because A \in \mathcal{F}_1 \Rightarrow A \in \mathcal{F}_2)$
 $= \int_A X dP$

And $\int_A X dP = \int_A E[X | \mathcal{F}_1] dP$

$\therefore$ So for all $A \in \mathcal{F}_1$ $\int_A E[X_2 | \mathcal{F}_1] dP = \int_A E[X | \mathcal{F}_1] dP$

$\therefore E[E[X | \mathcal{F}_2] | \mathcal{F}_1] = E[X | \mathcal{F}_1]$ (a.s.)

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[Step 1] $X = \mathbb{I}_E(\omega)$ ($E \in \mathcal{F}$) and $A \in \mathcal{F}$.

$$\begin{aligned}
 \int_A XY \, dP &= \int_A \mathbb{I}_E(\omega) \cdot Y \, dP \\
 &= \int_{A \cap E} Y \, dP \quad (A \cap E \in \mathcal{F}) \\
 &= \int_{A \cap E} E[Y|\mathcal{F}] \, dP \\
 &= \int_A \mathbb{I}_E \cdot E[Y|\mathcal{F}] \, dP \\
 &= \int_A \underbrace{X E[Y|\mathcal{F}]} \, dP
 \end{aligned}$$

↳ This is also $\mathcal{F}$ -measurable

Hence,

$$E[XY|\mathcal{F}] = X \cdot E[Y|\mathcal{F}]$$

[Step 2] $X = a_1 \mathbb{I}_{E_1}(\omega) + a_2 \mathbb{I}_{E_2}(\omega) + \dots + a_k \mathbb{I}_{E_k}(\omega)$

$\{a_j\}_{j=1}^k \subseteq [0, \infty)$, $\{E_j\}_{j=1}^k \subseteq \mathcal{F}$: disjoint

By [15] (1): Linearity and [Step 1]

We have $E[XY|\mathcal{F}] = X \cdot E[Y|\mathcal{F}]$.

Step 3 $X \geq 0$, $Y \geq 0$

We take $\{X_n\}_{n \geq 1}$: simple measurable functions

st $X_n \uparrow X$.

For each $n \geq 1$, $E[X_n Y | \mathcal{F}] = X_n E[Y | \mathcal{F}]$

by Step 2.

By [15] (3), $X_n Y \uparrow XY$

$\Rightarrow E[X_n Y | \mathcal{F}] \uparrow E[XY | \mathcal{F}]$

$$\begin{aligned} \therefore \lim_{n \rightarrow \infty} E[X_n Y | \mathcal{F}] &= \lim_{n \rightarrow \infty} X_n E[Y | \mathcal{F}] \\ &= E[XY | \mathcal{F}] \end{aligned}$$

$$\therefore X E[Y | \mathcal{F}] = E[XY | \mathcal{F}]$$

Step 4 $Y \geq 0$.

$$\text{By } \boxed{\text{Step 3}}, \quad E[X^T Y | \mathcal{F}] = X^T E[Y | \mathcal{F}] \quad \dots \textcircled{1}$$

$$E[XY | \mathcal{F}] = X E[Y | \mathcal{F}] \quad \dots \textcircled{2}$$

And $E|XY| < \infty$, implies that

$\textcircled{1} - \textcircled{2}$ is defined (a.s.)

$$\text{So } E[XY | \mathcal{F}] = X E[Y | \mathcal{F}] \quad (\text{a.s.})$$

Step 5 General case,

$$E[XY^T | \mathcal{F}] = X E[Y^T | \mathcal{F}] \quad \dots \textcircled{3}$$

$$E[XX^T | \mathcal{F}] = X E[X^T | \mathcal{F}] \quad \dots \textcircled{4}$$

(by Step 4)

By

$\textcircled{3} - \textcircled{4}$, we have the desired result.

[2] Let Z be a $\mathcal{F}$ -measurable function

We consider $\min_Z E[(X-Z)^2]$.

$$\begin{aligned}
 E[(X-Z)^2] &= E \circ E[(X-Z)^2 | \mathcal{F}] \\
 &= E[E[(X - E[X|\mathcal{F}] + E[X|\mathcal{F}] - Z)^2 | \mathcal{F}]] \\
 &= E \left[\underbrace{E[(X - E[X|\mathcal{F}])^2 | \mathcal{F}]}_{\textcircled{1}} + 2 \underbrace{E[(X - E[X|\mathcal{F}]) \cdot (E[X|\mathcal{F}] - Z) | \mathcal{F}]}_{\textcircled{2}} + \underbrace{E[(E[X|\mathcal{F}] - Z)^2 | \mathcal{F}]}_{\textcircled{3}} \right]
 \end{aligned}$$

$$\textcircled{1} = V[X|\mathcal{F}] \ (\geq 0)$$

$$\begin{aligned}
 \textcircled{2} &= 2 \cdot (E[X|\mathcal{F}] - Z) \underbrace{E[(X - E[X|\mathcal{F}]) | \mathcal{F}]}_{=0} \\
 &= 0
 \end{aligned}$$

$$= 0 \quad (\because \text{see [2]})$$

① is not related with Z .

So we just have to consider

$$\underset{Z}{\operatorname{Argmin}} E[\textcircled{3}]$$

$$\text{And } \textcircled{3} = (E[X|F] - Z)^2 \quad (\because \textcircled{2})$$

$$\geq 0$$

$$\text{So } E[\textcircled{3}] \geq 0$$

Equality holds when $\textcircled{3} = 0$ (as)

$\therefore Z = E[X|F]$ (as) minimizes it.

$$\boxed{22} \quad \mu(w, A) : (w \in \Omega, A \in \mathcal{S})$$

- For almost everywhere $w \in \Omega$

$\mu(w, \cdot) : \mathcal{S} \rightarrow [0, 1]$ is a measure

(i.e. $\exists \tilde{\Omega} \in \mathcal{F} \quad P(\tilde{\Omega}) = 1$,

$\forall w \in \tilde{\Omega} : \mu(w, \cdot) : \mathcal{S} \rightarrow [0, 1]$ is a measure)

- For all $A \in \mathcal{S}$

$\mu(\cdot, A) : \Omega \rightarrow [0, 1]$ is a conditional

probability = $P(X \in A | \mathcal{F})$.

Then $\mu(w, A)$ is a regular conditional distribution

for X given $\mathcal{F}$.

23

(1) When $\mathcal{F}_n \subseteq \mathcal{F}_{n+1}$ and $\{\mathcal{F}_n\}_{n \geq 1}$ is σ -algebras

then $\mathcal{F}_n$ is a filtration

(2) X_n is $\mathcal{F}_n$ -measurable for all $n \geq 1$

(3) $E|X_n| < \infty$

- X_n is $\mathcal{F}_n$ -measurable (adapted to $\mathcal{F}_n$)

- $E[X_{n+1} | \mathcal{F}_n] = X_n$

(for all $n \geq 1$)

(4) $E|X_n| < \infty$, X_n is $\mathcal{F}_n$ -measurable

$E[X_{n+1} | \mathcal{F}_n] \leq X_n$

(5) $E|X_n| < \infty$ X_n is $\mathcal{F}_n$ -measurable

$E[X_{n+1} | \mathcal{F}_n] \geq X_n$

24

$$\textcircled{1} E|S_n| \leq \sum_{j=1}^n E|\xi_j| \leq 2n E[\xi_j^+] < \infty$$

$$(\because E[\xi_j] = 0 \Rightarrow E[\xi_j^+] = E[\xi_j^-] < \infty)$$

$$\textcircled{2} S_n = \xi_1 + \xi_2 + \dots + \xi_n \\ = g(\xi_1, \xi_2, \dots, \xi_n)$$

$$\text{where } g(x_1, x_2, \dots, x_n) = x_1 + x_2 + \dots + x_n$$

Since $g: \mathbb{R}^n \rightarrow \mathbb{R}$ is continuous,

$g(\cdot)$ is Bore measurable

Let us recall ch 2 [9],

From its proof, we found that

$$g(\xi_1, \xi_2, \dots, \xi_n) \text{ is } \sigma\left[\bigcup_{j=1}^n \sigma[\xi_j]\right] - \text{measurable}$$

$$\parallel$$

$$\sigma[\xi_1, \xi_2, \dots, \xi_n]$$

$$\parallel$$

$$\mathbb{F}_n$$

③ Finally, $E[S_{n+1} | \mathcal{F}_n]$

$$= E[S_{n+1} + S_n | \mathcal{F}_n] = S_n + E[S_{n+1} | \mathcal{F}_n]$$

$\nearrow$ $(\because [12])$ $\nearrow$ $(\because [13])$

$$= S_n + E[S_{n+1}] = S_n$$

$$\therefore E[S_{n+1} | \mathcal{F}_n] = S_n$$

($\oplus$ $J_1 \sim J_{n+1}$: independent
 $\Rightarrow \sigma(J_1) \sim \sigma(J_{n+1})$: independent
 $\Rightarrow \sigma(J_1 \dots J_n), \sigma(J_{n+1})$ independent
 $\therefore$ ch 2, [2])

By ①~③, $\{S_n\}_{n \geq 1}$ is a martingale

25

$$① E[S_n^2] = n\sigma^2$$

$$\therefore E|S_n^2 - n\sigma^2| < 2n\sigma^2 < \infty$$

$$② S_n^2 - n\sigma^2 = g(J_1, J_2, \dots, J_n) - n\sigma^2 \quad \text{where}$$

$$g(x_1, x_2, \dots, x_n) = (x_1 + \dots + x_n)^2 \quad \text{is Borel-Measurable}$$

$$\text{So } S_n^2 - n\sigma^2 \text{ is } \sigma(J_1, J_2, \dots, J_n)\text{-measurable}$$

$$③ E[S_{n+1}^2 - (n+1)\sigma^2 | \underline{J_n}]$$

$$= E[(S_n + J_{n+1})^2 - (n+1)\sigma^2 | \underline{J_n}]$$

$$= E[S_n^2 + 2S_n J_{n+1} + J_{n+1}^2 | \underline{J_n}] - (n+1)\sigma^2 \quad \downarrow \text{[5.1]}$$

$$= \underbrace{S_n^2}_{(12)} + \underbrace{2S_n E[J_{n+1} | \underline{J_n}]}_{(20, 13)} + \underbrace{E[J_{n+1}^2]}_0 - (n+1)\sigma^2 \quad (13)$$

$$= S_n^2 + 0 - (n+1)\sigma^2 = S_n^2 - n\sigma^2$$

So the proof is complete.

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$$\textcircled{1} E|M_n| = \sum_{m=1}^n E|Y_m| < \infty$$

$$(\because E[Y_m] = 1 \Rightarrow E[Y_m], E[Y_m] < \infty)$$

$$\textcircled{2} M_n = g(Y_1, Y_2, \dots, Y_n) \quad \text{where } g(x_1, \dots, x_n) = x_1 x_2 \dots x_n$$

(= continuous $\Rightarrow$ Borel-measurable)

$\therefore M_n$ is $\sigma(Y_1, Y_2, \dots, Y_n)$ -measurable. ($\mathcal{F}_n$ -measurable)

$$\textcircled{3} E[M_{n+1} | \mathcal{F}_n] = E[Y_{n+1} \cdot M_n | \mathcal{F}_n] \rightarrow (\textcircled{2})$$

$$= M_n E[Y_{n+1} | \mathcal{F}_n] = M_n \cdot E[Y_{n+1}] = M_n$$

$\uparrow$
(13)

$\therefore$ By $\textcircled{3}$, $\{M_n\}_{n \geq 1}$ is a martingale.

97

Since $\{X_n\}_{n \geq 1}$ is a supermartingale,

$$E[X_n | \mathcal{F}_{n-1}] \leq X_{n-1}$$

By [5] (2),

$$E[E[X_n | \mathcal{F}_{n-1}] | \mathcal{F}_{n-2}] \leq E[X_{n-1} | \mathcal{F}_{n-2}] \leq X_{n-2} \quad \text{Supermartingale}$$

By [19] (2), LHS = $E[X_n | \mathcal{F}_{n-2}]$

So we have $E[X_n | \mathcal{F}_{n-2}] \leq X_{n-2}$

By repeating the same argument $n-m$ times

we have $E[X_n | \mathcal{F}_m] \leq X_m$.

[28] We only show that $E[\phi(X_{n+1}) | \mathcal{F}_n] \geq \phi(X_n)$.

By Jensen's inequality ([16]),

$$E[\phi(X_{n+1}) | \mathcal{F}_n] \geq \phi(\underbrace{E[X_{n+1} | \mathcal{F}_n]}_{\text{martingale}}) = \phi(X_n)$$

($\because \{X_n\}_{n \geq 1}$ martingale)

Now the proof is complete.

(For example, $\phi(x) = |x|^p$)

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By Jensen's inequality

$$E[\phi(X_{n+1}) | \mathcal{F}_n] \geq \phi \circ E[X_{n+1} | \mathcal{F}_n]$$

Since $\phi(\cdot)$ is an increasing function,

$$E[X_{n+1} | \mathcal{F}_n] \geq X_n \quad (\because \text{sub-martingale})$$

$$\text{implies } \phi \circ E[X_{n+1} | \mathcal{F}_n] \geq \phi(X_n)$$

$$\text{So } E[\phi(X_{n+1}) | \mathcal{F}_n] \geq \phi(X_n)$$

$\therefore \phi(X_n)$ is a sub-martingale.

30

(1) $\{H_n\}_{n \geq 1}$ is a predictable sequence

means that H_n is $\mathcal{F}_{n-1}$ -measurable.

$$(2) (H \cdot X)_n \stackrel{\text{def}}{=} \sum_{m=1}^n H_m (X_m - X_{m-1})$$

3] • $X_n \dots$ Supermartingale

• $H_n \geq 0$ and predictable, and bounded

$$\begin{aligned} \textcircled{1} \quad E|(H \cdot X)_n| &\leq \sum_{m=1}^n E|H_m|(X_m - X_{m-1}) \\ &\leq M \sum_{m=1}^n E[|X_m| + |X_{m-1}|] \\ &\quad (|H_n| \leq M < \infty \dots \therefore \text{Bounded}) \\ &\leq 2M(E|X_0| + \dots + E|X_n|) < \infty \end{aligned}$$

($\because \{X_n\}_{n \geq 1}$: supermartingale $\Rightarrow E|X_n| < \infty$)

$$\begin{aligned} \textcircled{2} \quad \sum_{m=1}^n \underbrace{H_m}_{\downarrow \mathcal{F}_{m-1}\text{-measurable}} \cdot \underbrace{(X_m - X_{m-1})}_{\downarrow \mathcal{F}_m\text{-measurable}} &\dots \mathcal{F}_n\text{-measurable} \end{aligned}$$

$$\begin{aligned} \textcircled{3} \quad E[(H \cdot X)_{n+1} | \mathcal{F}_n] \\ = E[(H \cdot X)_n + H_{n+1}(X_{n+1} - X_n) | \mathcal{F}_n] \end{aligned}$$

$$\begin{aligned}
 &= (H \cdot X)_n + E[H_{n+1}(X_{n+1} - X_n) | \mathcal{F}_n] \\
 &= (H \cdot X)_n + \underbrace{H_{n+1}}_{\geq 0} \cdot \underbrace{E[X_{n+1} - X_n | \mathcal{F}_n]}_{\geq 0} \quad (H_{n+1} \dots \mathcal{F}_n\text{-measurable}) \\
 &\quad \uparrow \\
 &\quad \text{(by assumption)} \quad \because \{X_n\}_{n \geq 1} \text{ supermartingale}
 \end{aligned}$$

$$\geq (H \cdot X)_n$$

So $\{(H \cdot X)_n\}_{n \geq 1}$ is also a supermartingale. $\square$

NOTE If $\{X_n\}_{n \geq 1}$ is a submartingale.

$$\begin{aligned}
 \text{Then } E[(H \cdot X)_{n+1} | \mathcal{F}_n] &= (H \cdot X)_n + \underbrace{H_{n+1}}_{\geq 0} \cdot \underbrace{E[X_{n+1} - X_n | \mathcal{F}_n]}_{\leq 0} \\
 &\leq (H \cdot X)_n
 \end{aligned}$$

So $\{(H \cdot X)_n\}_{n \geq 1}$ is a submartingale.

NOTE If $\{X_n\}_{n \geq 1}$ is a martingale. Then $\{(H \cdot X)_n\}_{n \geq 1}$

is a martingale. (H_n does not have to be positive.)

32

- $\{X_n\}_{n \geq 1}$ is supermartingale
- N is stopping time

Let $H_n \stackrel{\text{def}}{=} \mathbb{I}_{\{N \geq n\}}$.

