

Stochastic process

III Itô Formula and the Martingale Representation Theorem

Seongho Joo

SNU MILAB

The 1-dimensional Itô formula

The 1-dimensional Itô formula

Definition 1 (1-dimensional Itô processes)

Let B_t be 1-dimensional Brownian motion $(\Omega, \mathcal{F}, \mathbb{P})$. A Itô process (or stochastic integral) is a stochastic process X_t on $(\Omega, \mathcal{F}, \mathbb{P})$ of the form

$$X_t = X_0 + \int_0^t u(s, w) ds + \int_0^t v(s, w) dB_s \quad (1.1)$$

where $v \in \mathcal{W}_{\mathcal{H}}$, so that

$$\mathbb{P} \left(\int_0^t v(s, w)^2 ds < \infty \text{ for all } t \geq 0 \right) = 1 \quad (1.2)$$

We also assume that u is \mathcal{H}_t -adapted and

$$\mathbb{P} \left(\int_0^t |u(s, w)| ds < \infty \text{ for all } t \geq 0 \right) = 1 \quad (1.3)$$

If X_t is an Itô process of the form eq. (1.1), the eq. (1.1) sometimes written in the shorter differential form

$$dX_t = u dt + v dB_t \quad (1.4)$$

The 1-dimensional Itô formula

We are now ready to state the first main result in this chapter.

Theorem 1 (The 1-dimensional Itô formula)

Let X_t be an Itô process given by

$$dX_t = u dt + v dB_t$$

Let $g(t, x) \in C^2([0, \infty) \times \mathbb{R})$, then $Y_t = g(t, X_t)$ is again an Itô process, and

$$dY_t = \frac{\partial g}{\partial t}(t, X_t) dt + \frac{\partial g}{\partial x}(t, X_t) dX_t + \frac{1}{2} \frac{\partial^2}{\partial x^2}(t, X_t) \cdot (dX_t)^2 \quad (1.5)$$

where $(dX_t)^2$ is computed according to the rules

$$dt \cdot dt = dt \cdot dB_t = dB_t \cdot dt = 0, \quad dB_t \cdot dB_t = dt \quad (1.6)$$

Example We know that

$$\int_0^t B_s dB_s = \frac{1}{2} B_t^2 - \frac{1}{2} t$$

Apply the Itô formula to find the value of the integral with $X_t = B_t$ and $g(t, x) = \frac{1}{2} x^2$

Cont.

Then, $Y_t = g(t, B_t) = \frac{1}{2}B_t^2$ and by the Itô formula

$$dY_t = \frac{\partial g}{\partial t} dt + \frac{\partial g}{\partial x} dB_t + \frac{1}{2} \frac{\partial^2 g}{\partial x^2} (dB_t)^2 = B_t dB_t + \frac{1}{2} (dB_t)^2$$

$\implies dY_t = B_t dB_t + \frac{1}{2} dt$. In other words,

$$\frac{1}{2}B_t^2 = \int_0^t B_s dB_s + \frac{1}{2}t$$

Itô Integral의 정의를 사용하지 않고도 값을 구했다.

Example Find the value of $\int_0^t s dB_s$

뭔가 적분했을 때 tB_t term이 나올 것 같으니 Itô formula를 함수를 $g(t, x) = tx$ 로 설정하는 것이 좋을 듯 하다. By the Itô formula,

$$dY_t = B_t dt + t dB_t + 0$$

$$\implies tB_t = \int_0^t B_s ds + \int_0^t s dB_s, \quad \int_0^t s dB_s = tB_t - \int_0^t B_s ds$$

The 1-dimensional Itô formula

Theorem 2 (Integral by parts)

Suppose $f(s, w) = f(s)$ only depends on s and that f is continuous and of bounded variation in $[0, t]$. Then

$$\int_0^t f(s) \, dB_s = f(t)B_t - \int_0^t B_s \, df_s \quad (1.7)$$

Remark. f should not depend on w .

