

① Square Integrable Martingales,
Local Martingales, Quadratic
Variation

(Ω, \mathcal{F}, P) , $\mathbb{F} = \{\mathcal{F}_t\}_{t \geq 0}$ given. Assume \mathbb{F}
satisfies the usual conditions

Let $X = \{X_t\}_{t \geq 0}$ be a right-continuous
martingale.

If $E[X_t^2] < \infty \quad \forall t \geq 0$, we say that
 X is square integrable

• Additionally, if $X_0 = 0$ a.s. we
say that $X \in M_2$, i.e. we define

$M_2 = \{X \mid X \text{ is a right-continuous, square integrable martingale with } X_0 = 0 \text{ a.s.}\}$

• If X is continuous we say

$X \in M_2^c = \{X \in M_2 \mid X \text{ is continuous}\}$

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Matrix Structure on M_2 :

For $X \in M_2$ define

$$\|X\|_x^2 = \mathbb{E}[X_x^2] \quad x \geq 0$$

$$\|X\| = \sum_{n=1}^{\infty} \frac{\|X\|_{n-1}}{2^n}$$

Claim: if we identify indistinguishable processes as being equal then $\|\cdot\|$ is a metric.

Pf.

$$\|X - Y\| = 0 \Leftrightarrow \|X - Y\|_n = 0 \quad \forall n$$

$$\Leftrightarrow X_n = Y_n \quad \text{s.d.} \quad \forall n$$

$$\Leftrightarrow X_x = \mathbb{E}[X_n | \mathcal{F}_x] = \mathbb{E}[Y_n | \mathcal{F}_x] = Y_x$$

$$\forall x \geq 0$$

$\Rightarrow X, Y$ indistinguishable since they are also right continuous.

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Claim: M_2 is a complete metric space under $\|\cdot\|$ and M_2^c is a closed sub-space of M_2 .

pf

$\{X_n\}_{n=1,2,3,\dots}$ Cauchy in $(M_2, \|\cdot\|)$

$\Rightarrow \{X_n(t)\}_{n=1,2,3,\dots}$ Cauchy in $L^2(\Omega, \mathcal{F}_t, P) \forall t$.

$\Rightarrow X_n(t) \rightarrow X(t)$ in $L^2(\Omega, \mathcal{F}_t, P)$

Now for $\Delta \leq t$, $A \in \mathcal{F}_\Delta \subseteq \mathcal{F}_t$ we have

$$0 = \lim_{n \rightarrow \infty} E[\mathbb{1}_A(X_n(t) - X(t))]$$

$$= \lim_{n \rightarrow \infty} E[\mathbb{1}_A(X_n(\Delta) - X(\Delta))]$$

$\Rightarrow X = \{X(t)\}_{t \geq 0}$ (collection of limits)

is a square integrable martingale.

\Rightarrow If satisfies usual conditions

implies we can take a right

continuous modification, starting at 0

since $X_n(0) = 0 \quad \forall n$

\Rightarrow Using Dominated Convergence $\|X^n - X\| \rightarrow 0$

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This, $(M_2, \|\cdot\|)$ is complete.

Now, Assume $\{X_n\}_{n=1, \dots} \in M_2^c$ converges to $X \in M_2$ in $\|\cdot\|$.

~~$\{X^n - X\}_{t \in T}$~~ is a martingale

$\Rightarrow |X^n - X|$ is a sub-martingale

$$\Rightarrow P\left(\sup_{t \in T} |X^n - X|_t > \epsilon\right) \leq \frac{1}{\epsilon^2} E[|X^n - X|_T^2]$$

$$= \frac{1}{\epsilon^2} \|X^n - X\|_T^2 \rightarrow 0$$

\Rightarrow along some $\{n_k\}_{k=1, 2, \dots}$ we have

$$P\left(\sup_{t \in T} |X^{n_k} - X|_t > \frac{1}{k}\right) \leq \frac{1}{2k}$$

\Rightarrow Borel-Cantelli gives $X^{n_k} \rightarrow X$

uniformly on $[0, T]$ almost surely.

$\therefore X$ is continuous on $[0, T]$
a.s. $\forall T > 0$, hence on $[0, \infty)$.