Then $\{H_n\}_{n \geq 1}$ is a predictable sequence

because $\{N \geq n\}^c = \{N < n\} \in \mathcal{F}_{n-1}$

$\Rightarrow \{N \geq n\} \in \mathcal{F}_{n-1}, \quad \therefore H_n$ is $\mathcal{F}_{n-1}$ -measurable

Now we consider

$$(H \cdot X)_n = \sum_{m=1}^n H_m (X_m - X_{m-1})$$

~~case~~ $N \geq n$

$\Rightarrow N \geq m$ for all $m=1, \dots, n$

$\Rightarrow H_m = 1$ for all $m=1, \dots, n$

$$\therefore (A \cdot X)_n = X_n - X_0$$

$$\boxed{\text{case}} \quad N < n$$

$$\Rightarrow H_m = I\{N \geq m\} = 0 \quad \text{for all } m = N+1 \sim n$$

$$\begin{aligned} \Rightarrow (H \cdot X)_n &= \sum_{m=1}^n H_m (X_m - X_{m-1}) = \sum_{m=1}^N (\sim) + \sum_{m=N+1}^n (\sim) \\ &= \sum_{m=1}^N 1 \cdot (X_m - X_{m-1}) = X_N - X_0 \end{aligned}$$

$$\text{Hence } (H \cdot X)_n = X_{n \wedge N} - X_0 \quad (n \wedge N \stackrel{\text{def}}{=} \min\{n, N\})$$

Since $\{H_n\}_{n \geq 1}$ is positive, bounded

so $(H \cdot X)_n = X_{n \wedge N} - X_0$ is also a supermartingale.

It's easy to verify that $X_{n \wedge N}$ is a supermartingale.

(*) $\{X_n\}, \{Y_n\}$ are supermartingale $\Rightarrow \{X_n + cY_n\}$: supermartingale

Let $\overset{\text{def}}{Y}_n = X_0 \Rightarrow \{Y_n\}_{n \geq 1}$: martingale $\Rightarrow$ supermartingale

33 Here we prove the upcrossing inequality.

(1) For proof, we introduce $\{N_k\}_{k \geq 0}$, stopping times

$$N_0 \stackrel{\text{def}}{=} -1.$$

$$N_{2k+1} \stackrel{\text{def}}{=} \inf \{n \mid n > N_{2k}, X_n \leq a\}$$

$$N_{2k} \stackrel{\text{def}}{=} \inf \{n \mid n > N_{2k-1}, X_n \geq b\}$$

⊛ We can imagine that $\{X_n\}_{n \geq 1}$ is going back and forth around $[a, b]$.

N_{2k-1}, N_{2k} are the index number of $\{X_n\}_{n \geq 1}$ which goes above "b" or below "a".

Now we prove that $\{N_k\}_{k \geq 1}$ are stopping times by induction.

Suppose $N_1, N_2, \dots, N_{2k-2}$ are stopping times.

$$\begin{aligned}
 \text{Then } \{N_{2k-1} = n\} &= \bigcup_{m=1}^{n-1} \{N_{2k-2} = m\} \cap \{N_{2k-1} = n\} \\
 &= \bigcup_{m=1}^{n-1} \{N_{2k-2} = m\} \cap \{X_{m+1} > a\} \cap \dots \\
 &\quad \cap \{X_{n-1} > a\} \cap \{X_n \leq a\} \\
 &\in \mathcal{F}_n
 \end{aligned}$$

($\because$ By assumption $\{N_{2k-2} = m\} \in \mathcal{F}_m \subseteq \mathcal{F}_n$

and $\{X_{m+1} > a\} \dots \{X_n \leq a\} \in \mathcal{F}_n$.)

So N_{2k-1} is also a stopping time.

Similarly N_{2k} is also a stopping time.

Hence the proof is complete.

$$(2) H_m \stackrel{\text{def}}{=} \prod_{k=1}^{\infty} \{N_{2k-1} < m \leq N_{2k}\}$$

$$= \prod_{k=1}^{\infty} \{N_{2k-1} < m\} \cdot \prod \{m \leq N_{2k}\}$$

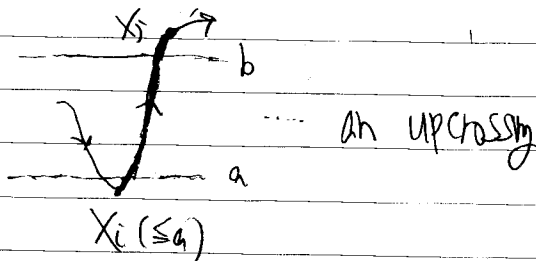
$$= \prod_{k=1}^{\infty} \underbrace{\{N_{2k-1} < m\}}_{\in \mathcal{F}_{m-1}} \cdot \prod \underbrace{\{N_{2k} < m\}^c}_{\in \mathcal{F}_m}$$

Hence, H_m is $\mathcal{F}_{m-1}$ -measurable.

$\therefore I_t$ is a predictable sequence.

$$(3) \text{ Now we define } U_n = \sup \{k \mid N_{2k} \leq n\}$$

U_n is the number of upcrossings seen between $\{X_1, X_2, \dots, X_n\}$.



(4) If $\{X_n\}_{n \geq 1}$ is a submartingale, then

$$(b-a) E[U_n] \leq E[(X_n - a)^+] - E[(X_0 - a)^+]$$

(proof)

$$\text{Let } \phi(x) = a + (x-a)^+. \quad ((x-a)^+ = \max\{x-a, 0\})$$

Then $\phi(x)$ is convex and monotone increasing.

$$Y_n \stackrel{\text{def}}{=} \phi(X_n).$$

By [29], Y_n is also a submartingale.

By [31], $(H \cdot Y)_n$ is also a submartingale.

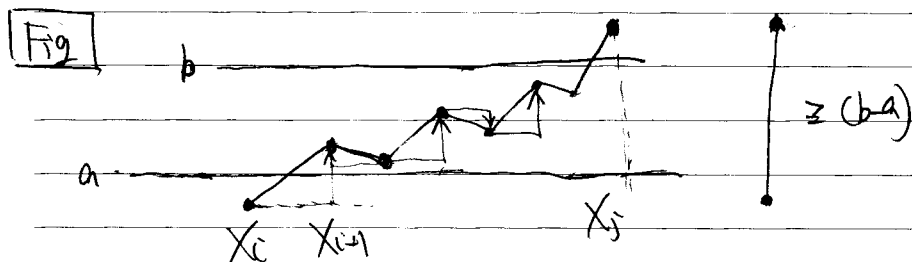
(See [31]. **NOTE** the statement still holds for submartingale)

(H_n is defined as (2))

$$\begin{aligned} \text{Now } (H \cdot Y)_n &= \sum_{m=1}^n H_m (Y_m - Y_{m-1}) \\ &= \sum_{m=1}^n \underbrace{\mathbb{I}_{\bigcup_{k=1}^m \{N_{2k-1} \leq m \leq N_{2k}\}}}_{\downarrow} (Y_m - Y_{m-1}) \end{aligned}$$

This part is 1 when X_n is going up from "a" to "b" otherwise it is 0. (See the definition of N_{2k-1} and N_{2k})

So if we sum up $(X_m - X_{m-1})$ for such m ,
 then it will exceed $(b-a)$ (x number of upcrossings)



$$\Rightarrow (X_{i+1} - X_i) + (X_{i+2} - X_{i+1}) + \dots + (X_j - X_{j-1}) = X_j - X_i \geq (b-a)$$

It's not difficult to see that we may
 change X_m to Y_m . ($Y_i \rightarrow 0$, $Y_{i+1} = X_{i+1}, \dots$)

$$\text{So } (HY)_n \geq (b-a) \cdot \underbrace{U_n}_{\text{number of upcrossings}}$$

number of upcrossings

Now let $K_m = 1 - H_m$

$$\text{Then } (K \cdot Y)_n + \underbrace{(H \cdot Y)_n}_{\geq (b-a)U_n} = Y_n - Y_0$$

$$\therefore Y_n - Y_0 \geq (K \cdot Y)_n + (b-a) \cdot U_n \dots (*)$$

We should notice that $K_m \geq 0$ and K_m is also a predictable sequence, hence $(K \cdot Y)_n$ is a submartingale.

$$\therefore E[(K \cdot Y)_n | \mathcal{F}_{n-1}] \geq (K \cdot Y)_{n-1}$$

$$\Rightarrow E[(K \cdot Y)_n] \geq E[(K \cdot Y)_{n-1}]$$

$$\Rightarrow E[(K \cdot Y)_n] \geq E[(K \cdot Y)_1]$$

$$(K \cdot Y)_1 = \prod_{k=1}^{\infty} \{N_{2k-1} < m \leq N_{2k}\}^c (Y_1 - Y_0) (= Y_1 - Y_0 \text{ or } 0) \\ \geq \min\{0, Y_1 - Y_0\}$$

$$\therefore E[(K \cdot Y)_1] \geq \min\{E[0], E[Y_1 - Y_0]\} \geq 0$$

($\because \{Y_n\}_{n \geq 1}$ submartingale)

$$\therefore E[(K \cdot Y)_n] \geq 0 \dots (**)$$

Finally $E[Y_n - Y_0] \geq (b-a) E[U_n]$ (by $(*)$, $(**)$)

34 We show the martingale convergence theorem

- (- $\{X_n\}_{n \geq 1} \dots$ submartingale
- (- $\sup_{n \geq 1} E[X_n^+] < \infty$

Recall the upcrossing inequality

$$(b-a) E[U_n] \leq E[(X_n - a)^+] - E[(X_0 - a)^+]$$

By definition of $U_n = \sup_{k \geq 1} \{k \mid N_k \leq n\}$

$$U_n \leq U_{n+1}. \quad \therefore U_n \uparrow U.$$

So monotone convergence theorem implies that

$$\lim_{n \rightarrow \infty} (b-a) E[U_n] = (b-a) E[U]$$

$$(\sim) \leq \liminf_{n \rightarrow \infty} (E[(X_n - a)^+] - E[(X_0 - a)^+])$$

$$\text{Since } (X_n - a)^+ \leq X_n^+ + |a|$$

$$\text{We have } (b-a) E[U] \leq \sup_{n \geq 1} E[X_n^+] + |a| - E[(X_0 - a)^+] < \infty$$

$$\text{So } E[U] < \infty \Rightarrow \underline{U < \infty} \text{ (a.s.)}$$

This means that the number of upcrossings in $\{X_n\}_{n \in \mathbb{N}}$ is finite almost surely.

Hence the number of $\{X_n\}_{n \in \mathbb{N}}$ which is larger than "b" is finite (a.s.). Similarly the number of $\{X_n\}_{n \in \mathbb{N}}$ which is smaller than "a" is finite (a.s.)

$$\text{So } P(\liminf X_n < a < b < \limsup X_n) = 0$$

$$(\because \limsup X_n \leq b \text{ (a.s.) ; } \liminf X_n \geq a \text{ (a.s.)})$$

(for all a, b)

$$\therefore P\left(\bigcup_{(a,b) \in \mathbb{Q}^2} \{\liminf X_n < a < b < \limsup X_n\}\right) = 0$$

($\mathbb{Q}$: rational numbers)

$$\Rightarrow P(\liminf X_n = \limsup X_n) = 1 \quad (\mathbb{Q}: \text{dense set})$$

$\therefore \lim_{n \rightarrow \infty} X_n$ exists almost surely.

Finally we show that $E|X| < \infty$

• By Fatou's lemma

$$E[\liminf X_n^+] = E[X^+] \leq \liminf E[X_n^+] \leq \sup_{n \geq 1} E[X_n^+] < \infty$$

$$\therefore E[X^+] < \infty$$

• Again by Fatou's lemma,

$$E[\liminf X_n^-] = E[X^-] \leq \liminf E[X_n^-] \quad \dots \textcircled{*}$$

$$\text{And } E[X_n^-] = E[X_n^+] - E[X_n]$$

Since $\{X_n\}_{n \geq 1}$ is a submartingale,

$$E[X_n] \geq E[X_{n-1}] \geq \dots \geq E[X_0]$$

$$\text{So } E[X_n] \leq E[X_n^+] - E[X_n] \leq E[X_n^+] - E[X_0]$$

$$\therefore \textcircled{*} \leq \sup_{n \geq 1} E[X_n^+] - E[X_0] < \infty$$

$$\therefore E[X^-] < \infty$$

[35] $X_n \geq 0$, X_n : supermartingale.

$\Rightarrow -X_n \leq 0$, $-X_n$: submartingale

$\Rightarrow (-X_n)^+ = 0 \quad \therefore \sup_{n \geq 1} E[(-X_n)^+] = 0 < \infty$

By [34], $-X_n$ converges (a.s.)

$\Leftrightarrow X_n$ converges (a.s.)

$\therefore X_n \rightarrow X$ (a.s.)

Since $\{X_n\}_{n \geq 1}$ is a supermartingale

$$E[X_n | \mathcal{F}_{n-1}] \leq X_{n-1} \Rightarrow E[X_n] \leq E[X_{n-1}] \leq E[X_0]$$

$$\text{By Fatou's lemma, } E[\liminf X_n] \leq \liminf E[X_n] \leq E[X_0]$$

$$\quad \quad \quad \parallel$$

$$\quad \quad \quad E[X]$$

So the proof is complete.

36

$$\text{Let } S_0 = 1, S_1 = 1 + Y_1, S_2 = 1 + Y_1 + Y_2, \dots$$

where $Y_1, Y_2, \dots, Y_n, \dots$ are iid random variables

$$P(Y_i = 1) = P(Y_i = -1) = \frac{1}{2}.$$

$$\text{Now let } \mathcal{F}_n = \sigma(Y_1, Y_2, \dots, Y_n).$$

$$\text{Then } E[S_{n+1} | \mathcal{F}_n] = E[Y_{n+1} + S_n | \mathcal{F}_n]$$

$$= \underbrace{S_n}_{(2)} + \underbrace{E[Y_{n+1} | \mathcal{F}_n]}_{(3)} = \underbrace{S_n}_{(2)} + \underbrace{0}_{(3)} = S_n$$

$$E[S_n] = 0 \in (-\infty, \infty)$$

S_n : $\mathcal{F}_n$ -measurable

$\Rightarrow S_0, \{S_n\}$ is a martingale

Let $N = \inf\{n \mid S_n = 0\}$.

Then N is a stopping time.

By [B2], $S_{n \wedge N}$ is a martingale.

($\because$ It's easy to verify B2 holds for martingales/submartingales)

Hence $S_{n \wedge N}$ is a supermartingale.

And $S_{n \wedge N} \geq 0$ ($\because N \stackrel{\text{def}}{=} \inf\{n \mid S_n = 0\}$
and $S_0 = 1 > 0$)

By [35] $S_{n \wedge N}$ converges almost surely. ($S_{n \wedge N} \xrightarrow{\text{a.s.}} S^*$)

And $\forall k \geq 1, P(S^* = k) > 0$.

because if $S_{n \wedge N} = k$, then $S_{(n+1) \wedge N} = k+1$ or $k-1$.

Therefore $P(S^* = 0) = 1 \quad \therefore S^* = 0$ (a.s.)

Now $E|S_{n \wedge N} - S^*| = E|S_{n \wedge N}| = E[S_{n \wedge N}] \geq 1 > 0 \quad (\forall n \geq 1)$

$\Rightarrow$ does not converge in L^1 .

37 Consider $\{X_k\}_{k \geq 1}$ which is defined as

- $X_k | X_{k-1} = 0 \quad \begin{cases} 1 & (\frac{1}{2k}) \\ -1 & (\frac{1}{2k}) \\ 0 & (1 - \frac{1}{k}) \end{cases}$
- $X_k | X_{k-1} \neq 0 \quad \begin{cases} kX_{k-1} & (\frac{1}{k}) \\ 0 & (1 - \frac{1}{k}) \end{cases}$
- $X_0 = 0$

$$\mathcal{F}_k = \sigma(X_1, X_2, \dots, X_k)$$

$$\text{then } E[X_k | \mathcal{F}_{k-1}] = E[X_k | X_{k-1}] = X_{k-1}$$

$$\text{because } E[X_k | X_{k-1} = 0] = \frac{1}{2k} \cdot (1) + \frac{1}{2k} \cdot (-1) = 0$$

$$E[X_k | X_{k-1} = x_{k-1}] = (kx_{k-1}) \cdot \frac{1}{k} = x_{k-1} \quad (\neq 0)$$

$$\therefore E[X_k] = E[X_{k-1}] = E[X_0] = 0$$

So $\{X_k\}_{k \geq 1}$ is a martingale

$$\begin{aligned}
 P(X_k=0) &= E[\mathbb{I}_{\{X_k=0\}}] = E \cdot E[\mathbb{I}_{\{X_k=0\}} | X_{k-1}] \\
 &= E \cdot P(X_k=0 | X_{k-1}) = E\left[1 - \frac{1}{k}\right] = 1 - \frac{1}{k}
 \end{aligned}$$

$$\therefore \lim_{k \rightarrow \infty} P(X_k=0) = 1 \quad ; \quad X_k \xrightarrow{P} 0.$$

Next we show that $X_k \not\rightarrow 0$ (a.s.)

