

Balint 6th: Itô calculus / 2

(16)

Itô's formula

$(\Omega, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P})$ filtered probability space

$B(t)$ BM, $\mathcal{F}_t^B \subseteq \mathcal{F}_t$, \mathcal{F}_t -martingale

Itô processes: (def).

$$X(t) = X(0) + \int_0^t u(s) ds + \int_0^t v(s) dB(s)$$

(i) $u(\cdot), v(\cdot)$ progressively measurable

(ii) $\mathbb{E} \left(\int_0^T |u(s)| ds \right) < \infty$

(iii) $\mathbb{E} \left(\int_0^T |v(s)|^2 ds \right) < \infty$

Notation:

$$dX(t) = u(t) dt + v(t) dB(t)$$

(17)

Theorem: Let $g: (0, \infty) \times \mathbb{R} \rightarrow \mathbb{R}$

be $C^{1,2}$; $X(t)$ Itô process with

$$dX(t) = \mu(t)dt + \sigma(t)dB(t);$$

$$Y(t) := g(t, X(t)).$$

Then $Y(t)$ is also Itô-process with

$$dY(t) = \left(\dot{g}(t, X(t)) + g'(t, X(t))\mu(t) + \frac{1}{2} g''(t, X(t))\sigma(t)^2 \right) dt + g'(t, X(t))\sigma(t)dB(t)$$

$$= \dot{g}(t, X(t))dt + g'(t, X(t))dX(t) + \frac{1}{2} g''(t, X(t))\sigma(t)^2 dt$$

cook-book:

$$(dB(t))^2 = dt, \quad (dB(t))^3 = dt dB(t) = (dt)^2 = 0$$

$$(dX(t))^2 = \nu(t)^2 dt, \quad (dX(t))^3 = dt dX(t) = (dt)^2 = 0$$

Examples: $X(t) = B(t)$ ($\mu \equiv 0, \nu \equiv 1$)

① $g(t, x) = g(x) \quad g \in C^2$

$$g(B(t)) - g(B(0)) = \int_0^t g'(B(s)) dB(s) + \frac{1}{2} \int_0^t g''(B(s)) ds$$

$$\int_0^t f(B(s)) dB(s) = F(B(t)) - F(B(0)) - \frac{1}{2} \int_0^t f'(B(s)) ds$$

$$f \in C^1$$

② $g(t,x) = x g(t)$ $g \in C^1$

$$\int g(s) dB(s) = g(t)B(t) - \int_0^t g(s)B(s) ds$$

(could be derived directly by integration by parts)

Full proof for $g(t,x) = g(x)$ at end of this set of notes

Proof of Itô's formula (sketch)

assume for simplicity $g \in C^{2,3}$

$$t_j = t \cdot \frac{j}{n} \quad j=0,1,\dots,n$$

$$Y(t) - Y(0) = \sum_{j=1}^n Y(t_j) - Y(t_{j-1})$$

$$Y(t_j) - Y(t_{j-1}) = g(t_j, X(t_j)) - g(t_{j-1}, X(t_{j-1}))$$
 Taylor expansion

$$g(t_{j-1}, X(t_{j-1})) (t_j - t_{j-1}) +$$

$$g'(t_{j-1}, X(t_{j-1})) (M(t_j) (t_j - t_{j-1}) + N(t_{j-1}) (B(t_j) - B(t_{j-1})))$$

+ %

$$+ \frac{1}{2} g''(t_{j-1}, X(t_{j-1})) \sigma(t_{j-1})^2 (B(t_j) - B(t_{j-1}))^2$$