⑤

Before proceeding to quadratic variation we first introduce the fundamental concept of a local martingale.

Let $X = \{X_t\}_{t \geq 0}$ be an adapted process. If there exists a sequence of stopping times $\{T_n\}_{n=1,2,3,\dots}$ s.t.

1) $T_n \rightarrow \infty$ a.s.

2) $X^{(n)} = \{X_{t \wedge T_n}\}_{t \geq 0}$ is a martingale $\forall n = 1, 2, 3, \dots$

then we say X is a local martingale

If additionally:

a) $X_0 = 0$ a.s., write $X \in M_{loc}$

b) $X_0 = 0$ a.s., X continuous we write $X \in M_{c,loc}$

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Important facts:

- ~~Local~~ Martingales are Local Martingales
($T_n = n$)
- Local Martingales need not be Martingales
 - we will come up with examples of uniformly integrable local martingales which are not martingales.

The Basic Issue:

X : (τ -cont, cont) local martingale

$\forall n$:

$$E[|X_{\tau \wedge T_n}|] < \infty \quad \forall \epsilon > 0$$

$$E[X_{\tau \wedge T_n} 1_A] = E[X_{\Delta \wedge T_n} 1_A]$$

$$\forall 0 \leq \Delta < \tau, A \in \mathcal{F}_\Delta$$

⑦

So, for X to be a martingale
we need

$\{X_{\tau \wedge T_n}\}_{n=1,2,\dots}$ to be u.i.

for each $\tau \geq 0$.

- not enough that $\{X_t\}_{t \geq 0}$ be u.i.

But, one can show:

1) Local Martingales of class DL are
Martingales

2) Non-negative Local Martingales are
super-Martingales

And, we can always assume, at least for
continuous processes, that local martingales
are uniformly bounded

1) $\{X_{\tau \wedge T_n}\}$ is a martingale

2) set $\sigma_n = \inf\{\tau \geq 0 \mid |X_t| = n\}$ - stopping

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with $\tilde{T}_n = \sigma_n \wedge T_n$ we have $\tilde{T}_n \rightarrow \infty$
a.s. and

$$|X_{t \wedge \tilde{T}_n}| \leq n$$

and one can use optional sampling to
show that $X_{t \wedge \tilde{T}_n}$ is a martingale.

- we can use \tilde{T}_n as a sequence of
stopping times.

Local Martingales are Ubiquitous!

Quadratic Variation.

Let $X \in M_2$. Then X^2 is a
(\mathbb{F} -cont) sub martingale which is
non-negative (hence of class DL) so
we can write

$$X_t^2 = M_t + A_t \quad *$$

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M : r -cont martingale

A : r -cont, \nearrow , natural

Also, if $X \in M_2$ then we can take M, A to be continuous.

Definition.

$X \in M_2$. Then the quadratic variation of X is the process

$$\langle X \rangle_t = A_t \quad \text{A710}$$

where A is from $*$. Thus, $\langle X \rangle$ is the unique (up to indistinguishability) natural, increasing process so that

$$X^2 - \langle X \rangle_t$$

is a martingale.

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Q. Why "Quadratic" Variation?

Let X be any process, and fix $t \geq 0$.

Let $\Pi = \{t_0, t_1, \dots, t_n\}$ be a partition of $[0, t]$

$$0 = t_0 < t_1 < \dots < t_n = t.$$

The quadratic variation of X over Π is defined by

$$V_t^X(\Pi) = \sum_{i=1}^n |X_{t_i} - X_{t_{i-1}}|^2 \quad (u)$$

- this looks like "real" quadratic variation.

Define the "size" $\|\Pi\|$ of the partition by

$$\|\Pi\| = \max_{i \leq n} (t_i - t_{i-1})$$

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The reason we say $\langle X \rangle$ is the quadratic variation is the following:

Thm.

Let $X \in \mathcal{M}_2^c$, $\epsilon > 0$. Then

$$\lim_{\|\pi\| \rightarrow 0} V_\epsilon^{(2)}(\pi) = \langle X \rangle_\epsilon \text{ in Probability.}$$

I.e. $\forall \epsilon > 0, \gamma > 0 \exists \delta > 0$ s.t.

$$\|\pi\| < \delta \Rightarrow P(|V_\epsilon^{(2)}(\pi) - \langle X \rangle_\epsilon| > \epsilon) < \gamma.$$

Why is this true?

