

①

Construction of the Stochastic Integral

Goal: Assign meaning to and analyze the object

$$I_t = \int_0^t X_s dM_s \quad t \geq 0$$

X : given (adapted, ???) process

~~for~~ $M \in \mathcal{M}_0^c$

Problem: with probability one, the paths of M have unbounded variation (first) so I cannot be defined pathwise.

Workaround: exploit the Itô isometry, density of simple processes to define I as an L^2 limit.

Plan: 1) construct I for simple processes.

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2) Extend I to "general" processes when $M = W \circ B.M.$

3) Extend I to "general" processes for general $M \in M_2^c$, $M \in M^{c,loc}$.

Preliminaries:

$(\Omega, \mathcal{F}, P)$, $\mathbb{F}$ given. $\mathbb{F}$: satisfies usual conditions (important here).

$M \in M_2^c$

A) Integral for Simple Processes.

We say $X = \{X_t\}_{t \geq 0}$ is simple if X admits the form

$$X_t(\omega) = 1_0(t) \xi_0(\omega) + \sum_{i=0}^{\infty} \xi_i(\omega) 1_{(t_i, t_{i+1}]}(t)$$

with

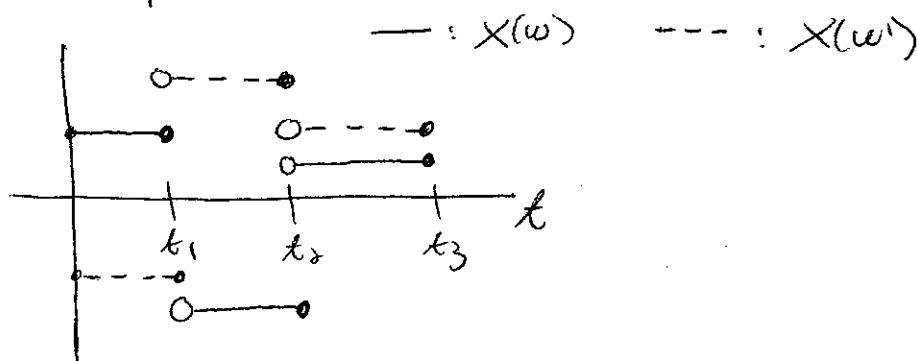
i) $t_i \nearrow \infty$

ii) ξ_i is $\mathcal{F}_{t_i}$ mbl

iii) $\sup_{i, \omega} |\xi_i(\omega)| \leq C < \infty.$

3)

Simple processes are random step functions



- note:
- 1) X is left continuous, banded
 - 2) X is progressively measurable

For X simple, $M \in \mathcal{M}_2^c$ we define the integral pathwise by

$$I_t(\omega) = I(X)_t(\omega) \triangleq \sum_{k=0}^{\infty} \overset{\substack{\text{left} \\ \text{end point}}}{\xi_k(\omega)} (M_{t \wedge t_{k+1}} - M_{t \wedge t_k})(\omega)$$

- essentially the discrete time martingale transform.

- note:
- $$0 \leq t < t_1: \quad I_t = \xi_0 M_t$$
- $$t_1 \leq t < t_2: \quad I_t = \xi_0 M_{t_1} + \xi_1 (M_t - M_{t_1})$$

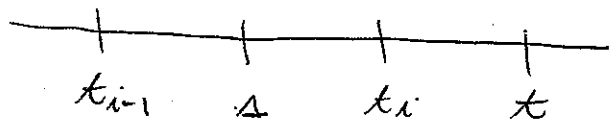
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Properties of I for X simple.

1) $I_0 = 0$ a.s. (clear)

2) I is a martingale.

pf.



(other cases similar)

$$I_t = \sum_{j=0}^{i-1} \xi_j (N_{t_{j+1}} - N_{t_j}) + \xi_i (N_t - N_{t_i})$$

$$I_\Delta = \sum_{j=0}^{i-1} \xi_j (N_{t_{j+1}} - N_{t_j}) + \xi_{i-1} (N_\Delta - N_{t_{i-1}})$$

$$\begin{aligned} \Rightarrow I_t - I_\Delta &= \xi_i (N_t - N_{t_i}) + \xi_{i-1} (N_{t_i} - N_{t_{i-1}}) \\ &\quad - \xi_{i-1} (N_\Delta - N_{t_{i-1}}) \\ &= \xi_i (N_t - N_{t_i}) + \xi_{i-1} (N_{t_i} - N_\Delta) \end{aligned}$$

$$\begin{aligned} E[\xi_i (N_t - N_{t_i}) | \mathcal{F}_\Delta] &= E[\xi_i E[N_t - N_{t_i} | \mathcal{F}_{t_i}] | \mathcal{F}_\Delta] \\ &= 0 \end{aligned}$$

$$E[\xi_{i-1} (N_{t_i} - N_\Delta) | \mathcal{F}_\Delta] = \xi_{i-1} E[N_{t_i} - N_\Delta | \mathcal{F}_\Delta] = 0$$

□

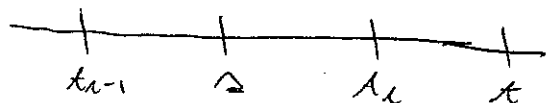
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3) I is continuous (clear)

4) $I \in \mathcal{M}_2^c$ with $E[I_x^2] = E[\int_0^x X_u^2 d\langle \Pi \rangle_0]$
and $\langle I \rangle_x = \int_0^x X_u^2 d\langle \Pi \rangle_0$

- Ito isometry.

pt.



