

Thm Let X take the form $dX_t = U_t dt + Y_t dB_t$.

$$\text{Then } V_t^{(2)}(X, \Pi) \xrightarrow[\|\Pi\| \rightarrow 0]{P} \int_0^t Y_s^2 ds$$

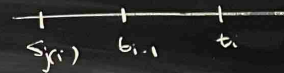
Pf STEP 0 Reduce to $U_s = 0$.

STEP 1 ($Y \in V_0$) Assume $Y_s = 1_{[0, t_0]}^{(s)} Z_0 + \sum_{i=1}^m Z_i 1_{(s_{i-1}, s_i]}(s)$

Assume $\Pi = \{0 = t_0 < \dots < t_k = t\}$ s.t. $\|\Pi\| = \max_{i=1, \dots, k} |s_i - s_{i-1}|$

Denote $I := \{i = 1, \dots, k : \{s_0, \dots, s_m\} \cap [t_{i-1}, t_i] \neq \emptyset\}$

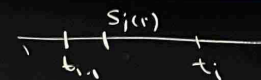
$$j(i) := \max \{j = 1, \dots, m : s_j \leq t_i\}$$



Then:

$$i \in I \Rightarrow t_{i-1} \leq s_{j(i)} \leq t_i$$

$$i \notin I \Rightarrow t_{i-1} > s_{j(i)} \wedge s_{j(i)+1} > t_i$$



$$\begin{aligned} \text{So } V_t^{(2)}(X, \Pi) &= \sum_{i=1}^k (X_{t_i} - X_{t_{i-1}})^2 \\ &= \sum_{i \notin I} Z_{j(i)+1}^2 (B_{t_i} - B_{t_{i-1}})^2 \quad \leftarrow 1^{st} \\ &\quad + \sum_{i \in I} \left(Z_{j(i)} (B_{s_{j(i)}} - B_{t_{i-1}}) + Z_{j(i)+1} (B_{t_i} - B_{s_{j(i)+1}}) \right)^2 \quad \leftarrow 2^{nd} \end{aligned}$$

Denote $C := \max_{j=0 \dots m} \|Z_j\|_\infty$.

$$2^{\text{nd}} \leq 2C^2 \sum_{i \in I} [(B_{s_{ij}} - B_{t_{i-1}})^2 + (B_{t_i} - B_{s_{ij}})^2]$$

$$E(2^{\text{nd}}) \leq 2C^2 \sum_{i \in I} (t_i - t_{i-1}) \leq 2C^2 |I| \|\Pi\| \leq 2C^2 (m+1) \|\Pi\|$$

$$1^{\text{st}} - \int_0^t Y_s^2 ds = \sum_{i \notin I} Z_{j(i)+1}^2 [(B_{t_i} - B_{t_{i-1}})^2 - (t_i - t_{i-1})]$$

$$\underbrace{\sum_{i=1}^n Z_{j(i)+1}^2 (t_i - t_{i-1})}_{3^{\text{rd}}} - \sum_{i \in I} Z_{j(i)+1}^2 (t_i - t_{i-1}) \leq C^2 (m+1) \|\Pi\|$$

$$E(3^{\text{rd}}) \leq C^4 \sum_{i \notin I} \text{Var}((B_{t_i} - B_{t_{i-1}})^2) = 3C^4 \sum_{i \notin I} |t_i - t_{i-1}|^2 \leq 3C^4 t \|\Pi\|$$

So we get $V_t^{(2)}(X, \Pi) \xrightarrow{\|\Pi\| \rightarrow 0} \int_0^t Y_s^2 ds$ for $Y \in V_0$.

STEP 2 ($Y \in V$) Need:

$$|V_t^{(2)}(X, \Pi)^{1/2} - V_t^{(2)}(\bar{X}, \Pi)^{1/2}|^2 \leq V_t^{(2)}(X - \bar{X}, \Pi)$$

Let $\{Y^{(n)}\}_{n \in \mathbb{N}} \in \mathcal{V}_0^M$ be s.t. $\|Y - Y^{(n)}\| \xrightarrow{n \rightarrow \infty} 0$.