Key observation: $0 \leq \Delta < \epsilon \leq U < V$

$$E[(X_\epsilon - X_\Delta)(X_V - X_U)]$$

$$= E[(X_\epsilon - X_\Delta) E[(X_V - X_U) | \mathcal{F}_U]]$$

$$= 0$$

- cross-product terms all go to zero in non-overlapping intervals.

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In particular, for $0 \leq t \leq u < v$

$$\begin{aligned}
E[(X_v - X_t)^2 | \mathcal{F}_t] &= E[X_v^2 - 2X_v X_t + X_t^2 | \mathcal{F}_t] \\
&= E[X_v^2 - 2X_t E[X_v | \mathcal{F}_t] + X_t^2 | \mathcal{F}_t] \\
&= E[X_v^2 - X_t^2 | \mathcal{F}_t] \\
&= E[X_v^2 - \langle X \rangle_v + \langle X \rangle_v - (X_t^2 - \langle X \rangle_t) - \langle X \rangle_t | \mathcal{F}_t] \\
&= E[\langle X \rangle_v - \langle X \rangle_t | \mathcal{F}_t]
\end{aligned}$$

• $X^2 - \langle X \rangle$ a martingale.

~~Summary~~

Summarizing: 3 equalities

- 1) $E[(X_v - X_t)^2 | \mathcal{F}_t] = E[X_v^2 - X_t^2 | \mathcal{F}_t]$
- 2) $E[(X_v - X_t)^2 - (\langle X \rangle_v - \langle X \rangle_t) | \mathcal{F}_t] = 0$
- 3) $E[X_v^2 - X_t^2 - (\langle X \rangle_v - \langle X \rangle_t) | \mathcal{F}_t] = 0$

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Using these equalities the proof is straight forward, using a localization technique:

Step 1: Assume $|X_t| \leq K$ for $t \in \mathcal{E}$.

Then $E[(V_{\mathcal{E}}^{(2)}(\pi))^2] \leq 6K^4$

pf

$$\begin{aligned} E[(V_{\mathcal{E}}^{(2)}(\pi))^2] &= E\left[\sum_{i=1}^n (X_{t_i} - X_{t_{i-1}})^4\right] \\ &\quad + 2 E\left[\sum_{i=1}^n \sum_{j=i+1}^n (X_{t_i} - X_{t_{i-1}})^2 (X_{t_j} - X_{t_{j-1}})^2\right] \end{aligned}$$

$$\stackrel{\textcircled{1}}{\leq} 4K^2 E\left[\sum_{i=1}^n (X_{t_i} - X_{t_{i-1}})^2\right] \quad \text{ ~~$(X_{t_i} - X_{t_{i-1}})^2$~~ }$$

$$|X_{t_i} - X_{t_{i-1}}| \leq 2K.$$

$$= 4K^2 E\left[\sum_{i=1}^n X_{t_i}^2 - X_{t_{i-1}}^2\right] = 4K^2 E[X_t^2 - X_0^2]$$

$$\leq 4K^4.$$

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$$\mathbb{E} \left[\sum_{i=1}^n \sum_{j=i+1}^n (X_{t_i} - X_{t_{i-1}})^2 (X_{t_j} - X_{t_{j-1}})^2 \right]$$

$$= \mathbb{E} \left[\sum_{i=1}^n (X_{t_i} - X_{t_{i-1}})^2 \mathbb{E} \left[\sum_{j=i+1}^n (X_{t_j} - X_{t_{j-1}})^2 \mid \mathcal{F}_{t_i} \right] \right]$$

$$= \mathbb{E} \left[\sum_{i=1}^n (X_{t_i} - X_{t_{i-1}})^2 \mathbb{E} \left[\sum_{j=i+1}^n (X_{t_j}^2 - X_{t_{j-1}}^2) \mid \mathcal{F}_{t_i} \right] \right]$$