(other cases similar)

$$E[(I_x - I_s)^2 | \mathcal{F}_s]$$

$$= E[\xi_s^2 (\Pi_x - \Pi_{x_i})^2 + 2\xi_s \xi_{x_{i-1}} (\Pi_x - \Pi_{x_i}) (\Pi_{x_i} - \Pi_s) + \xi_{x_{i-1}}^2 (\Pi_{x_i} - \Pi_s)^2 | \mathcal{F}_s]$$

$$= E[\xi_s^2 (\langle \Pi \rangle_x - \langle \Pi \rangle_{x_i}) + \xi_{x_{i-1}}^2 (\langle \Pi \rangle_{x_i} - \langle \Pi \rangle_s) | \mathcal{F}_s]$$

$$+ E[\xi_s^2 ((\Pi_x - \Pi_{x_i})^2 - (\langle \Pi \rangle_x - \langle \Pi \rangle_{x_i})) + \xi_{x_{i-1}}^2 ((\Pi_{x_i} - \Pi_s)^2 - (\langle \Pi \rangle_{x_i} - \langle \Pi \rangle_s)) | \mathcal{F}_s]$$

⑥

$$= E \left[\int_0^t X_u^2 d\langle M \rangle_u \mid \mathcal{I}_t \right].$$

$$\begin{aligned} - X_u &= \zeta_{u-1} & u \in (t_{i-1}, t_i] \\ &= \zeta_i & u \in (t_i, t]. \end{aligned}$$

∴

$$\begin{aligned} 1) \quad E[I_t^2] &= E[E[I_t^2 \mid \mathcal{I}_0]] \\ &= E[E[\int_0^t X_u^2 d\langle M \rangle_u \mid \mathcal{I}_0]] \\ &= E[\int_0^t X_u^2 d\langle M \rangle_u] \end{aligned}$$

$$\Rightarrow E[I_t^2] < \infty \quad \forall t \quad \text{since } |X| \leq C.$$

$$\Rightarrow I \in M_2^c$$

$$2) \quad \left(I_t^2 - \int_0^t X_u^2 d\langle M \rangle_u \right)_{t \geq 0} \text{ is a martingale}$$

$$\Rightarrow \langle I \rangle_t = \int_0^t X_u^2 d\langle M \rangle_u. \quad \blacksquare$$

(subtle point: X prog. mbl ~~implies~~ simple $\Rightarrow \int_0^t X_u^2 d\langle M \rangle_u$ adapted)

⑦

$$5) I \text{ is linear in that } I(\alpha X + \beta Y) \\ = \alpha I(X) + \beta I(Y)$$

-clear after enlarging partition to contain both $\{t_i\}, \{s_i\}$

Summarizing

X simple, $\Pi \in M_2^c$ implies $I(X) \in M_2^c$

with $\langle I \rangle_t = \int_0^t X_s^2 d\langle \Pi \rangle_s$. The map

$\otimes X \rightarrow I(X)$ is linear.

≡

Extending I to General Integrands.

Basic Idea: find X^n simple s.t. $X^n \rightarrow X$

in some sense. Then, show that I^n converges

to some process $\hat{I}$. Define $I(X) = \hat{I}$.

⑧

Approximation.

Lemma.

Let $X = \{X_t\}_{t \geq 0}$ be bounded (t, ω) adapted, measurable. There is a sequence of simple processes s.t.

$$\sup_{T > 0} \lim_{n \rightarrow \infty} \mathbb{E} \left[\int_0^T |X_t^n - X_t|^2 dt \right] = 0$$

pf.

suffices to find $\{X^{(n,T)}\}_n$ simple s.t.

$$\lim_{n \rightarrow \infty} \mathbb{E} \left[\int_0^T |X_t^{(n,T)} - X_t|^2 dt \right] = 0 \quad \forall T > 0$$

$\Rightarrow \forall m, \exists n_m$ s.t.

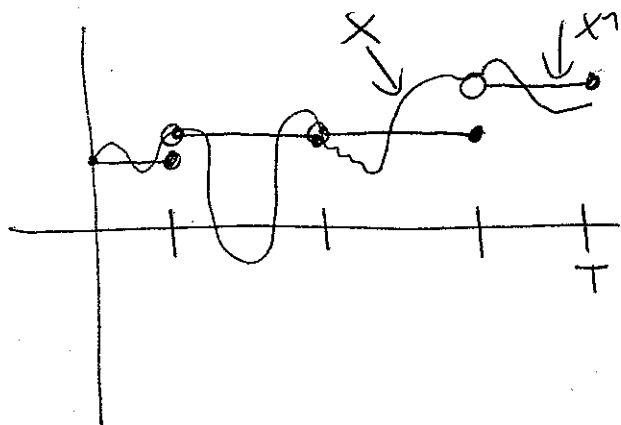
$$\mathbb{E} \left[\int_0^m |X_t^{m,n_m} - X_t|^2 dt \right] \leq 1/n$$

so $\{X^{m,n_m}\}$ works.