Consider the probability $P(X_k=0 \text{ for all } k \geq K)$

$$\begin{aligned}
 &= \underbrace{P(X_K=0)}_{\leq 1} \cdot \underbrace{P(X_{K+1}=0 | X_K=0)}_{= 1 - \frac{1}{K+1}} \cdot \underbrace{P(X_{K+2}=0 | X_{K+1}=0)}_{= 1 - \frac{1}{K+2}} \cdots \\
 &\leq 1 \cdot \left(1 - \frac{1}{K+1}\right) \cdot \left(1 - \frac{1}{K+2}\right) \cdots
 \end{aligned}$$

$$\therefore P(X_k=0 \text{ for all } k \geq K) \leq \prod_{k=K+1}^{\infty} \left(1 - \frac{1}{k}\right) = 0 \dots (*)$$

$$(*) \quad \forall x \in \mathbb{R}, \quad e^x \geq 1+x \Rightarrow e^{-x} \geq 1-x \Rightarrow -x \geq \ln(1-x)$$

$$\Rightarrow \ln\left(1 - \frac{1}{k}\right) \leq -\frac{1}{k} \Rightarrow \sum_{k=K+1}^{\infty} \ln\left(1 - \frac{1}{k}\right) = -\infty \Rightarrow \prod_{k=K+1}^{\infty} \left(1 - \frac{1}{k}\right) = 0$$

And X_k is integer-valued.

So $X_k \rightarrow 0 \Leftrightarrow \exists K$ s.t. $X_k=0$ for all $k \geq K$.

$$\therefore P(X_k \rightarrow 0) = P\left(\bigcup_{K=1}^{\infty} \{X_k=0 \text{ for all } k \geq K\}\right) \leq \sum_{K=1}^{\infty} P(\sim) = 0$$

$$\boxed{38} \quad Z_n = X_n \vee Y_n = \max \{X_n, Y_n\}$$

is also a submartingale

$$\textcircled{1} \quad |Z_n| \leq |X_n| + |Y_n|$$

$$E|Z_n| \leq E|X_n| + E|Y_n| < \infty$$

$\textcircled{2} \quad Z_n$ is $\mathcal{F}_n$ -measurable (X_n, Y_n are $\mathcal{F}_n$ -measurable)

$$\textcircled{3} \quad Z_{n+1} = X_{n+1} \vee Y_{n+1} \geq X_{n+1} \quad (\geq Y_{n+1})$$

$$\therefore E[Z_{n+1} | \mathcal{F}_n] \geq E[X_{n+1} | \mathcal{F}_n] \geq X_n$$

$\therefore X_n$ is a submartingale

Similarly $E[Z_{n+1} | \mathcal{F}_n] \geq Y_n$

$$\therefore E[Z_{n+1} | \mathcal{F}_n] \geq \max \{X_n, Y_n\}$$

$$\quad \quad \quad \parallel$$

$$\quad \quad \quad Z_n$$

$\therefore \{Z_n\}_{n \geq 1}$ is a sub-martingale

39

(1) We use the supermartingale convergence theorem [35]

$X_n \geq 0$ and X_n is a martingale ($\Rightarrow$ supermartingale)

$\therefore X_n \rightarrow X$ (a.s.)

Next, $X_n \rightarrow X$ (a.s.) $\Rightarrow X_n \xrightarrow{P} X$.

It's enough to show that $X_n \xrightarrow{P} 0$

$$P(|Y_m - 1| > 0) = P(Y_m \neq 1) > 0$$

$\Rightarrow$ We may find small $\epsilon > 0$ st $P(|Y_m - 1| \geq \epsilon) > 0$

Let $\delta > 0$ be an arbitrary positive number

$$P(|X_{n+1} - X_n| \geq \delta \epsilon) = P(|X_{n+1} - 1| |X_n| \geq \delta \epsilon)$$

$$\geq P(|Y_{n+1} - 1| \geq \epsilon \cap |X_n| \geq \delta)$$

$$= \underbrace{P(|Y_{n+1} - 1| \geq \epsilon)}_{> 0} P(|X_n| \geq \delta)$$

$$\text{And LHS: } P(|X_{n+1} - X_n| \geq \delta \epsilon) \leq P\left(|X_{n+1} - X| \geq \frac{\delta \epsilon}{2}\right) + P\left(|X_n - X| \geq \frac{\delta \epsilon}{2}\right)$$

$\rightarrow 0$ as $n \rightarrow \infty$. ($\because X_n \rightarrow X$)

Hence $RHS \rightarrow 0$ as $n \rightarrow \infty$

So $P(|X_n| \geq \delta) \rightarrow 0$ as $n \rightarrow \infty$

Since $\delta > 0$ is arbitrary, so $X_n \xrightarrow{P} 0$.

Now the proof is complete.

$$(2) \frac{1}{n} \log \prod_{m=1}^n Y_m = \frac{\log Y_1 + \dots + \log Y_n}{n}$$

$$\text{By SLLN. } \xrightarrow{a.s.} E[\log Y_1] = c$$

It's enough to show that $E[\log Y_1] < 0$

By Jensen's inequality. $E[\log Y_1] \leq \log E[Y_1] = 0$.

Its equality holds when $Y_1 = \text{constant}$ (a.s.)

But by assumption. $P(Y_1 = 1) < 1$

So equality does not hold

$$\therefore c = E[\log Y_1] < 0$$

[40] Let $W_n \stackrel{\text{def}}{=} \frac{X_n}{\prod_{m=1}^n (1+Y_m)}$

Then $E[W_{n+1} | \mathcal{F}_n] = E\left[\frac{X_{n+1}}{\prod_{m=1}^{n+1} (1+Y_m)} \mid \mathcal{F}_n\right]$

$= \frac{1}{\prod_{m=1}^n (1+Y_m)} E[X_{n+1} | \mathcal{F}_n]$ by assumption

[20] $\left(\frac{1}{\prod_{m=1}^n (1+Y_m)}\right) : \mathcal{F}_n\text{-measurable}$

$\leq \frac{1}{\prod_{m=1}^n (1+Y_m)} (1+Y_n) X_n = \frac{X_n}{\prod_{m=1}^n (1+Y_m)} = W_n$

So $\{W_n\}_{n \geq 1}$ is a Supermartingale.

Since $\{X_n\}_{n \geq 1}, \{Y_n\}_{n \geq 1}$ are positive, $W_n \geq 0$

We may apply [35] to $\{W_n\}_{n \geq 1}$

$\therefore W_n \rightarrow W$ (a.s.)

Next. We consider convergence of $\prod_{m=1}^n (1+Y_m)$.

$$\ln \prod_{m=1}^n (1+Y_m) = \sum_{m=1}^n \ln(1+Y_m)$$

$$1+x \leq e^x \Rightarrow \ln(1+x) \leq x \Rightarrow \sum_{m=1}^n \ln(1+Y_m) \leq \sum_{m=1}^n Y_m$$

$$\therefore 0 \leq \sum_{m=1}^n \ln(1+Y_m) \leq \sum_{m=1}^n Y_m$$

(Y_m positive, so $\ln(1+Y_m) \geq 0$)

Since $\sum_{n=1}^{\infty} Y_n < \infty$ (a.s.), $\sum_{n=1}^{\infty} \ln(1+Y_n)$ also converges to a finite limit (a.s.).

Hence $\prod_{m=1}^n (1+Y_m)$ converges (a.s.)

$$\therefore \frac{X_n}{\prod_{m=1}^n (1+Y_m)} = \prod_{m=1}^n (1+Y_m) = X_n \quad \text{converges (a.s.)}$$

41

Step 1

Let $N_k \stackrel{\text{def}}{=} \inf \{n \mid X_n \leq -k\}$

Since $D = \{ \limsup X_n = \infty \cap \liminf X_n = -\infty \}$

If $\omega \in D^c$, then $\textcircled{1} \liminf X_n(\omega) > -\infty$ or $\textcircled{2} \limsup X_n(\omega) < \infty$

Now suppose $\textcircled{1} \liminf_{n \rightarrow \infty} X_n(\omega) > -\infty$.

We may find a positive number $\tilde{K}(\omega) < \infty$

such that $\liminf_{n \rightarrow \infty} X_n(\omega) > -\tilde{K}(\omega)$

For sufficiently large m (say $m > M(\omega)$), $X_m(\omega) > -\tilde{K}(\omega)$

Let $-K(\omega) = \min \{X_1(\omega), \dots, X_{M(\omega)}(\omega), -\tilde{K}(\omega)\}$

Then $X_m(\omega) > -K(\omega)$ for all $m \geq 1$.

$$\begin{aligned} \therefore \{ \liminf X_n(\omega) > -\infty \} &\subseteq \bigcup_{k=1}^{\infty} \{ X_m(\omega) > -k \text{ for all } m \geq 1 \} \\ &\subseteq \bigcup_{k=1}^{\infty} \{ N_k = \infty \} \end{aligned}$$

(Similar argument is valid for $\textcircled{2}$)

Step 2

- Again $N_k \stackrel{\text{def}}{=} \inf \{n \mid X_n \leq -k\}$
 where k is a constant and is not related with ω .
- Since N_k is a stopping time,
 $X_{n \wedge N_k}$ is also a martingale by [32].
- $X_{n \wedge N_k} + k + M \geq 0$
 ($\because X_{n \wedge N_k - 1} > -k$ and $|X_m - X_{m+1}| \leq M < \infty$)
- We apply [35] to $\{X_{n \wedge N_k} + k + M\}_{n \geq 1}$
 $\lim_{n \rightarrow \infty} X_{n \wedge N_k} + k + M$ exists and finite (a.s.)
- $$(0 \leq E[\lim_{n \rightarrow \infty} X_{n \wedge N_k} + k + M] \leq E[X_1 + k + M] < \infty$$

$$\therefore \lim_{n \rightarrow \infty} X_{n \wedge N_k} + k + M < \infty \text{ (a.s.)}) \quad (\text{we suppose } X_0 = 0)$$
- Hence $\lim_{n \rightarrow \infty} X_{n \wedge N_k}$ exists and finite (a.s.)
- If $\omega \in \{N_k = \infty\}$, Then $X_{n \wedge N_k} = X_n$
 So $\lim_{n \rightarrow \infty} X_n$ exists and finite.

Step 3 Now recall Step 1

$$\omega \in D^c \stackrel{(*)}{\Rightarrow} \omega \in \bigcup_{k=1}^{\infty} \{N_k = \infty\}$$

$$\Rightarrow \exists K \in \mathbb{N} \text{ s.t. } N_K = \infty \quad \downarrow \text{Step 2}$$

$$\Rightarrow \lim_{n \rightarrow \infty} X_n \text{ exists and finite (a.s.)}$$

$$\Rightarrow \omega \in C \quad (\text{a.s.})$$

↑
(Except ω in a measure zero set)

$$\therefore P(C \cup D) = 1$$



(*) We should both discuss $\omega \in \left\{ \liminf X_n > -\infty \right\}$
and $\omega \in \left\{ \limsup X_n < \infty \right\}$

but they are quite similar

42 (Doob's decomposition)

$$A_n \stackrel{\text{def}}{=} \sum_{m=1}^n E[X_m - X_{m-1} | \mathcal{F}_{m-1}]$$

$$M_n \stackrel{\text{def}}{=} X_n - A_n$$

Then $X_n = M_n + A_n$

① A_n is a predictable and increasing sequence

- A_n is $\mathcal{F}_{n-1}$ -measurable by its definition.

$$E[X_m - X_{m-1} | \mathcal{F}_{m-1}] = \underbrace{E[X_m | \mathcal{F}_{m-1}]}_{\geq X_{m-1}} - X_{m-1} \geq 0$$

So A_n is increasing.

② M_n is a martingale

- $E[A_{n+1} | \mathcal{F}_n] = A_{n+1}$ ($\because A_{n+1}$ is $\mathcal{F}_n$ -measurable and by 42)

- $E[M_{n+1} | \mathcal{F}_n] = E[X_{n+1} - A_{n+1} | \mathcal{F}_n]$

$$= E[X_{n+1} | \mathcal{F}_n] - A_{n+1}$$

$$(X_{n+1} - X_n + X_n)$$

$$= E[X_{n+1} - X_n | \mathcal{F}_n] + E[X_n | \mathcal{F}_n] - A_{n+1} \\ \left(\sum_{m=1}^{n+1} E[X_m - X_{m-1} | \mathcal{F}_{m-1}] \right)$$

$$= E[X_n | \mathcal{F}_n] - \sum_{m=1}^n E[X_m - X_{m-1} | \mathcal{F}_{m-1}]$$

$$= X_n - \sum_{m=1}^n E[X_m - X_{m-1} | \mathcal{F}_{m-1}]$$

$$= X_n - A_n = M_n$$

$\therefore M_n$ is a martingale

$$\mathcal{F}_0 = \{\emptyset, \Omega\}, A_n \in \mathcal{F}_n$$

[43] The statement in the textbook may be wrong.

Here we prove that $P(\{A_n\}_{n \geq 0} \triangle \{\sum_{m=1}^{\infty} P(A_m | \mathcal{F}_{m-1}) = \infty\}) = 0$

(proof) $X_n \stackrel{\text{def}}{=} \sum_{m=1}^n \mathbb{I}_{A_m}$

$$E[X_{n+1} | \mathcal{F}_n] = X_n + \underbrace{E[\mathbb{I}_{A_{n+1}} | \mathcal{F}_n]}_{\geq 0} \geq X_n$$

$\therefore \{X_n\}_{n \geq 1}$ is a submartingale

By [42], $X_n = M_n + K_n$ (Doob's decomposition)

where M_n : martingale, K_n : increasing predictable sequence

$$M_n = X_n - \sum_{m=1}^n E[X_m - X_{m-1} | \mathcal{F}_{m-1}]$$

$$= \sum_{m=1}^n \mathbb{I}_{A_m} - \sum_{m=1}^n E[\mathbb{I}_{A_m} | \mathcal{F}_{m-1}]$$

$$= \sum_{m=1}^n (\mathbb{I}_{A_m} - P(A_m | \mathcal{F}_{m-1}))$$

$$|M_n - M_{n-1}| = |\mathbb{I}_{A_n} - P(A_n | \mathcal{F}_{n-1})| \leq 2$$

So we may apply [41]

$$C = \left\{ \lim_{n \rightarrow \infty} M_n \text{ exists and is finite} \right\}$$

$$D = \left\{ \limsup_{n \rightarrow \infty} M_n = \infty \text{ and } \liminf_{n \rightarrow \infty} M_n = -\infty \right\}$$

$$P(C \cup D) = 1$$

① $w \in C$

$$\text{If } \sum_{n=1}^{\infty} \mathbb{I}_{A_n}(w) = \infty, \text{ then } \sum_{n=1}^{\infty} P(A_n | \mathcal{F}_{n-1}) = \infty$$

$$(\because M_n = \sum_{k=1}^n \{ \mathbb{I}_{A_k} - P(A_k | \mathcal{F}_{k-1}) \}$$

$$\text{and } \lim_{n \rightarrow \infty} M_n \text{ exists and finite})$$

$$\text{Similarly } \sum_{n=1}^{\infty} P(A_n | \mathcal{F}_{n-1}) = \infty \Rightarrow \sum_{n=1}^{\infty} \mathbb{I}_{A_n}(w) = \infty$$

$$\therefore \left\{ \sum_{n=1}^{\infty} \mathbb{I}_{A_n}(w) = \infty \right\} \Leftrightarrow \left\{ \sum_{n=1}^{\infty} P(A_n | \mathcal{F}_{n-1}) = \infty \right\}$$

② $w \in D$

$$\sum_{n=1}^{\infty} \mathbb{I}_{A_n}(w) = \infty \text{ and } \sum_{n=1}^{\infty} P(A_n | \mathcal{F}_{n-1}) = \infty$$

$$\text{So if } w \in C \cup D \Rightarrow \left\{ \sum_{n=1}^{\infty} \mathbb{I}_{A_n}(w) = \infty \right\} \Leftrightarrow \left\{ \sum_{n=1}^{\infty} P(A_n | \mathcal{F}_{n-1}) = \infty \right\}$$

$$\therefore \left\{ \sum_{n=1}^{\infty} \mathbb{I}_{A_n}(w) = \infty \right\} \triangleq \left\{ \sum_{n=1}^{\infty} P(A_n | \mathcal{F}_{n-1}) = \infty \right\} \subseteq (C \cup D)^c$$

$$\text{"(Axiom)"}$$

$$P(C \cup D)^c = 0. \text{ The statement is true. } \square \text{ CHASHIN}$$

44 $\{Z_i^{(n)}\}_{i=1}^n \dots n!d$, non-negative integer valued.

$$Z_{n+1} \stackrel{\text{def}}{=} \begin{cases} \sum_{i=1}^{Z_n^{(n)}} 1 + \sum_{i=1}^{Z_n^{(n)}} Z_i^{(n)} & \text{if } Z_n \geq 1 \\ 0 & \text{if } Z_n = 0 \end{cases}$$

Then $\{Z_n\}_{n \geq 1}$ is called a Galton-Watson process

[45] Z_n is a Galton-Watson process

where $E[Z_1^M] = \mu > 0$.