+ error_j

$$|\text{error}_j| \leq C \left((\Delta t_j)^2 + |\Delta B_j|^3 \right)$$

\uparrow $t_j - t_{j-1}$ \uparrow $B(t_j) - B(t_{j-1})$

$$\sum_j |\text{error}_j| \leq C \sum_j (|\Delta t_j|^2 + |\Delta B_j|^3) \rightarrow 0$$

as $n \rightarrow \infty$

...
 $Y(t) - Y(0) =$ this is I²-limit

$$\lim_{n \rightarrow \infty} \sum_j \left\{ g_j(\Delta t_j) + g'_j(u_j(\Delta t_j) + \sigma_j \Delta B_j) + \frac{1}{2} g''_j \sigma_j^2 (\Delta t_j) \right\} =$$

$$\int_0^t \left(g(s, X(s)) + g'(s, X(s)) u(s) + \frac{1}{2} g''(s, X(s)) \sigma(s)^2 \right) ds$$

$$+ \int_0^t g'(s, X(s)) \sigma(s) dB(s)$$

□

Multidimensional Itô formula:

$$B = (B_1, B_2, \dots, B_m)^T \quad \text{Brownian motion in } \mathbb{R}^m$$

$$u = (u_1, u_2, \dots, u_n)^T \quad \text{process in } \mathbb{R}^n$$

$$v = \left(v_{ji} \right)_{\substack{i=1, \dots, m \\ j=1, \dots, n}} \quad \text{process in } \mathbb{R}^{n \times m}$$

$$dX(t) = u(t) dt + v(t) dB(t)$$

$$dX_j(t) = u_j(t) dt + v_{ji}(t) dB_i(t)$$

$$g = [0, \infty) \times \mathbb{R}^n \rightarrow \mathbb{R}^p, \quad C^{1,2}$$

$$g(t, x) = \left(g_k(t, x_1, \dots, x_n) \right)_{k=1, \dots, p}$$

$$Y(t) := g(t, X(t)).$$

Thm Multidimensional Ito formula.

(22)

$$dY_k(t) =$$

$$\left(\frac{\partial g^k}{\partial t}(t, X(t)) + \sum_{j=1}^n \frac{\partial g^k}{\partial X_j}(t, X(t)) \mu_j(t) + \right.$$

$$\left. + \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 g^k}{\partial X_i \partial X_j}(t, X(t)) \sum_{r=1}^m \sigma_{ir}(t) \sigma_{jr}(t) \right) dt$$

$$+ \sum_{j=1}^n \sum_{r=1}^m \frac{\partial g^k}{\partial X_j}(t, X(t)) \sigma_{jr}(t) dB_r(t).$$

Mind the indices!

Applications of Ito: see exercises

Representation of random variables and martingales as Ito integrals:

$$(\Omega, \mathcal{F}, \mathbb{P}) \quad t \mapsto B(t) \quad \text{BM}$$

now this is the filtration considered!!! $\rightarrow (\mathcal{F}_t^B)_{t \geq 0}$ natural filtration of B .

Theorem (Ito representation Thm)

$$\left\{ X \in L^2(\Omega, \mathcal{F}_T^B, \mathbb{P}) \right\}$$

||

$$\left\{ c + \int_0^t \varphi(s) dB(s) : c = \mathbb{E}(X) \in \mathbb{R}; \varphi \in \mathcal{V}_T \right\}$$

Theorem (martingale representation thm)

Let $M(t)$ be a (\mathcal{F}_t^B) -martingale, L^2

Then $\exists! \varphi \in \mathcal{V}_T$:

$$M(t) = \int_0^t \varphi(s) dB(s)$$

$h : [0, \infty) \rightarrow \mathbb{R}$ deterministic, s.t.

$$(\forall t < \infty) : \int_0^t |h(s)|^2 ds < \infty$$

$$Y_h(t) := \exp\left(\int_0^t h(s) dB(s) - \frac{1}{2} \int_0^t h(s)^2 ds\right)$$

Lemma 1

$$Y_h(t) = 1 + \int_0^t Y_h(s) h(s) dB(s)$$

Proof apply Ito with $dX(t) = h(s)dB(s) - \frac{1}{2}h(s)^2 ds$

$$g(t, x) = e^x$$

$$\text{get } dY_h(t) = Y_h(s) h(s) dB(s) \quad \square$$

Note : $t \mapsto Y_h(t)$ is a martingale!