9)

fix $T > 0$. a) If X is continuous set

$$X^n_x(\omega) = X_0(\omega) \mathbb{1}_{\{0\}}(x) + \sum_{k=0}^{n-1} X_{\frac{kT}{n}}(\omega) \mathbb{1}_{(\frac{kT}{n}, \frac{(k+1)T}{n}]}(x).$$



X^n simple b/c X bounded, adapted, mbl.

$$E\left[\int_0^T |X^n_x - X_x|^2 dx\right] \rightarrow 0 \quad \text{b/c bounded conv thm.}$$

b) If X is progressively measurable then

i) for fixed m , approximate X by $\tilde{X}^m$ via.

$$F_x(\omega) = \int_0^{x \wedge T} X_s(\omega) ds \quad ; \quad \cancel{\tilde{X}_x(\omega) = m}$$

$$\tilde{X}_x^{(m)} = m(F_x(\omega) - F_{(x-1/m) \wedge T}(\omega))$$

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technical: need X prog mbl to have
 F adapted, hence X adapted

$\tilde{X}^n$ is continuous, bounded so $\exists \tilde{X}^n$

$$\text{s.t. } \lim_n E \left[\int_0^T |\tilde{X}_t^n - \tilde{X}_t^m|^2 dt \right] = 0.$$

But, for fixed ω the set $\leftarrow$ or lim d.n.s.

$$A_\omega = \{t \leq T \mid \lim_n \tilde{X}_t^n(\omega) \neq X_t(\omega)\}$$

is $B[0, T]$ mbl (X prog. mbl) and has
Lebesgue measure 0 (FTOC)

$$\Rightarrow \tilde{X}_t^n \mapsto X_t \quad \text{a.s. } P \times \text{Leb}[0, T].$$

$$\Rightarrow \lim_{n \rightarrow \infty} E \left[\int_0^T |\tilde{X}_t^n - X_t|^2 dt \right] = 0 \quad \text{by}$$

bounded convergence

$$\Rightarrow \exists X^n \text{ simple s.t. } \lim_n E \left[\int_0^T |X_t^n - X_t|^2 dt \right] = 0$$

- double subsequences.

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c) X bounded, mbl, adapted

- $F_x = \int_0^{t \wedge T} X_s ds$ might not be adapted.

- X has a prog mbl modification. Y

a) $G_x = \int_0^{t \wedge T} Y_s ds$ is adapted.

b) $P(G_x = F_x) = 1 \quad \forall x.$

c) F_x is adapted since $\mathcal{F}_x$ contains all P null sets.

- repeat argument from b).

Extension of Integral for $M = W$

Notation:

$$\mu_w(A) \triangleq E \left[\int_0^T 1_A(x, w) dx \right] \quad A \in \mathcal{B}([0, T] \times \Omega).$$

$$E[x]_T^2 = E \left[\int_0^T x_s^2 ds \right]$$

(L^2 norm on $\Omega \times [0, T]$ for μ_w .)

$$[x] = \sum_{n=1}^{\infty} \frac{1 \wedge [x]_n}{2^n}$$

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$$\mathcal{L}_w = \{X \text{ mbl, adapted s.t. } [X]_T < \infty \forall T \geq 0\}$$

$[]$: metric on $\mathcal{L}$.

$$\mathcal{L}_0 = \{X \in \mathcal{L} \mid X \text{ simple}\} = \{X \text{ simple}\}.$$

Lemma

$\mathcal{L}_0$ is dense in $\mathcal{L}_w$.

pf.

If $X \in \mathcal{L}_w$ is bounded, it follows by previous lemma. Else, fix n and define

$$X^n_t(\omega) = X_t(\omega) \mathbb{1}_{|X_t(\omega)| \leq n}$$

By Dominated Convergence,

$$[X^n - X]_T^2 = E\left[\int_0^T X_t^2 \mathbb{1}_{|X_t| > n} dt\right] \xrightarrow{n \rightarrow \infty} 0$$

Thus, result follows by approximating X^n and then X .

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Notes: if $x \mapsto \langle \mu \rangle_x$ is a.c. w/ prob 1
 the above result also works for

$$\mathcal{L}_\mu = \{X \text{ mbl adapted s.t. } [X]^{(n)}_T < \infty\}$$

$$[X]^{(n)}_T = E \left[\int_0^T X_x^2 d\langle \mu \rangle_x \right]$$

$$[X]^{(n)} = \sum_n \frac{1 \wedge [X]^{(n)}_n}{2^n}$$

Idea: X bounded, adapted, mbl $\Rightarrow X^n \rightarrow X$

$$\text{in that } \sup_{T>0} E \left[\int_0^T |X_x^n - X_x|^2 dx \right] = 0$$

$$\Rightarrow \exists X^{n_k} \text{ s.t. } P \otimes \text{Lab} \left(\left[\xi_x, \omega \right] \mid X^{n_k}_x(\omega) = X_x(\omega) \right)^c \\ \Rightarrow 0$$

$$\Rightarrow [X - X^n]^{(n)}_T = E \left[\int_0^T |X_x^n - X_x|^2 d\langle \mu \rangle_x dx \right]$$

$\rightarrow 0$ by bounded convergence.