Let $W_n = \frac{Z_n}{\mu^n}$.

① W_n is $\mathcal{F}_n$ -measurable

- $Z_1 = \sum_{i=1}^{Z_0} \dots$ $\mathcal{F}_1$ -measurable

- Suppose Z_k is $\mathcal{F}_k$ -measurable

$$\begin{aligned} \text{Then } Z_{k+1} &= \mathbb{I}_{\{Z_k > 0\}} \cdot \left(\sum_{j=1}^{Z_k} Y_j^{(k+1)} \right) \\ &\in \mathcal{F}_k \subseteq \mathcal{F}_{k+1} \quad \in \sigma[\mathcal{F}_k \cup \{Y_j^{(k+1)}\}_{j=1}^{Z_k}] \\ &\subseteq \mathcal{F}_{k+1} \end{aligned}$$

By induction Z_k is $\mathcal{F}_k$ -measurable (So is W_k)

$$\begin{aligned} \text{② } E[W_{n+1} | \mathcal{F}_n] &= E\left[\frac{Z_{n+1}}{\mu^{n+1}} \mid \mathcal{F}_n\right] = E\left[\frac{Y_1^{(n+1)} + \dots + Y_{Z_n}^{(n+1)}}{\mu^{n+1}} \mid \mathcal{F}_n\right] \\ &= \frac{\mu \cdot Z_n}{\mu^{n+1}} = \frac{Z_n}{\mu^n} = W_n \quad (Y_j^{(n+1)} \sim Y_{Z_n}^{(n+1)} \text{ are independent of } \mathcal{F}_n) \end{aligned}$$

③ $E[W_{n+1}] = E[W_n]$ (by ②)

$$E[W] = E\left[\frac{Z}{\mu}\right] = E\left[\frac{Z_1}{\mu}\right] = 1$$

$$\therefore |E[W]| < \infty$$

16 We show that if $\mu < 1$, then $Z_n = 0$

for sufficiently large n .

$$W_n \stackrel{\text{def}}{=} \frac{Z_n}{\mu^n} \quad (\text{by 15 } E[W_n] = 1 \Rightarrow E[Z_n] = \mu^n)$$

Since W_n is a martingale and non-negative

$$\text{by 35, } W_n \rightarrow W \text{ (a.s.)} \quad E[W] \leq E[W] = 1$$

Next we show that $W_n \xrightarrow{P} 0$.

$$\begin{aligned} P(|W_n| > \epsilon) &\leq P(W_n > 0) = P(Z_n > 0) \leq E[Z_n \cdot \mathbb{I}_{\{Z_n > 0\}}] \\ &\leq E[Z_n] = \mu^n \quad (E[\mathbb{I}_{\{Z_n > 0\}}] \leq E[Z_n \cdot \mathbb{I}_{\{Z_n > 0\}}] \\ &\quad \because Z_n > 0 \Rightarrow Z_n \geq 1) \end{aligned}$$

$$\text{Hence } n \rightarrow \infty \quad \mu^n \downarrow 0. \quad \therefore W_n \xrightarrow{P} 0. \quad \therefore W = 0.$$

$$\frac{Z_n}{\mu^n} \xrightarrow{\text{a.s.}} 0. \quad \text{Since } Z_n \text{ is non-negative integer-valued}$$

and $\mu^n \ll 1$ $Z_n = 0$ for sufficiently large n .

Otherwise, there exist infinitely many n such that

$$Z_n \geq 1. \quad \text{Then} \quad \limsup_{n \rightarrow \infty} \frac{Z_n}{n} = +\infty.$$

So the statement holds. $\blacksquare$

47 Step 1

Let $k \geq 1$ (integer) and $N \geq 1$ (integer)

Consider $P(Z_n = k \text{ for all } n \geq N)$

$$= P(Z_N = k, Z_{N+1} = k, \dots)$$

$$\leq P(Z_N = k) \cdot P(Z_{N+1} = k \mid Z_N = k) \cdot P(Z_{N+2} = k \mid \dots)$$

$$\dots Z_{N+1} = k, Z_N = k) \cdot \dots$$

$$= P(Z_N = k) \cdot \underbrace{P(Z_{N+1} = k \mid Z_N = k)}_p \cdot \underbrace{P(Z_{N+2} = k \mid Z_{N+1} = k)}_p \dots$$

$$\leq P(Z_N = k) \cdot p^m \quad (m \geq 1)$$

$p < 1$ because $p = 1$ only when $k = 1$ and

$$P(Y_i^{(n)} = 1) = 1.$$

But $P(Y_i^{(m)} = 1) < 1$ so $p < 1$

By taking $m \rightarrow \infty$, we have $RHS \rightarrow 0$.

So $LHS = 0$.

$$\therefore P(\{Z_n = k \text{ for all } n \in \mathbb{N}\}) = 0$$

$$\Rightarrow P\left(\bigcup_{N=1}^{\infty} \{Z_n = k \text{ for all } n \geq N\}\right) = 0$$

$$= P(\{Z_n = k \text{ for sufficiently large } n\}) = 0$$

for all $k \geq 1$.

Step 2

$W_n = \frac{Z_n}{n} = \bar{Z}_n$ is a martingale and non-negative

By [35], $\bar{Z}_n \rightarrow \bar{Z}$ (a.s) and $\bar{Z}(\infty)$ (a.s)

Similar to [46], $\bar{Z}_n = \bar{Z}$ for sufficiently large $n \geq 1$

because Z_n is integer-valued (so is $\bar{Z}$).

Otherwise there exists infinitely many $n \geq 1$

$$\bar{Z}_n \neq \bar{Z} \Rightarrow \limsup |\bar{Z}_n - \bar{Z}| \geq 1.$$

Step 3 Finally by **Step 1** and **Step 2**,

we have $P(\{Z_n = 0 \text{ for sufficiently large } n\}) = 1$

so $\bar{Z} = 0$ (a.s)

48 $\{X_n\}_{n \geq 1}$: submartingale

N : Stopping time with $P(N \leq k) = 1$

We show that $E[X_0] \leq E[X_N] \leq E[X_k]$.

(proof)

By [32], $X_{n \wedge N}$ is also a submartingale, hence

$$E[X_{n \wedge N} | \mathcal{F}_m] \geq X_{m \wedge N}$$

$$\Rightarrow E[X_{n \wedge N}] \geq E[X_{m \wedge N}] \quad (\because E \circ E[X_{n \wedge N} | \mathcal{F}_m] = E[X_{n \wedge N}])$$

$$\therefore E[X_{0 \wedge N}] \leq \dots \leq E[X_{k \wedge N}]$$

$$\Rightarrow E[X_0] \leq E[X_N] \quad \dots (1)$$

Let $K_m \stackrel{\text{def}}{=} \mathbb{I}_{\{N \leq m\}}$... this is a predictable sequence

By [31], $(K_n X)_n$ is a submartingale.

$$(K_n X)_n = \sum_{m=1}^n \underbrace{K_m}_{\text{predictable}} (X_m - X_{m-1}) =$$

$$\begin{pmatrix} \bullet N \leq m-1 \Rightarrow 1 \\ \bullet N \geq m \Rightarrow 0 \end{pmatrix} \Rightarrow$$

$$\therefore \begin{pmatrix} \bullet \text{ if } N \geq n, & K_m = 0 & \text{for } m=1 \sim n \\ \bullet \text{ if } N \leq n-1, & K_m = 1 & \text{for } m=1 \sim n-1 \\ & K_m = 0 & \text{for } m=n \end{pmatrix}$$

$$= \begin{cases} 0 & (\text{if } N \geq n) \\ \sum_{m=N+1}^n 1 \cdot (X_m - X_{m-1}) = X_n - X_N & (\text{if } N \leq n-1) \end{cases}$$

$$= X_n - X_{n \wedge N} \quad (\wedge \dots \min\{\dots\})$$

So $\{X_n - X_{n \wedge N}\}_{n \geq 0}$ is a submartingale

$$\therefore E[X_k - X_{k \wedge N}] \geq E[X_0 - X_{0 \wedge N}] = 0$$

$$\Rightarrow E[X_k] \geq E[X_{k \wedge N}] - E[X_N] \dots \textcircled{2}$$

$$\textcircled{1} \text{ and } \textcircled{2} \Rightarrow E[X_0] \leq E[X_N] \leq E[X_k]$$

49 (Doob's inequality)

- $\{X_n\}_{n \geq 1} \dots$ submartingale
- $\overline{X}_n \stackrel{\text{def}}{=} \max_{0 \leq m \leq n} \{X_m^+\}$
- $A \stackrel{\text{def}}{=} \{X_n \geq \lambda\} \quad (\lambda > 0)$

We show that $\lambda P(A) \leq E[X_n | A] \leq E[X_n^+]$

(proof)

- $N \stackrel{\text{def}}{=} \inf \{m \mid X_m \geq \lambda\} \wedge n$ (and-min $\{a, b\}$)
- When $\omega \in A$, $\max_{0 \leq m \leq n} \{X_m^+\} \geq \lambda \Rightarrow$

There exists $m_0^{(\omega)} (0 \leq m_0 \leq n)$ st $X_{m_0}^+ \geq \lambda (> 0)$

$\Rightarrow X_{m_0} \geq \lambda \Rightarrow X_N \geq \lambda$ ($\because$ Definition of N)

$\therefore A \subseteq \{X_N \geq \lambda\}$

$$\therefore E[\lambda \cdot \mathbb{I}_A] \leq E[X_N \cdot \mathbb{I}_A],$$

$$\Rightarrow \lambda P(A) \leq E[X_N \cdot \mathbb{I}_A] \dots \textcircled{1}$$

Next since N is a stopping time and $P(N \leq n) = 1$,

$$\text{by } \textcircled{1} \quad E[X_N] \leq E[X_n] \dots \textcircled{2}$$

Furthermore, $\omega \in A^c \Rightarrow \max_{0 \leq m \leq n} \{X_m\} < \lambda \Rightarrow X_0, X_n < \lambda$

$$\Rightarrow \inf \{m \mid X_m \geq \lambda\} > n \Rightarrow N = n \neq X_N = X_n$$

$$\text{Thus } E[X_N \cdot \mathbb{I}_{A^c}] = E[X_n \cdot \mathbb{I}_{A^c}] \dots \textcircled{3}$$

$$\textcircled{2} \textcircled{3} \dots E[X_N \cdot \mathbb{I}_A] \leq E[X_n \cdot \mathbb{I}_A] \dots \textcircled{4}$$

$$\textcircled{1} \& \textcircled{4} \Rightarrow \underline{\lambda P(A) \leq E[X_n \cdot \mathbb{I}_A]}$$

$$\text{And } X_n \cdot \mathbb{I}_A \leq X_n^+ \cdot \mathbb{I}_A \leq X_n^+ \quad \therefore \underline{E[X_n \cdot \mathbb{I}_A] \leq E[X_n^+]}$$

$$\boxed{50} \quad S_n = \xi_1 + \xi_2 + \dots + \xi_n \quad (\xi_1, \xi_2, \dots \text{ independent})$$

$$E[\xi_m] = 0, \quad E[\xi_m^2] = \sigma_m^2 < \infty$$

$$\mathcal{F}_n = \sigma[\xi_1, \xi_2, \dots, \xi_n]$$

• $\{S_n^2\}_{n \geq 1}$ is a submartingale

$$E[S_{n+1}^2 | \mathcal{F}_n] = E[S_n^2 + 2S_n \xi_{n+1} + \xi_{n+1}^2 | \mathcal{F}_n]$$

$$= S_n^2 + 2S_n \underbrace{E[\xi_{n+1} | \mathcal{F}_n]}_0 + \underbrace{E[\xi_{n+1}^2 | \mathcal{F}_n]}_{\sigma_{n+1}^2}$$

$$\geq S_n^2$$

$$\bullet \stackrel{\text{def}}{A} = \left\{ \max_{1 \leq m \leq n} (S_m^2)^+ \geq x^2 \right\}$$

$$= \left\{ \max_{1 \leq m \leq n} S_m^2 \geq x^2 \right\}$$

$$= \left\{ \max_{1 \leq m \leq n} |S_m| \geq x \right\}$$

$$\text{By } \boxed{49} \quad x^2 \cdot P(A) \leq E[(S_n^2)^+] - E[S_n^2] = V[S_n]$$

$$\Rightarrow x^2 \cdot P\left(\max_{1 \leq m \leq n} |S_m| \geq x\right) \leq V[S_n]$$

[5] $\{X_n\}_{n \geq 1}$ is submartingale

$$\text{then } E[(\bar{X}_n)^p] \leq \left(\frac{p}{p-1}\right)^p E[(X_n^+)^p] \quad (p \in (1, \infty))$$

(proof)

$$\bar{X}_n = \max_{1 \leq m \leq n} \{X_m^+\}$$

let $M \in (0, \infty)$

$$\bullet \quad E[(\bar{X}_n \wedge M)^p] = \int_0^\infty p\lambda^{p-1} \cdot P(\bar{X}_n \wedge M > \lambda) d\lambda$$

$$(\because \text{[ch2]} \dots X \geq 0 \Rightarrow E[X^p] = \int_0^\infty p\lambda^{p-1} \cdot P(X > \lambda) d\lambda)$$

Since the number of points of discontinuity

is at most countable, and a countable

set has measure zero (Lebesgue measure),

$$\int_0^\infty p\lambda^{p-1} P(\bar{X}_n \wedge M > \lambda) d\lambda = \int_0^\infty p\lambda^{p-1} P(\bar{X}_n \wedge M \geq \lambda) d\lambda$$

$$\text{Let } A = \{X_n \wedge M \geq \lambda\}$$

$$X_n \wedge M = \max_{1 \leq m \leq n} \{X_m^+\} \wedge M = \max_{1 \leq m \leq n} \{X_m^+ \wedge M\}$$

$$= \max_{1 \leq m \leq n} \{(X_m \wedge M)^+\}$$

$$\text{Let } Y_n = (X_n \wedge M) \quad \nearrow$$

$$\text{Then } A = \{X_n \wedge M \geq \lambda\} = \{Y_n \geq \lambda\}$$

$$\text{By [49], } \lambda P(A) \leq E[Y_n \cdot \mathbb{I}_A]$$

$$= E[(X_n \wedge M) \cdot \mathbb{I}_A]$$

$$\leq E[X_n^+ \cdot \mathbb{I}_A]$$

$$\text{Hence, } \int_0^\infty \rho \lambda^{\rho-1} P(X_n \wedge M \geq \lambda) d\lambda = \int_0^\infty \rho \lambda^{\rho-2} \cdot \lambda P(A) d\lambda$$

$$\leq \int_0^\infty \rho \lambda^{\rho-2} \cdot E[X_n^+ \cdot \mathbb{I}_A] d\lambda$$

$$\begin{aligned}
&= \int_0^\infty p \lambda^{p-2} \int_{\Omega} X_n^+ \cdot \mathbb{I}_A \, dP \, d\lambda \\
&= \int_{\Omega} \int_0^\infty X_n^+ \cdot \mathbb{I}_A \cdot p \lambda^{p-2} \, d\lambda \, dP \quad \downarrow \text{Tonelli's Theorem} \\
&= \int_{\Omega} \int_0^{X_n \wedge M} X_n^+ \cdot p \lambda^{p-2} \, d\lambda \, dP \quad (A = \{X_n \wedge M \geq \lambda\}) \\
&= \int_{\Omega} X_n^+ \cdot \left(\frac{p}{p-1}\right) \cdot (X_n \wedge M)^{p-1} \, dP \\
&= \left(\frac{p}{p-1}\right) \int_{\Omega} X_n^+ \cdot (X_n \wedge M)^{p-1} \, dP \quad \downarrow \text{(By Hölder's inequality)} \\
&\leq \frac{p}{p-1} \cdot \left[E[(X_n^+)^p]\right]^{\frac{1}{p}} \cdot \left[E[(X_n \wedge M)^{p \cdot \frac{p-1}{p}}]\right]^{\frac{1}{p-1}} \quad \|fg\|_1 \leq \|f\|_p \|g\|_q \quad \left(\frac{1}{p} + \frac{1}{q} = 1\right) \\
&\quad f = X_n^+ \quad g = (X_n \wedge M)^{p-1} \\
&= \left(\frac{p}{p-1}\right) \cdot \left[E[(X_n^+)^p]\right]^{\frac{1}{p}} \cdot \left[E[(X_n \wedge M)^p]\right]^{\frac{1}{p-1}} \quad (\because p \cdot \frac{p-1}{p} = p-1 \Rightarrow p \cdot \frac{p-1}{p} = p-1)
\end{aligned}$$

$$\begin{aligned}
\text{So LHS} &= E[(X_n \wedge M)^p] \\
&\leq \text{RHS} = \left(\frac{p}{p-1}\right) \left[E[(X_n^+)^p]\right]^{\frac{1}{p}} \left[E[(X_n \wedge M)^p]\right]^{\frac{1}{p-1}}
\end{aligned}$$

$$\therefore \left[E[(X_n \wedge M)^p]\right]^{\frac{1}{p-1}} \leq \left(\frac{p}{p-1}\right) \left[E[(X_n^+)^p]\right]^{\frac{1}{p}}$$

$$\therefore \left[E[(X_n \wedge M)^p]\right]^{\frac{p}{p-1}} \leq \left(\frac{p}{p-1}\right)^p \cdot E[(X_n^+)^p]$$

$$\uparrow \\
\left(\frac{1}{p} + \frac{1}{q} = 1 \Rightarrow 1 + \frac{p}{q} = p \Rightarrow p - \frac{p}{q} = 1\right)$$

$$\therefore \text{LHS} = E[(\bar{X}_n \wedge M)^p]$$

$$\leq \text{RHS} = \left(\frac{p}{p-1}\right)^p \cdot E[(X_n^+)^p] \quad (\text{for all } M > 0)$$

By taking $M \nearrow \infty$, and by monotone convergence theorem, $\text{LHS} \nearrow E[(\bar{X}_n)^p]$

$$\text{So we have } E[(\bar{X}_n)^p] \leq \left(\frac{p}{p-1}\right)^p E[(X_n^+)^p] \quad \square$$

52 (LP-convergence theorem)

$\{X_n\}_{n \geq 1}$ martingale $\sup_{n \geq 1} E|X_n|^p < \infty$ (p>1)

Then $X_n \rightarrow X$ (a.s) and $X_n \xrightarrow{L^p} X$.