$= X_{t_j}^2 - X_{t_{j-1}}^2 \leq K^2$

$$\leq K^2 \mathbb{E} \left[\sum_{i=1}^n (X_{t_i} - X_{t_{i-1}})^2 \right]$$

$$\leq K^4 \quad (\text{previous calculation})$$

∴ (1) + 2(2) ≤ 4K^4 + 2K^4

Step 2: Assume $|X_t| \leq K$ $0 \leq t \leq T$

Then $\lim_{\|\pi\| \rightarrow 0} \mathbb{E} \left[V_t^{(4)}(\pi) \triangleq \sum_{i=1}^n (X_{t_i} - X_{t_{i-1}})^4 \right] = 0$

P.F.
 Note $V_t^{(4)}(\pi) \leq V_t^{(3)}(\pi) \cdot \max_{i=1, \dots, n} |X_{t_i} - X_{t_{i-1}}|^2$

(13)

$$\text{Set } m_{\epsilon}(X; \delta) = \sup_{\substack{0 \leq u < v \leq \epsilon \\ v - u \leq \delta}} |X_v - X_u|$$

- mbl r.v. b/c X continuous

$$\Rightarrow V_{\epsilon}^{(4)}(\pi) \leq V_{\epsilon}^{(2)}(\pi) \cdot m_{\epsilon}(X; \|\pi\|)^2$$

Now.

$$1) m_{\epsilon}(X; \|\pi\|) \leq \cancel{2K} 2K \quad |X_t| \leq K$$

$$2) \lim_{\|\pi\| \rightarrow 0} m_{\epsilon}(X; \|\pi\|) = 0 \quad \text{a.s. b/c}$$

X is continuous, hence uniformly continuous on $[0, t]$.

$$\begin{aligned} \therefore E[V_{\epsilon}^{(4)}(\pi)] &\leq E\left[\left(V_{\epsilon}^{(2)}(\pi)\right)^2\right]^{1/2} E\left[m_{\epsilon}^4(X; \|\pi\|)\right]^{1/2} \\ &\leq \sqrt{6K^4} E\left[m_{\epsilon}^4(X; \|\pi\|)\right]^{1/2} \end{aligned}$$

$$\rightarrow 0 \quad \text{as } \|\pi\| \rightarrow 0 \text{ by}$$

the dominated convergence thm.

□

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Step 3: Assume $|X|_{\Delta} \leq K, \langle X \rangle_{\Delta} \leq K$
for $\Delta \leq t$. Then

$$\lim_{\|\pi\| \rightarrow 0} E[(V_{\pi}^{(X)}(\pi) - \langle X \rangle_{\pi})^2] = 0$$

so $V_{\pi}^{(X)}(\pi) \rightarrow \langle X \rangle_{\pi}$ in L^2 , Prob.

PE

$$E[(V_{\pi}^{(X)}(\pi) - \langle X \rangle_{\pi})^2]$$

$$= E\left[\left(\sum_{k=1}^n (X_{t_k} - X_{t_{k-1}})^2 - (\langle X \rangle_{t_n} - \langle X \rangle_{t_0})\right)^2\right]$$

$$= \sum_{k=1}^n E\left[\left((X_{t_k} - X_{t_{k-1}})^2 - (\langle X \rangle_{t_k} - \langle X \rangle_{t_{k-1}})\right)^2\right]$$

- equality 2) before

$$\leq 2 \sum_{k=1}^n E[(X_{t_k} - X_{t_{k-1}})^4]$$

$$+ 2 \sum_{k=1}^n (\langle X \rangle_{t_k} - \langle X \rangle_{t_{k-1}})^2$$

$$\textcircled{17} \leq 2 E[V_t^{(4)}(\pi)] + 2 E[\langle X \rangle_t \cdot m_t(\langle X \rangle; \pi)]$$

↑ pull out max, taking sup sum with $\langle X \rangle_0 = 0$

$$\rightarrow 0$$

b/c $\langle X \rangle_t \leq K$ + previous lemma.