15)

Integral for $X \in \mathcal{L}_W$:

• recall the metric $\|\cdot\|$ for M_2^c given by $\|X\| = \sum_{n=1}^2 \frac{1 \wedge E[X_n^2]}{2^n}$

• Itô Isometry: X, Y simple implies for

~~$I(X), I(Y) \in M_2^c$~~ $I(X), I(Y) \in M_2^c$

that

$$E[(I(X) - I(Y))_t^2]$$

$$= E[I(X - Y)_t^2] \quad (\text{linearity})$$

$$= E\left[\int_0^t (X - Y)_u^2 du\right] \quad (\text{Itô Isometry for } \langle \Pi \rangle_t = t)$$

$$= [X - Y]_t$$

$$\Rightarrow \|I(X) - I(Y)\| = [X - Y]$$

(16)

So, here is what we do:

Let $X \in \mathcal{L}_W$. Take $X^n \in \mathcal{L}_0$

s.t. $[X^n - X] \rightarrow 0$. Thus

$$\|I(X^n) - I(X^m)\| = [X^n - X^m] \rightarrow 0 \quad n, m \rightarrow \infty$$

$\Rightarrow \{I(X^n)\}$ is Cauchy in $(M_2^c, \|\cdot\|)$

$\Rightarrow \exists$ a process $I \in M_2^c$ $\leftarrow$ unique up to indistinguishability
s.t.

$$\|I(X^n) - I(X)\| \rightarrow 0$$

We define the stochastic integral to be $I(X)$ and write

$$I(X)_t = \int_0^t X_u dW_u \quad t \geq 0.$$

1) $I \in M_2^c \Rightarrow I_0 = 0$ a.s., I is a martingale.

17)

ii) fix $\Delta < t$. Note $\{I_\Delta(x^n)\}_n, \{I_\Delta(x)\}_n$ converge in mean-squares to $I_\Delta(x), I_\Delta(x)$.

$$\Rightarrow E[I_\Delta(x)_t - I_\Delta(x)_\Delta]^2$$

$$= \lim_n E[I_\Delta(x^n)_t - I_\Delta(x^n)_\Delta]^2$$

$$= \lim_n E[I_\Delta \int_\Delta^t (x^n)^2 du] \quad (\text{Isometry})$$

$$= E[I_\Delta \int_\Delta^t X_u^2 du]$$

$$\Rightarrow E[(I(x)_t - I(x)_\Delta)^2 | \mathcal{F}_\Delta] = E[\int_\Delta^t X_u^2 du | \mathcal{F}_\Delta] *$$

now

for $X \in \mathcal{L}$, one can show $\int_0^\cdot X_u^2 du$ is adapted

$$\Rightarrow \langle I \rangle_t = \int_0^t X_u^2 du$$

Also * gives

$$E[I(x)_t^2] = E[\int_0^t X_u^2 du] \Rightarrow \|I(x)\| = [x].$$

(18)

Lastly, linearity and that $I(x)$ is well defined are clear.

Integral for General $\mu \in M_2^c$.

- If $x \mapsto \langle \mu \rangle_x$ is a.c. then since $\mathcal{L}_0$ is dense in $\mathcal{L}_\mu$ w.r.t. $\|\cdot\|_\mu$ [the construction of I is the exact same

$$1) \text{ Take } x^n \in \mathcal{L}_0 \text{ s.t. } [x^n - x](\mu) \rightarrow 0$$

$$2) \|I(x^n) - I(x^m)\| = [x^n - x^m](\mu) \rightarrow 0$$

$$\therefore I(x^n) \mapsto I(x) \in M_2^c$$

$$I(x)_t \triangleq \int_0^t x_0 d\mu_0.$$

$$\langle I \rangle_t = \int_0^t x_0^2 d\langle \mu \rangle_0 \quad \text{since}$$

$$\int_0^t x_0^2 d\langle \mu \rangle_0 \text{ is adapted}$$

$$\|I(x)\| = [x](\mu).$$

(19)

- If $x \mapsto \langle \mu \rangle_x$ is not a.c. we slightly modify our argument.

1) Set $\mathcal{Y}_M^* = \{X \text{ prog. mbl s.t. } [X]_T^*(n) < \infty \forall T > 0\}$

$$[X]_T^*(n) = E\left[\int_0^T X_s^2 d\langle \mu \rangle_s\right].$$

2) Show $\mathcal{Y}_0$ is dense in $\mathcal{Y}_M^*$ w.r.t.

~~$[\cdot](M)$~~

-technical proof on pgs 135-137.

3) For $X \in \mathcal{Y}_M^*$ construct $I(X)$ exactly as before

$$\Rightarrow I(X) \in M_2^c$$

$$\langle I \rangle_t = \int_0^t X_s^2 dM_s$$

(20)

Analysis of the Integral.