(proof)

Step 1 $\sup_{n \geq 1} E[X_n^p] < \infty$

because $(E[X_n^p])^p \leq (E|X_n|^p)^p \leq E[X_n^{p^2}]$
 $\uparrow$
 (Jensen's inequality)

$\leq \sup_{n \geq 1} E|X_n|^p < \infty$

So we have $\sup_{n \geq 1} E[X_n^p] < \infty$.

By martingale convergence theorem, (**34**)

$X_n \rightarrow X$ (a.s) $(\because \sup_{n \geq 1} E[X_n^p] < \infty)$
 $\{X_n\}_{n \geq 1}$ martingale
 $\Rightarrow$ submartingale

Step 2 Use L^p -maximal inequality (51)

$$Y_n \stackrel{\text{def}}{=} |X_n|$$

$$E[(Y_n)^p] \leq \left(\frac{p}{p-1}\right)^p E[(Y_n)^p]$$

||

$$\begin{aligned} E\left[\left(\max_{1 \leq m \leq n} |X_m|\right)^p\right] &\leq \left(\frac{p}{p-1}\right)^p E[|X_n|^p] \\ &\leq \left(\frac{p}{p-1}\right)^p \sup_{n \geq 1} E[|X_n|^p] < \infty \end{aligned}$$

By taking $n \rightarrow \infty$

$$\text{LHS} = E\left[\left(\sup_{n \geq 1} |X_n|\right)^p\right] < \infty$$

So $\sup_{n \geq 1} |X_n|$ is L^p -integrable.

Step 7 L.D.C.T

$$|X_n - X|^p \leq \left(2 \sup_{n \in \mathbb{N}} |X_n|\right)^p = 2^p \cdot \left(\sup_{n \in \mathbb{N}} |X_n|\right)^p \dots \text{integrable}$$

By Lebesgue's Dominated Convergence Theorem

$$\lim_{n \rightarrow \infty} E[|X_n - X|^p] = E\left[\lim_{n \rightarrow \infty} |X_n - X|^p\right] = 0$$

$$\stackrel{L^p}{=} X_n \rightarrow X \dots$$

Now the proof is complete $\square$

53

$$(1) \text{ Let } K_m \stackrel{\text{def}}{=} \mathbb{I}_{\{M < m \leq N\}} \geq 0$$

K_m is a predictable sequence

and K_m is bounded.

Hence, by [3], $(K \cdot X)_n$ is also a

submartingale.

$$(K \cdot X)_n = \sum_{m=1}^n K_m \cdot (X_m - X_{m-1})$$

$$= \begin{cases} 0 & (n \leq M) \\ X_n - X_M & (M < n \leq N) \\ X_N - X_M & (n > N) \end{cases}$$

$$= X_{n \wedge N} - X_{M \wedge N}$$

$$\text{So } E[(K \cdot X)_n | \mathcal{F}_{n-1}] \geq (K \cdot X)_{n-1} \quad (\text{submartingale})$$

$$\Rightarrow E[(K \cdot X)_n] \geq E[(K \cdot X)_{n-1}] \geq \dots \geq E[(K \cdot X)_1] \geq 0 \text{ as } E[X_1 - X_0] \geq 0$$

$$\geq 0 \quad (\because \{X_k\}_{k \geq 1} \text{ submartingale})$$

$$\therefore E[X_{M \wedge N}] \geq E[X_{M \wedge N}]$$

submartingale)

$n=k$... We have $E[X_N] \geq E[X_M]$

(2) We use the fact $\boxed{2}$:

$\left(\begin{array}{l} M \leq N \text{ and } A \in \mathcal{F}_M \\ \text{then } L \stackrel{\text{def}}{=} } \begin{cases} M & (w \in A) \\ N & (w \notin A) \end{cases} \right.$ is a stopping time

• By (1), $L \leq N \Rightarrow E[X_L] \leq E[X_N] \dots \textcircled{1}$

• $w \in A \Rightarrow L = M \Rightarrow E[X_L \cdot \mathbb{I}_A] = E[X_M \cdot \mathbb{I}_A] \dots \textcircled{2}$

• $w \in A^c \Rightarrow L = N \Rightarrow E[X_L \cdot \mathbb{I}_{A^c}] = E[X_N \cdot \mathbb{I}_{A^c}] \dots \textcircled{3}$

$\textcircled{1} - \textcircled{3} \Rightarrow E[X_L \cdot \mathbb{I}_A] \leq E[X_N \cdot \mathbb{I}_A] \dots \textcircled{4}$

$\textcircled{2} \& \textcircled{4} \Rightarrow E[X_M \cdot \mathbb{I}_A] \leq E[X_N \cdot \mathbb{I}_A] \dots \textcircled{5}$

$\textcircled{5} = \int_A X_M dP \leq \int_A X_N dP \quad \text{for all } A \in \mathcal{F}_M$
 $\qquad \qquad \qquad \int_A E[X_N | \mathcal{F}_M] dP \quad (\text{by definition})$

We should thus be $\mathcal{F}_M$ -measurable

Finally we show that X_M is a $\mathcal{F}_M$ -measurable function.

Let $c \in \mathbb{R}$. We show that $\{X_M > c\} \in \mathcal{F}_M$

$$\begin{aligned} & \{X_M > c\} \cap \{M = m\} \\ &= \underbrace{\{X_m > c\}}_{\in \mathcal{F}_m} \cap \underbrace{\{M = m\}}_{\in \mathcal{F}_m} \in \mathcal{F}_m \end{aligned}$$

So for $\forall m$ $\{X_m > c\} \cap \{M = m\} \in \mathcal{F}_m$

$\Rightarrow \{X_M > c\} \in \mathcal{F}_M$. $\therefore X_M$ is $\mathcal{F}_M$ -measurable

$$\forall A \in \mathcal{F}_M \quad \int_A X_M dP \leq \int_A E[X_M | \mathcal{F}_M] dP$$

$$\Rightarrow X_M \leq E[X_M | \mathcal{F}_M] \quad (a.s.)$$

54

$$A \stackrel{\text{def}}{=} \{\omega \in \Omega \mid \max_{1 \leq m \leq n} |S_m| > \lambda\}$$

$$N \stackrel{\text{def}}{=} \inf \{m \mid |S_m| > \lambda\} \wedge n \quad (\text{This is a stopping time})$$

Since $P(N \leq n) = 1$, we may apply 4.2

$$\begin{array}{ccc} E[S_1^2 - S_1^2] & \leq & E[S_N^2 - S_N^2] \\ \parallel & & \Downarrow \\ 0 & & \end{array}$$

$$\therefore 0 \leq E[(S_N^2 - S_N^2) \cdot \mathbb{I}_A + (S_N^2 - S_N^2) \cdot \mathbb{I}_A^c]$$

$$\bullet \omega \in A \Rightarrow \exists m = n \text{ such that } |S_m| > \lambda$$

$$\therefore |S_N| > \lambda \text{ but } |S_{N+1}| \leq \lambda \quad (* S_0 \stackrel{\text{def}}{=} 0)$$

$$\text{So we have } |S_N| \leq (\lambda + K) \quad (\because |X_i| \leq K)$$

$$\bullet \omega \in A^c \Rightarrow N = n \quad (\text{see definition of } N)$$

$$\text{And } |S_n| \leq \lambda$$

$$(0 \leq)$$

$$\text{Therefore } E[(S_N^2 - S_N^2) \cdot \mathbb{I}_A + (S_N^2 - S_N^2) \cdot \mathbb{I}_{A^c}]$$

$$\leq E[S_N^2 \cdot \mathbb{I}_A + (S_N^2 - S_N^2) \cdot \mathbb{I}_{A^c}]$$

$$\leq E[(x+k)^2 \cdot \mathbb{I}_A + (x^2 - \underbrace{S_N^2}_{=V[S_N]}) \cdot \mathbb{I}_{A^c}]$$

$$\therefore 0 \leq \underbrace{p(A) \cdot (x+k)^2 + (x^2 - V[S_N]) \cdot (1-p(A))}_{\parallel} \\ (x+k)^2 - (x+k)^2(1-p(A))$$

$$\therefore 0 \leq (x+k)^2 + \{x^2 - V[S_N] - (x+k)^2\}(1-p(A))$$

$$\Rightarrow \underbrace{(1-p(A)) \{ (x+k)^2 - x^2 + V[S_N] \}}_{\parallel} \leq (x+k)^2$$

$$\geq (1-p(A)) \cdot V[S_N]$$

$$\therefore \underbrace{(1-p(A))}_{\parallel} \leq \frac{(x+k)^2}{V[S_N]}$$

$$p(A^c) = p\left(\max_{1 \leq m \leq N} |S_m| \leq x\right)$$

55

Consider $E[Y_{n-1}(X_n - X_{n-1}) | \mathcal{F}_{n-1}]$

$$= Y_{n-1} \cdot \left(\underbrace{E[X_n | \mathcal{F}_{n-1}]}_{= X_{n-1}} - X_{n-1} \right) = 0$$

$$\therefore E \circ E[Y_{n-1}(X_n - X_{n-1}) | \mathcal{F}_{n-1}] = 0$$

⊗ By Cauchy Schwartz's Inequality

$$E[X_n Y_{n-1}]^2 \leq E[X_n^2] E[Y_{n-1}]^2$$

$< \infty$
So RHS finite

$$\therefore E[Y_{n-1}(X_n - X_{n-1})] = 0 \Rightarrow E[X_n Y_{n-1}] = E[X_{n-1} Y_{n-1}] \dots \textcircled{1}$$

Similarly we have $E[X_{n-1} Y_n] = E[X_{n-1} Y_{n-1}] \dots \textcircled{2}$

$$\begin{aligned} E[(X_n - X_{n-1})(Y_n - Y_{n-1})] &= E[X_n Y_n - \underbrace{X_{n-1} Y_n}_{= E[X_{n-1} Y_n]} - \underbrace{X_n Y_{n-1}}_{= E[X_{n-1} Y_{n-1}]} + X_{n-1} Y_{n-1}] \\ &= E[X_n Y_n] - E[X_{n-1} Y_{n-1}] \end{aligned}$$

$$\text{So } \sum_{m=1}^n E[(X_m - X_{m-1})(Y_n - Y_{n-1})] = \sum_{m=1}^n (E[X_m Y_m] - E[X_{m-1} Y_{m-1}])$$

And we have the desired result.

56

(1) $\{X_i\}_{i \in I}$... uniformly integrable

$$\lim_{M \rightarrow \infty} \sup_{i \in I} E[|X_i| \cdot \mathbf{I}_{\{|X_i| > M\}}] = 0$$

(2) We show that

$$\textcircled{1} \quad \lim_{M \rightarrow \infty} \sup_{i \in I} E[|X_i| \cdot \mathbf{I}_{\{|X_i| > M\}}] = 0$$

$$\Leftarrow \textcircled{2} \quad \bullet \sup_{i \in I} E|X_i| < \infty \quad (\text{Absolutely Continuous})$$

$$\bullet \lim_{P(A) \rightarrow 0} \sup_{i \in I} E[|X_i| \cdot \mathbf{I}_A] = 0$$

AGF

$$(\text{i.e. } \forall \varepsilon > 0 \exists \delta > 0 \text{ st } \forall A \in \mathcal{F} \quad P(A) < \delta$$

$$\sup_{i \in I} E[|X_i| \cdot \mathbf{I}_A] < \varepsilon.)$$

Step 1 $\textcircled{1} \Rightarrow \textcircled{2}$

Let $A \in \mathcal{F}$.

$$E|X_i| \cdot \mathbf{I}_A = E|X_i| \cdot \mathbf{I}_{\{\omega \in A \mid |X_i| > M\}} + E|X_i| \cdot \mathbf{I}_{\{\omega \in A \mid |X_i| \leq M\}}$$

$$\leq E|X_i| \cdot \mathbb{I}(|X_i| > M) + M \cdot P(A)$$

$$\leq \sup_{i \in \mathbb{I}} E|X_i| \cdot \mathbb{I}(|X_i| > M) + MP(A)$$

By taking large $M(\epsilon) > 0$, we have $\sup_{i \in \mathbb{I}} E|X_i| \cdot \mathbb{I}(|X_i| > M) < \frac{\epsilon}{2}$.

And then $\forall A \in \mathcal{F}$ $P(A) < \frac{\epsilon}{2M} \dots \sup_{i \in \mathbb{I}} E|X_i| \cdot \mathbb{I}_A < \epsilon$.

If $A = \Omega$, $\sup_{i \in \mathbb{I}} E|X_i| \leq \sup_{i \in \mathbb{I}} E|X_i| \cdot \mathbb{I}(|X_i| > M) + M < \infty$

So $\{X_i\}_{i \in \mathbb{I}}$ is L^1 -bounded.

Step 2 ② $\Rightarrow$ ①

$$\forall A \in \mathcal{F}: P(A) \leq p_\epsilon \quad \sup_{i \in \mathbb{I}} E[|X_i| \cdot \mathbb{I}_A] < \epsilon$$

Let $M_\epsilon = \frac{1}{p_\epsilon} \sup_{i \in \mathbb{I}} E|X_i| < \infty$. If $M > M_\epsilon$,

$$\begin{aligned} \text{Then } P(|X_i| > M) &\leq P(|X_i| > M_\epsilon) \leq \frac{1}{M_\epsilon} E|X_i| \\ &\leq \frac{1}{M_\epsilon} \sup_{i \in \mathbb{I}} E|X_i| = p_\epsilon \quad (\text{for all } i \in \mathbb{I}) \end{aligned}$$

Hence $\sup_{i \in \mathbb{I}} E[|X_i| \cdot \mathbb{I}(|X_i| > M)] < \epsilon$

This implies $\lim_{M \rightarrow \infty} \sup_{i \in \mathbb{I}} E[|X_i| \cdot \mathbb{I}(|X_i| > M)] = 0$ CHA SHIN

57

Step 1 We show " $\forall \epsilon > 0 \exists \delta > 0$ st $\forall A \in \mathcal{F}$

$$P(A) < \delta \Rightarrow \int_A |X| dP < \epsilon$$

(proof)

$$\begin{aligned} \int_A |X| dP &= \int \{w \in A \mid |X| > M\} |X| dP \\ &\quad + \int \{w \in A \mid |X| \leq M\} |X| dP \end{aligned}$$

$$\leq \underbrace{E[|X| \cdot \mathbb{I}\{|X| > M\}]}_{\downarrow} + M P(A)$$

By LDGT. $M \nearrow \infty$.

We have $< \frac{\epsilon}{2}$

Then if $P(A) < \frac{\epsilon}{2M} \Rightarrow \int_A |X| dP < \epsilon$.