Lemma 2 :

linear span of $\{Y_h(t) : h \text{ locally bdd}\}$ is dense in $L^2(\Omega, \mathcal{F}_t^B, \mathbb{P})$.

Proof : tedious but no surprises

Itô repr. then follows from Lemma 1 + Lemma 2 \square

Proof of martingale representation theorem:

$M(t)$ be (\mathcal{F}_t^B) - martingale

by Itô's repres. theorem $(\forall t)$ $M(t) = \int_0^t \varphi(s) dB(s)$
mind dependence on t

Let $0 \leq t_1 \leq t_2$

$$\begin{aligned} \mathbb{E} (M(t_2) | \mathcal{F}_{t_1}^B) &= \mathbb{E} \left(\int_0^{t_2} \varphi^{t_2}(s) dB(s) \mid \mathcal{F}_{t_1}^B \right) = \\ &= \int_0^{t_1} \varphi^{t_2}(s) dB(s) \end{aligned}$$

$$M(t_1) = \int_0^{t_1} \varphi^{t_1}(s) dB(s)$$

It follows that:

$$\mathbb{E} \left(\int_0^{t_1} (\varphi^{t_2}(s) - \varphi^{t_1}(s))^2 ds \right) = 0$$

$$\Rightarrow \varphi^{t_2}(s) = \varphi^{t_1}(s) \quad (\Delta, \omega) - \text{a.e.} \quad \square$$

Some details Proof of Ito's formula: assume (26)

$$f(B(t)) - f(B(0)) =$$

$$\int_0^t f'(B(s)) dB(s) + \frac{1}{2} \int_0^t f''(B(s)) ds$$

$$f(B(t)) - f(B(0)) =$$

$$\sum_{j=0}^{n-1} \left(f(B(t_{j+1})) - f(B(t_j)) \right) =$$

$$f(B(t_{j+1})) - f(B(t_j)) =$$

$$f'(B(t_j)) (B(t_{j+1}) - B(t_j)) +$$

$$\left\{ \frac{1}{2} f''(B(t_j)) \left((B(t_{j+1}) - B(t_j))^2 - (t_{j+1} - t_j) \right) + \right.$$

$$\left. \frac{1}{2} f''(B(t_j)) (t_{j+1} - t_j) + \right.$$

$$\left. (\text{Taylor error})_j \right\}$$

~~... = ...~~

Taylor error_j =

$$\frac{1}{6} \cdot f'''(B(t_j^*)) (B(t_{j+1}) - B(t_j))^3$$

$$E \left(\left(\sum_{j=0}^{I-1} \text{Taylor error}_j \right)^2 \right) \xrightarrow{\text{as } n \rightarrow \infty} 0$$

$$E \left(\left(\sum_{j=0}^{I-1} f'''(B(t_j)) (B(t_{j+1}) - B(t_j))^2 - (t_{j+1} - t_j) \right)^2 \right)$$

$$= \sum_{j=0}^{I-1} E \left(\left(f'''(B(t_j)) \right)^2 \right) (\Delta t_j)^2 (n_4 - n_2)$$

$$\rightarrow 0$$

$$E \left(\left(\sum_{j=0}^{I-1} f'''(B(t_j^*)) (B(t_{j+1}) - B(t_j))^3 \right)^2 \right)$$

$$\ll \int \left(\sum_{j=0}^{\infty} \frac{|B_{t_j^*}|^2}{\Delta t_j} \right)$$

$$\int \left(\sum_{j=0}^{\infty} \frac{|B_{t_j}|^6}{\underbrace{\Delta t_j}_{(\Delta t_j)^2}} \right)$$

$$f(t, B(t)) - f(0, B(0)) =$$

$$\int_0^t \left(\dot{f}(s, B(s)) + \frac{1}{2} f''(s, B(s)) \right) ds$$

$$+ \int_0^t f'(B(s)) dB(s)$$