Going forward we assume $X \in \mathcal{Y}_n^*$ for $n \in \mathbb{M}_2^c$. Results also hold for $X \in \mathcal{Y}_n$ if $t \rightarrow \langle n \rangle_t$ is a.c. a.s.

Prop

Let $X, Y \in \mathcal{Y}_n^*$, $S \leq T$ stopping. Then

$$1) E[(I_{t \wedge T}(X) - I_{t \wedge S}(X))(I_{t \wedge T}(Y) - I_{t \wedge S}(Y)) | \mathcal{F}_S]$$

$$= E\left[\int_{t \wedge S}^{t \wedge T} X_0 Y_0 d\langle n \rangle_0 \mid \mathcal{F}_S\right] \quad t \geq 0.$$

$$\Rightarrow S = 1 \leq t$$

$$E[(I_t(X) - I_1(X))(I_t(Y) - I_1(Y)) | \mathcal{F}_1]$$

$$= E\left[\int_1^t X_0 Y_0 d\langle n \rangle_0 \mid \mathcal{F}_1\right]$$

$$\Rightarrow \langle I(X), I(Y) \rangle_t = \int_0^t X_0 Y_0 d\langle n \rangle_0.$$

(21)

pf

Since $I(x)$ is a continuous martingale we have $E[I_{x \wedge T}(x) | \mathcal{F}_S] = I_{x \wedge S}(x)$

pf

(Prob 1.3.24 (ii)) Opt. Stoch. gives

$$E[I_{x \wedge T}(x) | \mathcal{F}_{S \wedge x}] = I_{S \wedge x}(x).$$

on $\{S \leq x\}$, prob 1.2.17 (i) at $T=S$,
 $S=x$ gives

$$E[I_{x \wedge T}(x) | \mathcal{F}_{S \wedge x}] = E[I_{x \wedge T}(x) | \mathcal{F}_S]$$

on $\{S > x\} \in \mathcal{F}_{S \wedge x}$ for $A \in \mathcal{F}_{S \wedge x}$ we have

$$E[1_A I_{x \wedge T}(x) 1_{S > x}] = E[1_A I_{x \wedge S}(x) 1_{S > x}]$$

so on $\{S > x\}$,

$$E[I_{x \wedge T}(x) | \mathcal{F}_{S \wedge x}] = I_{x \wedge S}(x) \quad \square$$

Applying this same result to the
Martingale $I_x(x)^2 - \langle I \rangle_x$ gives for x
 $\in \mathcal{L}_n^*$ gives.

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$$E[(\cancel{I}_{\tau \wedge T}(\tilde{X}) - I_{\tau \wedge S}(\tilde{X}))^2 | \mathcal{F}_S]$$

$$= E[\cancel{I}_{\tau \wedge T}(\tilde{X})^2 - I_{\tau \wedge S}(\tilde{X})^2 | \mathcal{F}_S]$$

$$= E\left[\int_{\tau \wedge S}^{\tau \wedge T} \tilde{X}_s^2 d\langle n \rangle_s \mid \mathcal{F}_S\right]$$

Now, use this for $\tilde{X} = \frac{1}{4}(X+Y)$ and $\tilde{X} = \frac{1}{4}(X-Y)$ and subtract. ~~z~~

$$1) \quad I_{\tau \wedge T}(X) = I_{\tau}(X) \quad \tilde{X}_t = X_t 1_{t \leq T}$$

stopped integral = integral for killed integrand

pf

$$E[(I_{\tau \wedge T}(X - \tilde{X}) - I_{\tau \wedge T}(X - \tilde{X}))^2 | \mathcal{F}_A]$$

$$= E\left[\int_{\tau \wedge T}^{\tau \wedge T} (X - \tilde{X})_s^2 d\langle n \rangle_s \mid \mathcal{F}_A\right] = 0$$

- note this follows from 1) by first using 1) at $S = \tau \wedge T \leq T$ to get result for $\mathcal{F}_{\tau \wedge T}$ then using problem 1.2.17 (i) to switch to $\mathcal{F}_A$ on $\{A \leq T\}$

(23)

then noting that on $\{\Delta > T\}$ both sides of the above are 0.

$\Rightarrow I_{\cdot, \wedge T}(X - \tilde{X})$ has 0 quad var

$\Rightarrow I_{\cdot, \wedge T}(X - \tilde{X}) = 0$ (indistinguishability).

Similarly, for $N_t = I_t(\tilde{X}) - I_{t \wedge T}(\tilde{X}) \in \mathcal{M}_2^c$

$$E[N_t^2] = E\left[\int_{t \wedge T}^t \tilde{X}_0^2 d\langle N \rangle_0\right] \quad (\text{opt. sampling})$$

$$= 0$$

so

$$I_{t \wedge T}(X) - I_t(\tilde{X}) = I_{t \wedge T}(X - \tilde{X})$$

$$- (I_t(\tilde{X}) - I_{t \wedge T}(\tilde{X})) = 0$$

(indistinguishable) $\square$

Thm.

$$M, N \in \mathcal{M}_2^c. \quad X \in \mathcal{L}_M^*, \quad Y \in \mathcal{L}_N^*.$$