[Step 2]

$$E[|E[X|\mathcal{F}]| \cdot \mathbb{I}_{\{|E[X|\mathcal{F}]| > M\}}] \xrightarrow{\text{Jensen's inequality}} E[E[|X|\mathcal{F}] \cdot \mathbb{I}_{\{|E[X|\mathcal{F}]| > M\}}]$$

$$= E[E[|X|\mathcal{F}] \cdot \mathbb{I}_A] \quad (A \stackrel{\text{def}}{=} \{|E[X|\mathcal{F}]| > M\})$$

$$= \int_A E[|X|\mathcal{F}] \, dP \quad (A \in \mathcal{F}) \quad \downarrow \text{By the definition of Conditional expectation.}$$

$$= \int_A E|X| \, dP$$

Be careful of the fact that $P(A)$

$$= P(E[|X|\mathcal{F}] > M) \leq \frac{1}{M} E|X| < \delta \epsilon \quad (\text{if } M > \frac{1}{\delta \epsilon} E|X|)$$

for all $A \in \mathcal{F}$

Hence by [Step 1], $\int_A E|X| \, dP < \epsilon$.

$$\text{So } \sup_{\mathcal{F}} E[|E[X|\mathcal{F}]| \cdot \mathbb{I}_{\{|E[X|\mathcal{F}]| > M\}}] < \epsilon$$

Now the proof is complete. $\square$

58

$$E[|X_i| \cdot \mathbb{I}\{|X_i| > M\}]$$

$$= E\left[\frac{|X_i|}{\phi(|X_i|)} \cdot \mathbb{I}\{|X_i| > M\} \cdot \phi(|X_i|)\right]$$

* By assumption, $\forall \varepsilon > 0 \exists M_\varepsilon$ s.t. $M > M_\varepsilon$

$$\frac{|X_i|}{\phi(|X_i|)} \mathbb{I}\{|X_i| > M\} < \frac{\varepsilon}{C}$$

$$(\because \frac{\phi(x)}{x} \nearrow \infty \text{ as } x \nearrow \infty)$$

$$< E\left[\frac{\varepsilon}{C} \phi(|X_i|)\right]$$

$$\leq \frac{\varepsilon}{C} \sup_{i \in I} E[\phi(|X_i|)]$$

$$\leq \frac{\varepsilon}{C}, C = \varepsilon$$

$$\therefore \lim_{M \nearrow \infty} \sup_{i \in I} E[|X_i| \cdot \mathbb{I}\{|X_i| > M\}] = 0$$

59 Suppose $\{X_n\}_{n \in \mathbb{N}}$ $X_n \xrightarrow{P} X$,

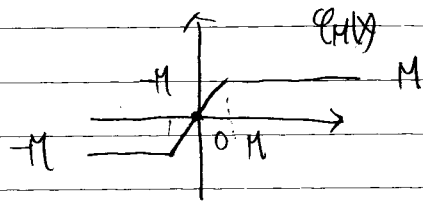
Now the following statements are equivalent.

- (1) $\{X_n\}_{n \in \mathbb{N}}$ Uniformly Integrable
 (2) $X_n \xrightarrow{L^1} X$
 (3) $E|X_n| \rightarrow E|X| < \infty$ ($\oplus$ We should assume $E|X| < \infty$ for all $n \in \mathbb{N}$)

(proof) (1) $\Rightarrow$ (2)

Consider a bounded continuous function $\varphi_M(x)$.

$$\varphi_M(x) = \begin{cases} M & x \geq M \\ x & |x| \leq M \\ -M & x \leq -M \end{cases}$$



By triangle inequality, we have

$$\begin{aligned} |X_n - X| &= |X_n - \varphi(X_n) + \varphi(X_n) - \varphi(X) + \varphi(X) - X| \\ &\leq \underbrace{|X_n - \varphi(X_n)|}_{(i)} + \underbrace{|\varphi(X_n) - \varphi(X)|}_{(ii)} + \underbrace{|\varphi(X) - X|}_{(iii)} \end{aligned}$$

(i) and (ii) :

$$|x - \varphi(x)| = \begin{cases} 0 & \dots |x| \leq M \\ |x| - M & \dots |x| > M \end{cases}$$

$$\leq \begin{cases} 0 & \dots |x| \leq M \\ |x| & \dots |x| > M \end{cases}$$

$$\leq |x| \cdot \mathbb{I}_{\{|x| > M\}}$$

Therefore (i) $\leq |x_n| \cdot \mathbb{I}_{\{|x_n| > M\}}$

(ii) $\leq |x| \cdot \mathbb{I}_{\{|x| > M\}}$

$$E[|x_n - x|] \leq E[|x_n| \cdot \mathbb{I}_{\{|x_n| > M\}}] + E[|x| \cdot \mathbb{I}_{\{|x| > M\}}] \\ + E[|\varphi(x_n) - \varphi(x)|]$$

And $\leq \sup_{n \in \mathbb{N}} E[|x_n| \cdot \mathbb{I}_{\{|x_n| > M\}}] + E[|x| \cdot \mathbb{I}_{\{|x| > M\}}] \\ + E[|\varphi(x_n) - \varphi(x)|]$

The first term $\rightarrow 0$ as $M \rightarrow \infty$ because $\{x_n\}_{n \in \mathbb{N}}$ uniformly integrable

$$(\because |X| \cdot \mathbb{I}_{\{|X| > M\}} \leq |X| \in L^1)$$

↑

• The second term $\rightarrow$ as $M \rightarrow \infty$ by LDCT

because $E|X| < \infty$.

(ch 4 [56])
(2)

(*) $\{X_n\}_{n \geq 1}$ uniformly integrable $\Rightarrow \sup_{n \geq 1} E[|X_n|] < \infty$,

$X_n \xrightarrow{P} X \Rightarrow |X_n| \xrightarrow{P} |X| \Rightarrow$ By Fatou's lemma (ch 2 [47])

$$E|X| \leq \liminf E[|X_n|] \leq \sup_{n \geq 1} E[|X_n|] < \infty$$

Hence $E|X| < \infty$

• Finally the third term $\rightarrow$ as $n \rightarrow \infty$

by bounded convergence theorem ($\because \phi(X_n) \xrightarrow{P} \phi(X)$
(ϕ continuous))

$$\text{So } \lim E[|X_n - X|] = 0 \quad \because \underbrace{X_n \xrightarrow{L^1} X}$$

(proof) ② $\Rightarrow$ ③ $\otimes$

$$|E[X_n] - E[X]| = |E[X_n - X]| \leq E[|X_n - X|] \rightarrow 0 \text{ (as } n \rightarrow \infty)$$

$$\text{So } E[X_n] \rightarrow E[X]$$

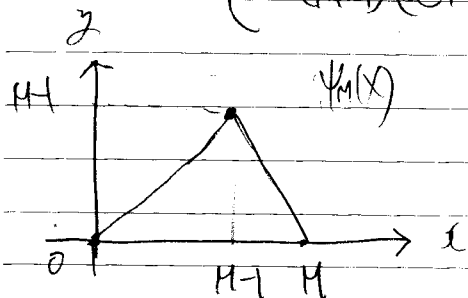
$\otimes$ " $X_n \xrightarrow{L} X$ " : We suppose this implies $E|X_n|, E|X| < \infty$

Hence $|E[X]| < \infty$ ($\Rightarrow E|X| < \infty$) and $\otimes$ is valid.

(proof) ③ $\Rightarrow$ ①

We consider a continuous and bounded function $\psi_M(x)$.

$$\psi_M(x) = \begin{cases} x & 0 \leq x \leq M-1 \\ 0 & M \leq x < \infty \\ -(M-1)(x-M) & M-1 \leq x \leq M \end{cases}$$



Step 1

$$X \geq 0 \Rightarrow \psi_M(X) \leq X \text{ and } \psi_M(X) \uparrow X \text{ (as } M \uparrow \infty).$$

$$\text{So } \psi_M(M) \uparrow M \text{ as } M \uparrow \infty$$

Hence by Monotone Convergence Theorem,

$$\lim_{M \uparrow \infty} E[\psi_M(X)] = E[X] \quad (\infty)$$

$$\forall \varepsilon > 0 \exists M(\varepsilon) > 0 \text{ st } E[X] - E[\psi_M(X)] < \frac{\varepsilon}{2}$$

Step 2 (Here $M = M(\varepsilon)$ given in Step 1.)

$$\text{Next, } 0 \leq \psi_M(|X_n|) \leq (M-1) \dots \text{ bounded}$$

$$\text{and } \psi_M(\cdot) \text{ is continuous } X_n \xrightarrow{P} X \Rightarrow |X_n| \xrightarrow{P} |X| \Rightarrow$$

$$\Rightarrow \psi_M(|X_n|) \xrightarrow{P} \psi_M(|X|) \Rightarrow E[\psi_M(|X_n|)] \rightarrow E[\psi_M(|X|)]$$

$\hookrightarrow$ Bounded Convergence Theorem in Ch 2. 4.11

Step 3

$$\text{Finally we consider } E[X_n \cdot \mathbb{I}_{\{|X_n| \leq M\}}]$$

$$= E[|X_n|] - E[|X_n| \cdot \mathbb{I}_{\{|X_n| > M\}}] \leq E[|X_n|] - E[\psi_M(|X_n|)]$$

$$\begin{cases} 0 \leq X \leq M \dots \\ X \cdot \mathbb{I}_{\{X \leq M\}} \geq \psi_M(X) \\ X > M \dots \\ X \cdot \mathbb{I}_{\{X > M\}} = \psi_M(X) = 0 \end{cases}$$

$$E[X_n] - E[\psi_M(X_n)]$$

$$\leq E[X_n] - E[\psi_M(X)] + \frac{\varepsilon}{4}$$

$$\left(\begin{array}{l} \because \text{By Step 2} \quad \forall \varepsilon > 0 \quad \exists N_1(M, \varepsilon) > 0, \text{ st} \\ \forall n > N_1 \quad |E[\psi_M(X_n)] - E[\psi_M(X)]| < \frac{\varepsilon}{4} \end{array} \right)$$

$$\left(\begin{array}{l} \textcircled{*} N_1(M, \varepsilon) \text{ is determined by } M, \varepsilon. \text{ But} \\ M \text{ is also determined by } \varepsilon > 0. \\ N_1 \text{ actually only depends on } \varepsilon > 0 \end{array} \right)$$

$$\leq E[X] - E[\psi_M(X)] + \frac{\varepsilon}{4} + \frac{\varepsilon}{4}$$

$$\left(\begin{array}{l} \because E[X_n] \rightarrow E[X] \Rightarrow \exists N_2(\varepsilon) > 0 \text{ such that} \\ \forall n > N_2(\varepsilon) \quad |E[X] - E[X_n]| < \frac{\varepsilon}{4} \\ \text{So we take } n > N \stackrel{\text{def}}{=} \max\{N_1(\varepsilon), N_2(\varepsilon)\} \end{array} \right)$$

$$< \frac{\varepsilon}{2} + \frac{\varepsilon}{2} \quad (\text{by Step 1})$$

$$\text{Hence } \sup_{n > N(\varepsilon)} \{E[X_n \cdot \mathbb{I}_{\{X_n > L\}}]\} < \varepsilon \quad (\forall L > M(\varepsilon))$$

Moreover, for $i=1 \sim N(\epsilon)$,

By taking large $K_1(\epsilon), K_2(\epsilon), \dots, K_N(\epsilon)$,

$$\text{We have } \begin{cases} \cdot E[|X_i| \cdot \mathbb{I}_{\{|X_i| > K_i\}}] < \epsilon \\ \cdot E[|X_{N(\epsilon)}| \cdot \mathbb{I}_{\{|X_{N(\epsilon)}| > K_{N(\epsilon)}\}}] < \epsilon \end{cases} \quad \left(\begin{array}{l} \text{So we should} \\ \text{assume } E|X_n| < \infty \\ \text{for all } n \in \mathbb{N} \end{array} \right)$$

(C.L.P.G.T)

$$\text{Let } K = \max\{K_1(\epsilon), \dots, K_{N(\epsilon)}\}$$

$$\text{Then } \forall L > \max\{K(\epsilon), M(\epsilon)\}$$

$$\sup_{n \in \mathbb{N}} E[|X_n| \cdot \mathbb{I}_{\{|X_n| > L\}}] < \epsilon$$

So $\{X_n\}_{n \in \mathbb{N}}$ is uniformly integrable. ■

[60] Let $\{X_n\}_{n \geq 1}$ - submartingale

① $\Rightarrow$ ②

($\because$ [56], (2))

$\{X_n\}_{n \geq 1}$: Uniformly Integrable $\Rightarrow \sup_{n \geq 1} E|X_n| < \infty$

$\Rightarrow \sup_{n \geq 1} E[X_n^+] < \infty \Rightarrow X_n \rightarrow X$ (a.s.) ($\because$ [34])

$E|X| < \infty$.

Next, $X_n \rightarrow X$ (a.s.) $\Rightarrow X_n \xrightarrow{P} X$

By [59], $X_n \xrightarrow{P} X$ and $\{X_n\}_{n \geq 1}$: Uniformly Integrable

$\Rightarrow X_n \xrightarrow{L} X$

② $\Rightarrow$ ③

This is trivial

③ $\Rightarrow$ ①

$X_n \xrightarrow{L} X \Rightarrow X_n \xrightarrow{P} X$ and By [59], $\{X_n\}_{n \geq 1}$

is uniformly integrable

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$$|E[X_n \cdot I_A] - E[X \cdot I_A]| \leq E[|X_n - X| \cdot I_A]$$

$$\leq E[|X_n - X|] \rightarrow 0 \quad (\text{as } n \rightarrow \infty)$$

$$(\because |X_n - X| \cdot I_A \leq |X_n - X|)$$

[62] Let $A \in \mathcal{F}_n$ (an arbitrary $\mathcal{F}_n$ -measurable set)

- $E[X_{n+1} | \mathcal{F}_n] = X_n$ ($\because$ martingale)
- $E[X_{n+2} | \mathcal{F}_n] = E[E[X_{n+2} | \mathcal{F}_{n+1}] | \mathcal{F}_n] = E[X_{n+1} | \mathcal{F}_n] = X_n$
 (By [9]) X_{n+1} ($\because$ martingale)

So similarly $E[X_m | \mathcal{F}_n] = X_n$ (for all $m \geq n+1$)

$$\Rightarrow \int_A E[X_m | \mathcal{F}_n] dP = \int_A X_n dP \quad \dots \textcircled{1}$$

($m \geq n+1$)

Moreover,

$$\int_A E[X_m | \mathcal{F}_n] dP = \int_A X_m dP \quad (A \in \mathcal{F}_n; \text{ By the definition of conditional expectation})$$

$\dots \textcircled{2}$

By $\textcircled{1}$ and $\textcircled{2}$ $\int_A X_m dP = \int_A X_n dP$ ($\forall m \geq n+1$) $\dots \textcircled{3}$

And [61] implies $\int_A X_n dP \rightarrow \int_A X dP$

But $\textcircled{3}$ implies $\int_A X_n dP = \int_A X dP$

In conclusion, $\forall A \in \mathcal{F}_n$ $\int_A X dP = \int_A E[X | \mathcal{F}_n] dP = \int_A X_n dP$

So $E[X | \mathcal{F}_n] = X_n$ by uniqueness of conditional expectation ([9] (3)).

[63] $\{X_n\}_{n \geq 1}$ martingale Then ① ~ ④ are equivalent

① $\Rightarrow$ ②

$\{X_n\}_{n \geq 1}$ martingale $\rightarrow$ submartingale

$\Rightarrow$ By [60]: this immediately holds

② $\Rightarrow$ ③

Similarly, by [60], this also holds

③ $\Rightarrow$ ④

By [62], $X_n \xrightarrow{L} X \Rightarrow X_n = E[X | \mathcal{F}_n]$

↓
(We assume $X_n \xrightarrow{L} X$ implies
 X is integrable.)

So ④ holds

④ $\Rightarrow$ ①

$$X_n = E[X | \mathcal{F}_n] \quad (X: \text{integrable})$$

$$\{X_n\}_{n \in \mathbb{N}} = \{E[X | \mathcal{F}_n]\}_{n \in \mathbb{N}} \quad \dots \text{ By } \boxed{57}$$

this is uniformly integrable

So ① holds $\blacksquare$

$$\boxed{64} \quad \mathcal{F}_n \uparrow \mathcal{F}_\infty (= \sigma[\bigcup_{n=1}^{\infty} \mathcal{F}_n]) \quad (\text{and } E|X| < \infty)$$

We show that $E[X|\mathcal{F}_n] \xrightarrow{a.s.} E[X|\mathcal{F}]$ and
 $E[X|\mathcal{F}_n] \xrightarrow{L^1} E[X|\mathcal{F}]$.

(proof)

$$\text{Let } Y_n = E[X|\mathcal{F}_n]$$

$$\text{By [19] (ch.4), } E[E[X|\mathcal{F}_{n+1}]|\mathcal{F}_n] = E[X|\mathcal{F}_n]$$

$$\Rightarrow E[Y_{n+1}|\mathcal{F}_n] = Y_n$$

Hence $\{Y_n\}_{n=1}^{\infty}$ is a martingale.

And also by [57], $\{Y_n\}_{n=1}^{\infty}$ is uniformly integrable.