Denote $X_t^{(n)} := \int_0^t Y_s^{(n)} dB_s$. Then

$$E\left(\left|V_t^{(2)}(X, \Pi)^{1/2} - V_t^{(2)}(X^{(n)}, \Pi)^{1/2}\right|^2\right)$$

$$\leq E V_t^{(2)}(X - X^{(n)}, \Pi)$$

$$\stackrel{\text{martingale}}{=} E\left((X_t - X_t^{(n)})^2\right)$$

$$\stackrel{H_0}{=} \int_0^t (Y_s - Y_s^{(n)})^2 ds = \|Y - Y^{(n)}\|_{L^2([0, t] \times \Omega)}^2 \xrightarrow{n \rightarrow \infty} 0$$

$$E\left(\left|\left(\int_0^t Y_s^2 ds\right)^{1/2} - \left(\int_0^t (Y_s^{(n)})^2 ds\right)^{1/2}\right|^2\right)$$

$$\leq E\left(\int_0^t (Y_s - Y_s^{(n)})^2 ds\right) =$$

$$\xrightarrow{n \rightarrow \infty} 0$$

Now apply 3ε argument under convergence in prob.

STEP 3 ($Y \in V^{loc}$) Then $T^{(M)} := \inf \{t \geq 0: \int_0^t Y_s^2 ds \geq M\}$.
 on $T^{(M)} > t$ we have $X_u = X_{T^{(M)} \wedge u}$ $u \leq t$
 and $X_{T^{(M)} \wedge t} = \int_0^t Y_s 1_{T^{(M)} > s} dB_s$ where $\{Y_s 1_{T^{(M)} > s} : s \geq 0\} \in \mathcal{V}$
 so $V_t^{(2)}(X_{T^{(M)} \wedge \cdot}, \Pi) \xrightarrow[\|\Pi\| \rightarrow 0]{p} \int_0^t Y_s^2 1_{T^{(M)} > s} ds = \int_0^t Y_s^2 ds$
 $\xrightarrow{\|\Pi\| \rightarrow 0} V_t^{(2)}(X, \Pi)$ $\square$

Def Given a generalized diffusion $dX_t = U_t dt + Y_t dB_t$,
 denote $\langle X \rangle_t := \int_0^t Y_s^2 ds$
 and call it the quadratic variation process.

Thm (Doob-Meyer decomposition for stoch. integral)

Let X take the form $dX_t = Y_t dB_t$ for $Y \in V_{loc}$.

Then $\{X_t^2 - \langle X \rangle_t : t \geq 0\}$ is a local martingale

Pf. ($Y \in V$) Let $u \leq t$. Then

$$E(X_t^2 - X_u^2 | \mathcal{F}_u) = E((X_t - X_u)^2 | \mathcal{F}_u)$$

conditional
Ito

$$E\left(\int_u^t Y_s^2 ds | \mathcal{F}_u\right) = E(\langle X \rangle_t - \langle X \rangle_u | \mathcal{F}_u)$$

So $E(X_t^2 - \langle X \rangle_t | \mathcal{F}_u) = X_u^2 - \langle X \rangle_u$ and
the above is a (proper) martingale.

($Y \in V_{loc}$) $X_{T^{(M)}_t} = \int_0^t Y_s 1_{T^{(M)}_s < s} dB_s$ where $\{Y_s 1_{T^{(M)}_s < s} : s \geq 0\} \in V$

$$\langle X \rangle_{T^{(M)}_t} = \int_0^t Y_s^2 1_{T^{(M)}_s < s} ds$$

So $\{X_{T^{(M)}_t}^2 - \langle X \rangle_{T^{(M)}_t} : t \geq 0\}$ is a martingale.

Since $T^{(M)} \xrightarrow[M \rightarrow \infty]{} \infty$, $\{X_t^2 - \langle X \rangle_t : t \geq 0\}$ is local martingale. $\square$

Stochastic integral w.r.t. semimartingales

Def Given a generalized diffusion $dX_t = U_t dt + Y_t dB_t$ and a process $\{Z_t; t \geq 0\}$ that is adapted, jointly-measurable and obeys: $\forall t \geq 0: \int_0^t \|U_s\| |Z_s| ds < \infty \wedge \int_0^t Z_s^2 Y_s^2 ds < \infty$ a.s.

we set
$$\int_0^t Z_s dX_s = \int_0^t Z_s U_s ds + \int_0^t Z_s Y_s dB_s.$$

Note: This is well-defined and independent of representation of X using U and Y processes.

Thm Let X be generalized diffusion. Then $\forall f \in C^2(\mathbb{R})$

$$\forall t \geq 0: f(X_t) = f(X_0) + \int_0^t f'(X_s) dX_s + \frac{1}{2} \int_0^t f''(X_s) d\langle X \rangle_s$$

(Itô's)

(generalized Itô formula)