Step 4: Un-localiza....

$$T_n = \inf\{t \geq 0 \mid |X|_t \geq n \text{ or } \langle X \rangle_t \geq n\}$$

$\Rightarrow \{X_{t \wedge T_n}\}_{t \geq 0}$; $\{X_{t \wedge T_n}^2 - \langle X \rangle_{t \wedge T_n}\}_{t \geq 0}$ bdd
martingales

Also, $\langle X_{\cdot \wedge T_n} \rangle_t = \langle X \rangle_{t \wedge T_n}$ by uniqueness in Doob-decomposition.

By step 3:

$$\lim_{\|\pi\| \rightarrow 0} E \left[\left(\sum_{k=1}^n (X_{t_k \wedge T_n} - X_{t_{k-1} \wedge T_n})^2 - (\langle X \rangle_{t_k \wedge T_n} - \langle X \rangle_{t_{k-1} \wedge T_n}) \right)^2 \right] = 0$$

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Since $T_n \rightarrow \infty$ a.s. it follows that

$$V_t^{(n)}(\pi) \rightarrow \langle X \rangle_t \quad \text{in Prob.}$$

-make sure you can show this! ~~z~~

Covariations

Let $X, Y \in M_2$. We define the covariation, or cross-variation, of X and Y as

$$\langle X, Y \rangle_t = \frac{1}{4} [\langle X+Y \rangle_t - \langle X-Y \rangle_t] \quad \text{1710}$$

$$\bullet X+Y, X-Y \in M_2$$

Note

$$ab = \frac{1}{4} ((a+b)^2 - (a-b)^2) \quad a, b \in \mathbb{R}.$$

$$\Rightarrow \sum_{i=1}^n (X_{t_i} - X_{t_{i-1}})(Y_{t_i} - Y_{t_{i-1}}) = \frac{1}{4} \sum_{i=1}^n \left[(X_{t_i} + Y_{t_i} - (X_{t_{i-1}} + Y_{t_{i-1}}))^2 - (X_{t_i} - Y_{t_i} - (X_{t_{i-1}} - Y_{t_{i-1}}))^2 \right]$$

(19) so if $X, Y \in M^c$ then

$$\sum_{k=1}^n (X_{t_k} - X_{t_{k-1}})(Y_{t_k} - Y_{t_{k-1}}) \xrightarrow{\text{prob.}} \langle X, Y \rangle_t$$

as $\|\pi\| \rightarrow 0$ for $t \geq 0$

- why we call it covariation.

Also

$$1) X_t Y_t - \langle X, Y \rangle_t$$

$$= \frac{1}{4} \left[(X_t + Y_t)^2 - \langle X + Y \rangle_t \right.$$

$$\left. - \left((X_t - Y_t)^2 - \langle X - Y \rangle_t \right) \right]$$

so $XY - \langle X, Y \rangle$ is a martingale.

2) We say X, Y are orthogonal
if $\langle X, Y \rangle_t = 0 \quad \forall t \geq 0$ with
probability one.

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so $\langle X, Y \rangle$ orthogonal implies
 XY is a martingale.

3) $\langle X \rangle_t = \langle X, X \rangle_t$ for $X \in M_2$

4) Another way to think about
 $\langle X, Y \rangle$ for $X, Y \in M_2^c$:

$\langle X, X \rangle$: unique continuous process
A of bounded variation
with $A_0 = 0$ s.t. $\{X_t^2 - A_t\}_{t \geq 0}$
is a martingale.

pf.

If $\exists A, B$ s.t. statement holds
then $B - A$ is a continuous
martingale of finite variation

$\Rightarrow B - A \equiv 0$. (Read Thm 5.13)

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5) $\langle X, Y \rangle$ well defined for X, Y in $M^{c,loc}$

$\exists!$ adapted, continuous process $A, A_0 = 0$

s.t. $XY - A \in M^{c,loc}$,

with $A = \langle X, Y \rangle$

idea:

$X_t^n = X_{t \wedge T_n}, Y_t^n = Y_{t \wedge T_n} \in M_2^c$ so

A^n s.t. $X^n Y^n - A^n \in M_2^c$. By

uniqueness $A^n = A^m$ on $T_n \wedge T_m$

since $T_n \rightarrow \infty$ we define $A = A^n$ on

$[0, T_n]$ and A is consistently defined

$\forall t \geq 0$ and $XY - A \in M^{c,loc}$ with

$\{T_n\}$