By [63], $\{Y_n\}_{n=1}^{\infty}$ is a uniformly integrable martingale.

$$\Rightarrow Y_n \rightarrow Y \text{ (a.s.) and (L}^1\text{)}$$

By [62], $\{Y_n\}_{n=1}^{\infty}$ is a martingale and $Y_n \xrightarrow{L^1} Y$

$$\Rightarrow Y_n = E[Y|\mathcal{F}_n] \quad \text{--- } \otimes$$

Now consider a family of sets $\mathcal{D}$.

$$\mathcal{D} \stackrel{\text{def}}{=} \{A \in \mathcal{F}_\infty \mid \int_A X dP = \int_A Y dP\}$$

• $\mathcal{F}_n \subseteq \mathcal{D}$

(proof)

$$A \in \mathcal{F}_n \Rightarrow \int_A X dP = \int_A E[X | \mathcal{F}_n] dP$$

(By definition of conditional expectation)

$$= \int_A Y_n dP = \int_A E[Y | \mathcal{F}_n] dP = \int_A Y dP$$

$(Y_n = E[X | \mathcal{F}_n])$ (By $\circledast$)

(By definition of conditional expectation)

So $\int_A X dP = \int_A Y dP$ for all $A \in \mathcal{F}_n$

Therefore $\mathcal{F}_n \subseteq \mathcal{D}$

• $\mathcal{D}$ is a Dynkin-system

(proof)

• $\Omega \in \mathcal{D} \quad (\because \mathcal{F}_n \in \mathcal{D} \text{ and } \Omega \in \mathcal{F}_n \rightarrow \Omega \in \mathcal{D})$

• $A, \tilde{A} \in \mathcal{D} \quad (A \subseteq \tilde{A})$

$$\int_A X dP = \int_A Y dP \quad (\text{finite}) \quad (\because X \text{ integrable})$$

$$\int_{\tilde{A}} X dP = \int_{\tilde{A}} Y dP \quad (\text{finite})$$

$$\therefore \int_{\tilde{A} \setminus A} X dP = \int_{\tilde{A} \setminus A} Y dP$$

$$\therefore \tilde{A} \setminus A \in \mathcal{D}$$

• $\{A_n\}_{n \in \mathbb{N}} \subset \mathcal{D} \quad A_n \uparrow A$

$$\int_{A_n} X dP = \int_{\Omega} X \cdot \mathbb{I}_{A_n} dP = \int_{A_n} Y dP = \int_{\Omega} Y \cdot \mathbb{I}_{A_n} dP$$

$$\text{Since } |X \cdot \mathbb{I}_{A_n}| \leq |X| \cdot \mathbb{I}_{A_n} \text{ integrable} \\ |Y \cdot \mathbb{I}_{A_n}| \leq |Y| \cdot \mathbb{I}_{A_n} \text{ integrable}$$

$$X \cdot \mathbb{I}_{A_n} \uparrow X \cdot \mathbb{I}_A, \quad Y \cdot \mathbb{I}_{A_n} \uparrow Y \cdot \mathbb{I}_A$$

$$\text{By LDCGT} \quad \lim_{n \rightarrow \infty} \int_{\Omega} X \cdot \mathbb{I}_{A_n} dP = \int_{\Omega} X \cdot \mathbb{I}_A dP = \int_A X dP$$

$$= \lim_{n \rightarrow \infty} \int_{\Omega} Y \cdot \mathbb{I}_{A_n} dP = \int_{\Omega} Y \cdot \mathbb{I}_A dP = \int_A Y dP$$

$$\therefore \int_A X dP = \int_A Y dP \Rightarrow A \in \mathcal{D}$$

$\therefore \mathcal{D}$ is a Dynkin-system

$$\underline{\mathcal{F}_{\infty} \subseteq \mathcal{D}}$$

(proof)

$$\mathcal{F}_n \subseteq \mathcal{D} \text{ (for all } n \in \mathbb{N})$$

$$\Rightarrow \bigcup_{n=1}^{\infty} \mathcal{F}_n \subseteq \mathcal{D}$$

$$\Rightarrow \mathcal{F}[\bigcup_{n=1}^{\infty} \mathcal{F}_n] \subseteq \mathcal{D}$$

($\mathcal{F}(\cdot)$: the smallest Dynkin system containing $\cdot$)

And $\bigcup_{n=1}^{\infty} \mathcal{F}_n$ is a π -system.

$$(\mathcal{F}_n \subseteq \mathcal{F}_m, A, B \in \bigcup_{n=1}^{\infty} \mathcal{F}_n \Rightarrow A \in \mathcal{F}_{n_1}, B \in \mathcal{F}_{n_2}$$

$$n_1 < n_2 \Rightarrow A, B \in \mathcal{F}_{n_2} \Rightarrow A \cap B \in \mathcal{F}_{n_2}$$

$$\Rightarrow A \cap B \in \bigcup_{n=1}^{\infty} \mathcal{F}_n)$$

$$\therefore \mathcal{F}\left[\bigcup_{n=1}^{\infty} \mathcal{F}_n\right] = \mathcal{G}\left[\bigcup_{n=1}^{\infty} \mathcal{F}_n\right] (= \mathcal{F}_{\infty})$$

(By π - λ theorem)

$$\Rightarrow \underline{\mathcal{F}_{\infty} \subseteq \mathcal{L}}$$

Finally, $\int_A X dP = \int_A Y dP \quad (\forall A \in \mathcal{F}_{\infty})$

"

 $\downarrow$

 (by the definition of conditional expectation)

 $\int_A E[X | \mathcal{F}_{\infty}] dP$

Y is also $\mathcal{F}_{\infty}$ -measurable

($\because Y_n$ is $\mathcal{F}_n$ -measurable $\Rightarrow \mathcal{F}_{\infty}$ -measurable

So, $\underset{(43)}{Y_n \rightarrow Y} \Rightarrow Y$ is also $\mathcal{F}_{\infty}$ -measurable)

So by uniqueness of conditional expectation

$$Y = E[X | \mathcal{F}_{\infty}] \quad (43)$$

[65] (Levy's 0-1 law)

This is just a special version of [64]

Let $X \sim \mathbb{I}_A$ and $A \in \mathcal{F}_\infty$ where

$$\mathcal{F}_n \nearrow \mathcal{F}_\infty \quad (\mathcal{F}_\infty = \sigma[\bigcup_{n=1}^{\infty} \mathcal{F}_n])$$

$$\text{By [64], } E[\mathbb{I}_A | \mathcal{F}_n] \xrightarrow{\text{a.s./L}} E[\mathbb{I}_A | \mathcal{F}_\infty]$$

$$\parallel$$

$$\mathbb{I}_A$$

66 (Dominated Convergence Theorem for conditional expectation)

$Y_n \rightarrow Y$ (a.s.) and $|Y_n| \leq Z$ for all $n \in \mathbb{N}$

Where $E[Z] < \infty$. And suppose $\mathcal{F}_n \uparrow \mathcal{F}_\infty = \sigma\left[\bigcup_{n=1}^{\infty} \mathcal{F}_n\right]$.

We show that " $E[Y_n | \mathcal{F}_n] \rightarrow E[Y | \mathcal{F}_\infty]$ (a.s.)"

(proof). (Our first goal is to show that $\limsup_{n \rightarrow \infty} E[|Y_n - Y| | \mathcal{F}_n] = 0$ (a.s.))

Let $W_N \stackrel{\text{def}}{=} \sup_{n, m \geq N} E[|Y_n - Y_m|]$.

$W_N \leq 2Z$ Hence $E[W_N] < \infty$

Now fix N , and suppose $n \geq N$.

$$|Y_n - Y| = \lim_{k \rightarrow \infty} |Y_n - Y_k| \leq \sup_{k \geq N} |Y_k - Y| = W_N$$

Thus $E[|Y_n - Y| | \mathcal{F}_n] \leq E[W_N | \mathcal{F}_n]$

By taking $\limsup$, we have:

$$\limsup_{n \rightarrow \infty} E[|Y_n - Y| | \mathcal{F}_n] \leq \lim_{n \rightarrow \infty} E[W_N | \mathcal{F}_n]$$

By 64, (W_N is integrable), $\text{RHS} = E[W_N | \mathcal{F}_\infty]$.

Now taking $N \rightarrow \infty$, $W \rightarrow 0$ (a.s).

because $\{Y_n(w)\}_{n \geq 1}$ converges almost surely

$\Rightarrow \{Y_n(w)\}_{n \geq 1}$ is a Cauchy sequence almost surely.

$$\Rightarrow "N \wedge \infty \Rightarrow \infty \wedge 0 \text{ (ad)}"$$

Further more, this implies that

$$N \uparrow \infty \quad E[W_N | \mathcal{F}_{n_0}] \downarrow 0 \quad (45) \quad \dots \quad \textcircled{A}$$

* To prove $(*)$ we recall that $[5] (3)$.

Consider $X_n \geq 0$ $X_n \downarrow 0$ $E[X_1] < \infty$

Let $Y_n \stackrel{\text{def}}{=}} X_1 - X_n$. Then $Y_n \geq 0$ $Y_n \uparrow X_1$.

By (5) (3) $\lim_{n \rightarrow \infty} E[n|F] = E[X|F] \quad (a.s.)$

$$\lim_{n \rightarrow \infty} E[X_1 - X_n | \mathcal{F}]$$

$$\lim_{n \rightarrow \infty} (E[X_1 | \mathcal{F}] - E[X_n | \mathcal{F}])$$

So $\lim_{n \rightarrow \infty} E[X_n | \mathcal{F}] = 0$ (a.s.)

$W_N \geq 0$. $W_N \downarrow 0$ (a.s.) and $W_1 \leq 2Z \Rightarrow E[W_1] < \infty$

So $E[W_N | \mathcal{F}_\infty] \downarrow 0$ (a.s.).

Therefore $\limsup_{n \rightarrow \infty} E[|Y_n - Y| | \mathcal{F}_n] = 0$ (a.s.)

$$\begin{aligned}
 \text{Finally, } & \limsup_{n \rightarrow \infty} |E[Y | \mathcal{F}_n] - E[Y | \mathcal{F}_\infty]| \\
 &= \limsup_{n \rightarrow \infty} |E[Y | \mathcal{F}_n] - E[Y_n | \mathcal{F}_n] + E[Y_n | \mathcal{F}_n] - E[Y | \mathcal{F}_\infty]| \\
 &\leq \underbrace{\limsup_{n \rightarrow \infty} E[|Y - Y_n| | \mathcal{F}_n]}_{\rightarrow 0 \text{ (a.s.)}} + \underbrace{\limsup_{n \rightarrow \infty} |E[Y_n | \mathcal{F}_n] - E[Y | \mathcal{F}_\infty]|}_{\text{By (64)} \rightarrow 0 \text{ (a.s.)}}
 \end{aligned}$$

Now the proof is complete. ~~Q.E.D.~~

[67] $X_n: \Omega \rightarrow [0, \infty)$ (r.v)

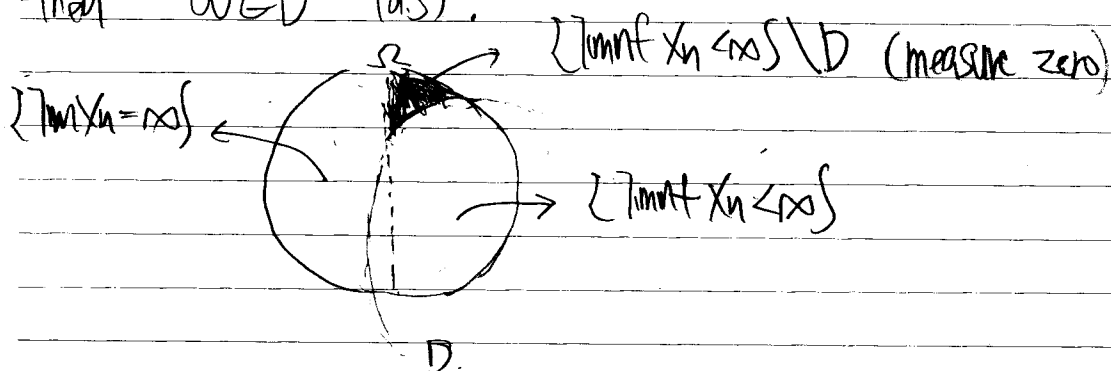
$$D = \{X_n = 0 \text{ for some } n \in \mathbb{N}\}$$

$$P(D | X_1, \dots, X_n) \equiv P(X) > 0 \text{ (a.s.) on } \{X_n \leq 2\}$$

We show that $P(D \cup \{\liminf X_n = \infty\}) = 1$.

(proof)

We try to show that if $\omega \in \{\liminf X_n < \infty\}$ then $\omega \in D$ (a.s.).



This will prove the statement of [67]

Step 1

Let $\mathcal{F}_n = \sigma[X_1, X_2, \dots, X_n]$ and $\mathcal{F}_\infty = \sigma[\bigcup_{n=1}^{\infty} \mathcal{F}_n]$.

$$D = \{X_n = 0 \text{ for some } n \geq 1\} \in \bigcup_{n=1}^{\infty} \mathcal{F}_n \subseteq \sigma[\bigcup_{n=1}^{\infty} \mathcal{F}_n] = \mathcal{F}_\infty$$

So $D \in \mathcal{F}_\infty$.

By Levy's 0-1 theorem [65],

$$\mathcal{F}_n \uparrow \mathcal{F}_\infty \text{ and } D \in \mathcal{F}_\infty \Rightarrow E[\mathbb{I}_D | \mathcal{F}_n] \rightarrow \mathbb{I}_D \text{ (a.s.)}$$

$$\Rightarrow P(D | X_1, \dots, X_n) \rightarrow \mathbb{I}_D \text{ (a.s.)}$$

Step 2

Now let $w \in \{\liminf X_n \leq M\} \quad (M < \infty)$

$$\Rightarrow w \in \{X_n \leq M \text{ i.o.}\}$$

$$(\because w \in \{X_n \leq M \text{ i.o.}\}^c \Rightarrow w \in \{X_n \leq M \text{ only for finite } n\})$$

$$\subseteq \{X_n > M \text{ for infinitely large } n\} \subseteq \{\liminf X_n > M\}$$

$$= \{\liminf X_n \leq M\}^c)$$

And $w \in \{X_n \leq M \text{ i.o.}\}$

$$\Rightarrow \{P(D | X_1, \dots, X_n) \geq \delta(M) \text{ i.o.}\} \text{ (a.s.)}$$

(Because $P(D | X_1, \dots, X_n) \geq \delta(M)$ (a.s.) on $\{X_n \leq M\}$)

$$= \{ E[I_D | \mathcal{F}_n] \geq f(M) : 0 \}$$

$$\subseteq \{ \limsup E[I_D | \mathcal{F}_n] \geq f(M) \}$$

By the conclusion of Step 1

$$\lim_{n \rightarrow \infty} P(D | X_1, \dots, X_n) = \lim_{n \rightarrow \infty} E[I_D | \mathcal{F}_n] = I_D \quad (a.s.)$$

$$\therefore \omega \in \{ \liminf X_n \leq M \} \Rightarrow \omega \in \{ I_D(\omega) \geq f(M) \} \quad (a.s.)$$

$$\text{Since } f(M) > 0, \omega \in \{ I_D(\omega) \geq f(M) \}$$

$$= \{ I_D(\omega) = 1 \}$$

$$= D$$

$$\forall M < \infty \quad \omega \in \{ \liminf X_n \leq M \} \Rightarrow \omega \in D \quad (a.s.)$$

$$\text{Therefore } \omega \in \bigcup_{M=1}^{\infty} \{ \liminf X_n \leq M \} \Rightarrow \omega \in D \quad (a.s.)$$

$$\parallel$$

$$\{ \liminf X_n < \infty \}$$

So the proof is complete $\square$

[68] Show that $\mathcal{F}_n \uparrow \mathcal{F}_\infty$ ($\mathcal{F}_n, \mathcal{F}_\infty$: σ -algebras)

and $Y_n \xrightarrow{L^1} Y \Rightarrow E[Y_n | \mathcal{F}_n] \xrightarrow{L^1} E[Y | \mathcal{F}_\infty]$

(proof)

Our goal is to show $\limsup_{n \rightarrow \infty} E |E[Y_n | \mathcal{F}_n] - E[Y | \mathcal{F}_\infty]|$

By triangle inequality,

$$\begin{aligned} & \limsup E |E[Y_n | \mathcal{F}_n] - E[Y | \mathcal{F}_\infty] + E[Y | \mathcal{F}_n] - E[Y | \mathcal{F}_\infty]| \\ & \leq \limsup E |E[Y_n | \mathcal{F}_n] - E[Y | \mathcal{F}_n]| \quad \text{--- ①} \\ & + \limsup E |E[Y | \mathcal{F}_n] - E[Y | \mathcal{F}_\infty]| \quad \text{--- ②} \end{aligned}$$

① $\dots = \limsup E |Y_n - Y| = 0 \quad (\because Y_n \xrightarrow{L^1} Y)$

② $\dots$ By [69], $E[Y | \mathcal{F}_n] \xrightarrow{L^1} E[Y | \mathcal{F}_\infty]$

$\therefore \text{②} \rightarrow 0.$

Now the proof is complete. ~~□~~

69

(1) Consider $\{X_n\}_{n \leq 0}$: a sequence of r.v.