Then

$$\langle I^M(X), I^N(Y) \rangle_t = \int_0^t X_0 Y_0 d\langle M, N \rangle_0$$

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Very important/Practical Result.

pf $X, Y \in \mathcal{L}_0$ (simple), this follows via a direct calculation.

e.g. $X_t = \xi_0 \mathbb{1}_{t \leq t_0}$; $Y_t = \theta_0 \mathbb{1}_{t \leq t_1}$

$$\Rightarrow I^M(X)_t = \xi_0 M_{t \wedge t_0}$$

$$I^N(Y)_t = \theta_0 N_{t \wedge t_1}$$

$$\begin{aligned} \Rightarrow \langle I^M(X), I^N(Y) \rangle_t &= \xi_0 \theta_0 \int_0^{t \wedge t_0 \wedge t_1} d\langle M, N \rangle_u \\ &= \int_0^t \xi_0 \mathbb{1}_{u \leq t_0} \theta_0 \mathbb{1}_{u \leq t_1} d\langle M, N \rangle_u \\ &= \int_0^t X_u Y_u d\langle M, N \rangle_u. \end{aligned}$$

to extend to $X \in \mathcal{L}_M^*$, $Y \in \mathcal{L}_N^*$ we proceed as follows:

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D (Kunita + Watanabe) with Prob 1

$$\int_0^t |X_u| |Y_u| d\check{\zeta}_u \leq \left(\int_0^t |X_u|^2 d\langle M \rangle_u \right)^{1/2} \times \left(\int_0^t |Y_u|^2 d\langle N \rangle_u \right)^{1/2}$$

$\check{\zeta}$: total variation (w-by-w) of $\langle M, N \rangle$.

pf $\check{\zeta}(w) \leq \psi(w) \triangleq \frac{1}{4} (\langle M \rangle(w) + \langle N \rangle(w))$

$$\langle M \rangle(w) \leq \psi(w)$$

$$\langle N \rangle(w) \leq \psi(w)$$

$\Rightarrow$ (think w fixed in a set of prob 1)

$$\langle M \rangle_t(w) = \int_0^t f_1(s, w) d\psi_s(w)$$

$$\langle N \rangle_t(w) = \int_0^t f_2(s, w) d\psi_s(w)$$

$$\langle M, N \rangle_t(w) = \int_0^t f_3(s, w) d\psi_s(w)$$

- R/N Derivatives of random measure on $[0, \infty)$.

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for $\alpha, B \in \mathbb{R} \quad \exists \Omega^{\alpha, B}$ w/ prob 1
s.t.

$$0 \leq \langle \alpha M + BN \rangle_t - \langle \alpha M + BN \rangle_s (w)$$

$$= \int_s^t (\alpha^2 f_1(\tau, w) + 2\alpha B f_3(\tau, w) + B^2 f_5(\tau, w)) d\langle M \rangle(\tau)$$

$\Rightarrow \forall \tau \in T_{\alpha, B}(w) \subseteq [0, \infty)$ with

$$\int_{T_{\alpha, B}(w)} d\langle M \rangle(\tau) = 0 \quad \text{w.p. 1}$$

$$\alpha^2 f_1(\tau, w) + 2\alpha B f_3(\tau, w) + B^2 f_5(\tau, w) \geq 0$$

$$\text{Set } \tilde{\Omega} = \bigcap_{\alpha, B \in \mathbb{Q}} \Omega^{\alpha, B}; \quad \tilde{T}(w) = \bigcup_{\alpha, B \in \mathbb{Q}} T_{\alpha, B}(w)$$

for $w \in \tilde{\Omega}$, $\tau \in \tilde{T}(w)$ w.p. 1

$\forall \alpha, B \in \mathbb{Q}$ that

$$\alpha^2 f_1(\tau, w) + 2\alpha B f_3(\tau, w) + B^2 f_5(\tau, w) \geq 0.$$

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Thus $*$ holds $\forall \alpha, \beta \in \mathbb{R}$.

Now, if $f_3(\tau, \omega) \geq 0$ take

$$\alpha = \alpha |X_{\tau}(\omega)|, \quad \beta = |Y_{\tau}(\omega)|$$

$$\Rightarrow 0 \leq \alpha^2 |X_{\tau}(\omega)|^2 f_1(\tau, \omega) + 2\alpha |X_{\tau}(\omega)| |Y_{\tau}(\omega)| f_3(\tau, \omega)$$

$$** + |Y_{\tau}(\omega)|^2 f_2(\tau, \omega).$$

If $f_3(\tau, \omega) \leq 0$ take $\alpha = \alpha |X_{\tau}(\omega)|$

$$\beta = -|Y_{\tau}(\omega)| \quad \text{so } ** \text{ holds as}$$

well.

Thus, integrating an $\int_0^t$ against $d\langle X, Y \rangle_u$ and plugging back in.

$$0 \leq \alpha^2 \int_0^t |X_u|^2 d\langle N \rangle_u + 2\alpha \int_0^t |X_u| |Y_u| d\check{N}_u + \int_0^t |Y_u|^2 d\langle N \rangle_u \quad (\text{as } t \geq 0)$$

Result follows by minimizing over α .