- X_n is $\mathcal{F}_n$ -measurable. (adapted to $\mathcal{F}_n$) ($\mathcal{F}_n \subseteq \mathcal{F}_{n+1}$)
- $E|X_n| < \infty$ ($n \leq 0$)
- $E[X_{n+1} | \mathcal{F}_n] = X_n$ for all $n \leq -1$

Then $\{X_n\}_{n \leq 0}$ is called a backward martingale.

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

We show that $\lim_{n \rightarrow -\infty} X_n$ exists

$$(X_{-\infty} \stackrel{\text{def}}{=} \lim_{n \rightarrow -\infty} X_n) \quad \text{and} \quad X_n \xrightarrow{L} X_{-\infty} \quad (\text{as } n \rightarrow -\infty)$$

Step 1

Consider upcrossing of $[a, b]$ by $\{X_{-n}, X_{-n+1}, \dots, X_1, X_0\}$

As we have shown in [33],

$$(b-a) E[U_n] \leq E[(X_0 - a)^+] - E[(X_{-n} - a)^+]$$

$$\leq E[(X_0 - a)^+] \leq E[X_0] + |a| < \infty$$

→ (MCT)

Similar to [33], $\lim_{h \rightarrow \infty} (b-a) E[U] = (b-a) E[U] < \infty$

(U : number of upcrossings of $[a, b]$ by $\{X_n\}_{n \geq 1}$)

$\therefore U < \infty$ (a.s.) $\Rightarrow$

$\omega \in \{X_n \leq a \text{ only for finite times}\} \cup \{X_n \geq b \text{ only for finite times}\}$

$\subseteq \{X_n > a \text{ for sufficiently large } n\} \cup \{X_n < b \text{ for sufficiently large } n\}$

$\subseteq \{\liminf X_n > a\} \cup \{\limsup X_n < b\}$

$\subseteq \{\liminf X_n \geq a\} \cup \{\limsup X_n \leq b\}$ (a.s.)

So $P((\{\liminf X_n \leq a\} \cup \{\limsup X_n \geq b\})^c) = 0$

$\therefore P(\liminf X_n < a < b < \limsup X_n) = 0$

$\therefore P(\bigcup_{\substack{a < b \\ a, b \in \mathbb{Q}}} \{\liminf X_n < a < b < \limsup X_n\}) = 0$

$\therefore P(\lim X_n \text{ does not exist}) = 0$

$\therefore \lim X_n \text{ exists (a.s.)}$

Step 2

We prove $X_n = E[X_0 | \mathcal{F}_n]$ for all $n \geq 0$.

Since $E[X_{n+1} | \mathcal{F}_n] = X_n$, by putting $n=1$

We have $E[X_1 | \mathcal{F}_1] = X_1$

Next $E[X_0 | \mathcal{F}_2] = E[\underbrace{E[X_0 | \mathcal{F}_1]}_{X_1} | \mathcal{F}_2]$

by [9] and $\mathcal{F}_2 \subseteq \mathcal{F}_1$.

So $E[X_0 | \mathcal{F}_2] = E[X_1 | \mathcal{F}_2] = X_2$ ($n=2$)

By induction, we have $E[X_0 | \mathcal{F}_n] = X_n$

Step 3

By Step 2 and [5], $\{X_n\}_{n \geq 1}$ is uniformly integrable.

By [5] $X_n \rightarrow X_\infty$ (a.s.) $\Rightarrow X_n \xrightarrow{P} X_\infty$

and $\{X_n\}_{n \geq 1}$ is uniformly integrable

$$\Rightarrow X_n \xrightarrow{L^1} X_\infty$$

[70] Suppose $\{X_n\}_{n \geq 1}$ is a backward martingale.

$$X_\infty = \lim_{n \rightarrow \infty} X_n \text{ and } \mathcal{F}_\infty = \bigcap_{n \geq 1} \mathcal{F}_n \text{ (--- this is given)}$$

We show that $X_\infty = E[X_0 | \mathcal{F}_\infty]$

[Step 1] X_∞ is $\mathcal{F}_\infty$ -measurable

Consider $\{X_\infty \geq a\}$.

$$= \{X_\infty \geq a\} = \left\{ \limsup_{n \rightarrow \infty} X_n \geq a \right\}$$

$$= \left\{ \limsup_{n \rightarrow \infty} \sup_{m \geq n} X_m \geq a \right\}$$

$$= \bigcap_{n \leq \infty} \underbrace{\left\{ \sup_{m \geq n} X_m \geq a \right\}}_{\mathcal{F}_n}$$

Here $\{X_\infty \geq a\} \in \mathcal{F}_n$ for all $n \leq \infty$

$$\Rightarrow \{X_\infty \geq a\} \in \bigcap_{n \leq \infty} \mathcal{F}_n$$

Step 2

$$X_n = E[X_0 | \mathcal{F}_n] \quad (\text{see [69]})$$

Let $A \in \mathcal{F}_\infty$.

$$\int_A X_n dP = \int_A E[X_0 | \mathcal{F}_n] dP = \int_A X_0 dP \quad \text{for all } n \leq 0$$

- $A \in \mathcal{F}_\infty \subseteq \mathcal{F}_n$ for all $n \in \mathbb{Z}$
- By the definition of conditional expectation

$$\int_A X_n dP = E[X_n \cdot I_A]$$

Since $X_n \xrightarrow{L} X_\infty$ (by [69]) and

$$\text{By [61], } E[X_n \cdot I_A] \rightarrow E[X_\infty \cdot I_A]$$

$$\text{But } \int_A X_n dP = E[X_n \cdot I_A] = \int_A X_0 dP \quad (\text{for all } n \leq 0)$$

So we find that $E[X_0 \cdot I_A] = E[X_\infty \cdot I_A]$.

Finally for $\forall A \in \mathcal{F}_{-\infty}$ we have

$$\int_A X_0 dP = \int_A X_{-\infty} dP.$$

||

$$\int_A E[X_0 | \mathcal{F}_{-\infty}] dP$$

↓ (by the definition of conditional expectation)

$$\text{Therefore, } X_{-\infty} = E[X_0 | \mathcal{F}_{-\infty}] \quad (a)$$

[7] Suppose $\mathcal{F}_{n+1} \subseteq \mathcal{F}_n$ ($n \leq 0$)

$$\mathcal{F}_\infty = \bigcap_n \mathcal{F}_n$$

We show that $E[Y|\mathcal{F}_n] \xrightarrow{\text{a.s./L}} E[Y|\mathcal{F}_\infty]$

(proof)

(Here we suppose $E|Y| < \infty$)

Let $X_n \stackrel{\text{def}}{=} E[Y|\mathcal{F}_n]$.

$$\begin{aligned} E[X_{n+1}|\mathcal{F}_n] &= E[E[Y|\mathcal{F}_{n+1}]|\mathcal{F}_n] \quad (\mathcal{F}_n \subseteq \mathcal{F}_{n+1}) \\ &= E[Y|\mathcal{F}_n] = X_n \quad (\text{by [9]}) \end{aligned}$$

So $\{X_n\}_{n \leq 0}$ is a backward martingale

$$\begin{aligned} \text{By [69] (2)} \quad X_n &\xrightarrow{\text{a.s.}} X_\infty \\ X_n &\xrightarrow{L} X_\infty \end{aligned}$$

$$\text{And by [70]} \quad X_\infty = E[X_0|\mathcal{F}_\infty]$$

$$= E[E[Y|\mathcal{F}_0]|\mathcal{F}_\infty]$$

$$= E[Y|\mathcal{F}_\infty] \quad \downarrow \text{ by [9]}$$

[72] $\{X_n\}_{n \geq 1}$ is a uniformly integrable submartingale.

We show that $\{X_{N \wedge n}\}_{n \geq 1}$ is uniformly integrable

where N is a stopping time.

[Step] $\sup_{n \geq 1} E[X_{N \wedge n}] < \infty$

- $\phi(x) = \max\{0, x\}$ is a convex and monotone increasing function. $\therefore X_n^+ = \phi(X_n)$ is also a submartingale (By [29])

- $N \wedge n$ is also a stopping time.

and $P(N \wedge n \leq n) = 1$

Applying [42] to $\{X_n^+\}$ and $N \wedge n$

we have $E[X_{N \wedge n}^+] \leq E[X_1^+]$

- $\{X_n\}_{n \geq 1}$: uniformly integrable $\Rightarrow \sup_{n \geq 1} E|X_n| < \infty$
 $\Rightarrow \sup_{n \geq 1} E[X_n^+] < \infty$ (By [56] (2))

$\therefore \sup_{n \geq 1} E[X_{N \wedge n}^+] < \infty$

Step 2 By martingale convergence theorem ([32], [34])

$$\underbrace{X_{nm}} \rightarrow \underbrace{X_N}_{\text{as}} \quad \text{and} \quad \underbrace{E|X_N|} < \infty.$$

(this is a submartingale)

Finally consider $\lim_{K \rightarrow \infty} \sup_{n \geq 1} E[|X_{nm}| \cdot \mathbb{I}_{\{|X_{nm}| > K\}}]$

$$E[|X_{nm}| \cdot \mathbb{I}_{\{|X_{nm}| > K\}}]$$

||

$$E[|X_{nm}| \cdot \mathbb{I}_{\{|X_{nm}| > K, N > n\}}] +$$

$$E[|X_{nm}| \cdot \mathbb{I}_{\{|X_{nm}| > K, N \leq n\}}]$$

||

$$E[|X_n| \cdot \mathbb{I}_{\{|X_n| > K, N > n\}}] +$$

$$E[|X_n| \cdot \mathbb{I}_{\{|X_n| > K, N \leq n\}}]$$

$$\leq \sup_{n \in \mathbb{N}} E[|X_n| \cdot \mathbb{I}_{\{|X_n| > K\}}] + E[|X_n| \cdot \mathbb{I}_{\{|X_n| > K\}}] \quad \text{not related to } n$$

$$\therefore \sup_{n \geq 1} E[|X_{nm}| \cdot \mathbb{I}_{\{|X_{nm}| > K\}}]$$

$$\leq \sup_{k \in \mathbb{N}} E[|X_n| \cdot \mathbb{I}_{\{|X_n| > k\}}] + E[|X_n| \cdot \mathbb{I}_{\{|X_n| \leq k\}}]$$

$$(E|X_n| < \infty)$$

By taking $k > 0$, we have $< \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon$

13

(1) $\{X_n\}_{n \geq 1}$... a uniformly integrable submartingale

N ... a stopping time

We show that $E[X_0] \leq E[X_N] \leq E[X_\infty]$

$$(X_\infty = \lim X_n)$$

(proof)

(Similar to 12)

Apply 12 to $\{X_n\}_{n \geq 1}$ and $\underline{N \wedge n}$

(This is also a stopping time
and $P(N \wedge n \leq n) = 1$)

Then we have $E[X_0] \leq E[X_{N \wedge n}] \leq E[X_n] \dots \circledast$

Recall [60] $\rightarrow$ $\left(\begin{array}{l} \{Y_n\}_{n \geq 1} \text{ submartingale} \\ \cdot \text{ Uniformly integrable} \Leftrightarrow \text{converge in } L^1 \end{array} \right)$

Since $\{X_{nm}\}, \{X_n\}$ are both submartingales ([32])
and are uniformly integrable ($\because$ [72]),
 $\{X_{nm}\}, \{X_n\}$ converges in L^1 .

This implies $E[X_{nm}] \rightarrow E[X_n]$ ($\because X_{nm} \xrightarrow{\text{a.s.}} X_n$ see [72])
 $E[X_n] \rightarrow E[X_\infty]$

By taking $n \rightarrow \infty$ to $\otimes$,

we have $E[X_0] \leq E[X_n] \leq E[X_\infty]$

(2) If $E|X_n| < \infty$ and $X_n \cdot \mathbb{I}_{\{N \geq n\}}$ is uniformly integrable.

Then $X_{n \wedge N}$ is also uniformly integrable and hence $E[X_0] \leq E[X_N]$

(proof)

$$\begin{aligned} & E[|X_{n \wedge N}| \cdot \mathbb{I}_{\{|X_{n \wedge N}| > M\}}] \\ &= E[|X_{n \wedge N}| \cdot \mathbb{I}_{\{|X_{n \wedge N}| > M, N \geq n\}}] \\ &+ E[|X_{n \wedge N}| \cdot \mathbb{I}_{\{|X_{n \wedge N}| > M, N \leq n\}}] \\ &= E[|X_n| \cdot \mathbb{I}_{\{|X_n| > M, N \geq n\}}] \quad \text{--- ①} \\ &+ E[|X_N| \cdot \mathbb{I}_{\{|X_N| > M, N \leq n\}}] \quad \text{--- ②} \end{aligned}$$

$$\begin{aligned} \text{①} &= E[|X_n| \cdot \mathbb{I}_{\{N \geq n\}} \cdot \mathbb{I}_{\{|X_n| \cdot \mathbb{I}_{\{N \geq n\}} > M\}}] \\ &\leq \sup_{n \geq 1} E[\cdot] \end{aligned}$$

$$\text{②} \leq E[|X_N| \cdot \mathbb{I}_{\{|X_N| > M\}}]$$

So by taking large $M > 0$ we have $< \frac{\epsilon}{2 + \epsilon}$

(If X_n is a submartingale)

X_{nN} is a uniformly integrable submartingale

By (1) $E[X_{0N}] \leq E[X_{nN}]$

So $E[X_0] \leq E[X_N]$.

(3) $\{X_n\}_{n \geq 1}$ is a non-negative supermartingale

$N(\leq \infty)$ is a stopping time

(proof) $N \wedge n$ is a stopping time and $P(N \wedge n \leq n) = 1$.

$-X_n$ is a submartingale

By [42], $E[-X_0] \leq E[-X_{N \wedge n}]$

$$\Rightarrow E[X_{N \wedge n}] \leq E[X_0] \text{ for all } n \geq 1$$

By Fatou's lemma

$$E[\liminf X_{N \wedge n}] \leq \liminf E[X_{N \wedge n}] \leq E[X_0]$$

$$\therefore E[X_N] \leq E[X_0]$$

[74] $\{X_n\}_{n \geq 1}$ submartingale, $E[|X_{n+1} - X_n| | \mathcal{F}_n] \leq B$ (a.s.)

N a stopping time $E[N] < \infty$

We show that $X_{N \wedge n}$ is uniformly integrable and hence $E[X_0] \leq E[X_N]$

[Step 1] $|X_{N \wedge n}| \leq |X_0| + \sum_{m=0}^{N \wedge n-1} |X_{m+1} - X_m| \cdot \mathbb{I}\{N > m\} \dots (*)$

because $X_{N \wedge n} = X_0 + \sum_{m=0}^{N \wedge n-1} (X_{m+1} - X_m)$ and

$$\begin{aligned} \text{so } |X_{N \wedge n}| &\leq |X_0| + \sum_{m=0}^{N \wedge n-1} |X_{m+1} - X_m| \\ &= |X_0| + \sum_{m=0}^{(N-1) \wedge n-1} |X_{m+1} - X_m| \\ &\leq |X_0| + \sum_{m=0}^{n-1} |X_{m+1} - X_m| \cdot \mathbb{I}\{N > m\} \\ &\leq |X_0| + \sum_{m=0}^{\infty} |X_{m+1} - X_m| \cdot \mathbb{I}\{N > m\} \end{aligned}$$

[Step 2] $E[|X_{m+1} - X_m| \cdot \mathbb{I}\{N > m\}]$

$$= E[E[|X_{m+1} - X_m| \cdot \mathbb{I}\{N > m\} | \mathcal{F}_m]] \quad , \quad \{N > m\} \in \mathcal{F}_m$$

$$= E[E[|X_{m+1} - X_m| | \mathcal{F}_m] \cdot \mathbb{I}\{N > m\}] \quad \text{by (20) and}$$

$$\leq E[B \cdot \mathbb{I}\{N > m\}] = B \cdot P(N > m)$$

Therefore Step $\otimes$:

$$\begin{aligned} E[\text{R.H.S.}] &= E[X_0] + \sum_{m=0}^{\infty} B \cdot P(N > m) \\ &= E[X_0] + B E[N] \quad \downarrow \text{ch2 [35]} \\ &< \infty \end{aligned}$$

$$|X_{N+m}| \leq \underbrace{|X_0| + \sum_{m=0}^{\infty} |X_{m+1} - X_m| \cdot \mathbb{I}\{N > m\}}_{\downarrow} = \underbrace{Y}_m$$

This is an integrable r.v not related to n .

$$\text{So } \lim_{M \rightarrow \infty} \sup_{n \geq 1} E[|X_{nm}| \cdot \mathbb{I}\{|X_{nm}| > M\}]$$

$$\leq \lim_{M \rightarrow \infty} E[Y \cdot \mathbb{I}\{Y > M\}] = 0 \quad (\because Y: \text{integrable})$$

$\therefore \{X_{nm}\}_{n \geq 1}$ is uniformly integrable