(28)

2) For $M, N \in M_2^c$, $X \in \mathcal{I}_M^*$, $Y \in \mathcal{I}_N^*$:

$$\langle I^M(X), I^N(Y) \rangle_t = \int_0^t X_u Y_u d\langle M, N \rangle_u.$$

pf

$X \in \mathcal{I}_M^*$ implies $\exists \{X^n\} \subseteq \mathcal{I}_0$ s.t.

$$\sup_T \lim_n \mathbb{E} \left[\int_0^T |X^n - X|^2 d\langle M \rangle_u \right] = 0$$

$\Rightarrow \forall T \exists$ sub-seq. (still labeled n) s.t.

$$\lim_n \int_0^T |X^n - X|^2 d\langle M \rangle_u = 0 \quad \text{o.s.}$$

$\Rightarrow t \leq T$ gives

$$|\langle I^M(X^n), N \rangle_t - \langle I^M(X), N \rangle_t|$$

$$\leq \langle I^M(X^n - X) \rangle_t \langle N \rangle_t \quad (\text{HW 1.5.7})$$

$$\leq \int_0^T |X^n - X|^2 d\langle M \rangle_u \cdot \langle N \rangle_T$$

$$\rightarrow 0 \quad \text{o.s.}$$

(29)

$$\Rightarrow \langle I^n(x), N \rangle_t = \lim_{n \rightarrow \infty} \langle I(x^n), N \rangle_t$$

$$= \lim_{n \rightarrow \infty} \int_0^t x_0^n d\langle \pi, N \rangle_0.$$

By Kunita-Watanabe.

$$\left| \int_0^t x_0^n d\langle \pi, N \rangle_0 - x_0 d\langle \pi, N \rangle_0 \right|$$

$$\leq \int_0^t |x_0^n - x_0| d\tilde{\zeta}_0$$

$\tilde{\zeta}$: total var of $\langle \pi, N \rangle$

$$\leq \sqrt{\int_0^t |x_0^n - x_0|^2 d\langle \pi \rangle_0} \sqrt{N_t}$$

$\rightarrow 0$

$$\therefore \langle I^n(x), N \rangle_t = \int_0^t x_0 d\langle \pi, N \rangle_0$$

Take $N = I^N(Y)$ $Y \in \mathcal{I}_N^*$, switch
roles of $I^n(x), N$ above.

(30)

$$\begin{aligned}\langle I^n(X), Y^n(Y) \rangle_t &= \int_0^t X_0 d\langle \Pi, I^n(Y) \rangle_0 \\ &= \int_0^t X_0 Y_0 d\langle \Pi, N \rangle_0 \quad \square.\end{aligned}$$

Integration wrt $\Pi \in M^{c,loc}$

- begin w/ a useful characterization of $I^n(X)$:

L
 $\Pi \in M^c_2, X \in \mathcal{L}^*_M$. Then $I^n(X)$ is the unique $\Phi \in M^c_2$ s.t.

$$\star \quad \langle \Phi, N \rangle_t = \int_0^t X_0 d\langle \Pi, N \rangle_0 \quad \forall N \in M^c_2.$$

pf.

~~Φ~~ $\int_0^t X_0 d\Pi_0$ satisfies $\star$.

If $\Phi \in M^c_2$ also satisfies $\star$:

(31)

$$\begin{aligned} \langle \Phi - I^n(X), N \rangle_t &= \int_0^t X_s d\langle n, N \rangle_s - \int_0^t X_s d\langle n, N \rangle_s \\ &= 0 \quad \text{a.s. } t \geq 0 \quad \forall N. \end{aligned}$$

Take $N = \Phi - I^n(X) \in M_2^c$

$$\Rightarrow \langle \Phi - I^n(X) \rangle_t = 0 \quad \text{a.s. } t \geq 0$$

$$\Rightarrow \Phi = I^n(X) \quad (\text{indistinguishability}).$$

Corollary

$$M \in M_2^c, X \in \mathcal{I}_M^*, N = I^M(X), Y \in \mathcal{I}_N^*$$

$$\Rightarrow 1) X Y \in \mathcal{I}_M^*$$

$$2) I^N(Y) = I^n(XY)$$

pf

$$E \left[\int_0^T Y_s^2 d\langle N \rangle_s \right] = E \left[\int_0^T Y_s^2 X_s^2 d\langle n \rangle_s \right]$$

$$\text{so } Y \in \mathcal{I}_N^* \Rightarrow XY \in \mathcal{I}_M^*.$$

(32)

Let $\tilde{N} \in \mathcal{M}_2^c$:

$$\langle I^M(XY), \tilde{N} \rangle = \int_0^t X_0 Y_0 d\langle M, \tilde{N} \rangle_0$$

$$= \int_0^t Y_0 d\langle N, \tilde{N} \rangle_0 = \langle I^N(X), \tilde{N} \rangle_{t, \#}$$

- this allows us to use the shorthand

$$dI^M(X)_t = X_t dM_t$$

$$dI^N(Y)_t = Y_t dN_t = Y_t X_t dM_t.$$

etc.

Localization Result.

$M, \tilde{M} \in \mathcal{M}_2^c$; $X \in \mathcal{L}_M^*$, $\tilde{X} \in \mathcal{L}_{\tilde{M}}^*$.

~~Suppose~~ $\exists$ s.t. T s.t.

$$i) X_{t \wedge T} = \tilde{X}_{t \wedge T}$$

$$ii) M_{t \wedge T} = \tilde{M}_{t \wedge T}$$

Then

$$I_{t \wedge T}^M(X) = I_{t \wedge T}^{\tilde{M}}(\tilde{X}) \quad \forall t \quad \text{n.s.}$$

(33)

pf

Let $N \in M_{\Sigma}^c$. Then

$$\langle \Pi - \tilde{\Pi}, N \rangle_{\mathcal{H}^1} = 0 \quad \text{a.s. } t \geq 0$$

$$(\text{use } \langle \Pi - \tilde{\Pi}, N \rangle = \lim_{n \rightarrow \infty} \sum \frac{(\Pi_{t_i} - \tilde{\Pi}_{t_i} - (\Pi_{t_{i+1}} - \tilde{\Pi}_{t_{i+1}}))}{(N_{t_i} - N_{t_{i+1}})})$$

to see this.)

$$\begin{aligned} \Rightarrow \langle \Pi^n(x) - \tilde{\Pi}^n(\tilde{x}), N \rangle_{\mathcal{H}^1} \\ &= \int_0^t x_0 d\langle \Pi, N \rangle_0 - \int_0^t \tilde{x}_0 d\langle \tilde{\Pi}, N \rangle_0 \\ &= 0 \end{aligned}$$

$$\text{Take } N = \Pi^n(x) - \tilde{\Pi}^n(\tilde{x})$$

Using, we extend Π^n to $N \in M_{\Sigma}^{\text{cloc}}$ Define: $\mathcal{P}_n = \{X \text{ mbl, adapted s.t.}$

$$P[\int_0^T x_0 d\langle \Pi, N \rangle_0 < \infty] = 1 \\ \forall T > 0 \}$$

(34)

$$P_M^* = \{x \in P_n \mid x \text{ prog mbl}\}.$$

Note

$$1) I_n \subseteq P_M; \quad I_n^* \subseteq P_M^* \quad \forall n \in \mathbb{N}_2.$$

2) we will define $I^n(x)$ for $x \in P_M^*$.

If $x \mapsto \langle n \rangle_x$ is a.c. we can define $I^n(x)$ for $x \in P_n$.

Main Idea

$$M \in M_{\text{cloc}} \Rightarrow \exists \{S_n\}, S_n \text{ stopping,}$$

$$S_n \nearrow \infty \text{ a.s. s.t. } M_{x \wedge S_n} \in M_2^c.$$

if $x \in P_M^*$, set

$$R_n = 1 \wedge \inf_{t \geq 0} \left\{ \int_0^t x_u d\langle n \rangle_u \geq n \right\}.$$

$$\Rightarrow R_n \text{ stopping, } R_n \nearrow \infty \text{ a.s.}$$

(35)

$$T_n = S_n \wedge R_n$$

$$M_t^n = M_{t \wedge T_n} \quad ; \quad X_t^n = X_t \mathbb{1}_{t \leq T_n}$$

$$\Rightarrow M^n \in M_2^c, \quad X^n \in \mathcal{I}_{M^n}^* \quad \text{since}$$

$$\int_0^{t \wedge T_n} (X_s^n)^2 d\langle M \rangle_s = \int_0^{t \wedge T_n} X_s^2 d\langle M \rangle_s \leq n.$$

$$\Rightarrow I^{M^n}(X^n) \in M_2^c \quad \text{well defined.}$$

$$\text{Since } M^n = M^m, \quad X^n = X^m \text{ on } [0, T_n \wedge T_m]$$

we know

$$I_x^{M^n}(X^n) = I_x^{M^m}(X^m) \quad t \leq T_n \wedge T_m$$

so, we define

$$I^n(X)_t \triangleq I^{M^n}(X^n)_t \quad t \leq T_m.$$

$$\Rightarrow I^n \text{ continuous,}$$

I^n a local martingale with

$\{T^n\}$ stopping.

(36)

In fact, we have:

$\hookrightarrow \eta \in M^{c,loc}, X \in P_N^*$ Then

$$1) I_0^\eta(X) = 0$$

$$2) I^\eta(\alpha X + \beta Y) = \alpha I^\eta(X) + \beta I^\eta(Y)$$

$$3) \langle I^\eta(X) \rangle_t = \int_0^t X_s d\langle \eta \rangle_s$$

$$4) I_{t \wedge T}^\eta(X) = I_t^\eta(\tilde{X}) \quad \tilde{X}_s = X_s 1_{s \leq T}$$

$$5) \langle I^\eta(X), I^\eta(Y) \rangle_t$$

$$= \int_0^t X_s Y_s d\langle \eta, \eta \rangle_s$$

$$\eta \in M^{c,loc}, Y \in P_N^*$$

- rule of thumb: sample path properties of $I^\eta(X)$, $\eta \in M^c, X \in P_N^*$ carry over

to $\eta \in M^{c,loc}, X \in P_N^*$. Expectation properties do not carry over.