Multi-dimensional Itô Formula

Multi-dimensional Itô Formula

- Now we consider m -dimensional Brownian motion

$B(t, w) := (B_1(t, w), \dots, B_m(t, w))$. If each of the processes $u_i(t, w)$ and $v_{ij}(t, w)$ satisfies the conditions given in [Definition 1](#) for $(1 \leq i \leq n, 1 \leq j \leq m)$, then we can form the following n Itô processes

$$dX(t) = u dt + v dB(t), \quad (2.1)$$

where

$$X(t) = \begin{pmatrix} X_1(t) \\ \vdots \\ X_n(t) \end{pmatrix}, \quad u = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}, \quad v = \begin{pmatrix} v_{11} & \dots & v_{1m} \\ \vdots & & \vdots \\ v_{n1} & \dots & v_{nm} \end{pmatrix}, \quad dB(t) = \begin{pmatrix} dB_1(t) \\ \vdots \\ dB_m(t) \end{pmatrix} \quad (2.2)$$

Such a process $X(t)$ is called an [n-dimensional Itô process](#).

Question: What is the result of Itô formula for $g(t, X_t)$ with smooth function g ?

Multi-dimensional Itô Formula

Theorem 3

Let $dX_t = u dt + v dB_t$ be an n -dimensional Itô process as above. Let $g(t, x) = (g_1(t, x), \dots, g_p(t, x))$ be a C^2 map from $[0, \infty) \times \mathbb{R}^n \rightarrow \mathbb{R}^p$. Then the process

$$Y(t, w) = g(t, X(t))$$

is again an Itô process, whose Y_k ($k = 1, \dots, p$) is given by

$$dY_k = \frac{\partial g_k}{\partial t}(t, X) dt + \sum_i \frac{\partial g_k}{\partial x_i}(t, X) dX_i + \frac{1}{2} \sum_{i,j} \frac{\partial^2 g_k}{\partial x_i \partial x_j}(t, X) dX_i dX_j$$

where $dB_i dB_j = \delta_{ij} dt$, $dB_i dt = dt dB_i = 0$

Example. Let $B = (B_1, \dots, B_n)$ be Brownian motion in \mathbb{R}^n , $n \geq 2$, and consider¹

$$R(t, w) = |B(t, w)| = (B_1^2(t, w) + \dots + B_n^2(t, w))^{\frac{1}{2}}$$

Then, by the Itô formula

$$dR = \sum_{i=1}^n \frac{B_i dB_i}{R} + \frac{n-1}{2R} dt$$

¹Although $g(t, x) = |x|$ is not C^2 at the origin, B_t never hits the origin, a.s. when $n \geq 2$.

Multi-dimensional Itô Formula

Proof. We deal with the case $p = 1$. Fix $t > 0$ and consider an increasing sequence $0 = t_0^{(n)} < \dots < t_{p_n}^{(n)} = t$ of subdivisions of $[0, t]$ whose mesh tends to 0. Then for every n ,

$$g(X_t) = g(X_0) + \sum_{i=0}^{p_n-1} (g(X_{t_{i+1}^{(n)}}) - g(X_{t_i^{(n)}}))$$

Now, for every $i \in \{0, 1, \dots, p_n - 1\}$, we apply the Taylor-Lagrange formula to the function $[0, 1] \ni \theta \mapsto g(X_{t_i^{(n)}} + \theta(X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}}))$, between $\theta = 0$ and $\theta = 1$, and we get

$$g(X_{t_{i+1}^{(n)}}) - g(X_{t_i^{(n)}}) = g'(X_{t_i^{(n)}})(X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}}) + \frac{1}{2}g_{n,i}(X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}})^2,$$

where $g_{n,i} = g''(X_{t_i^{(n)}} + c(X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}}))$ for some $c \in [0, 1]$. We also have

$$\lim_{n \rightarrow \infty} \sum_{i=0}^{p_n-1} g'(X_{t_i^{(n)}})(X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}}) = \int_0^t g'(X_s) dX_s$$

in probability. To complete the proof $p = 1$ case, it is enough to show that

$$\lim_{n \rightarrow \infty} \sum_{i=0}^{p_n-1} g_{n,i}(X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}})^2 = \int_0^t g''(X_s) \cdot (dX_t)^2$$

in probability.

Cont.

We observe that

$$\sup_{0 \leq i \leq p_n - 1} |g_{n,i} - g''(X_{t_i^{(n)}})| \leq \sup_{0 \leq i \leq p_n - 1} \left(\sup_{x \in [X_{t_i^{(n)}} \wedge X_{t_{i+1}^{(n)}}, X_{t_i^{(n)}} \vee X_{t_{i+1}^{(n)}}]} |g''(x) - g''(X_{t_i^{(n)}})| \right)$$

The right-hand side tends to 0 a.s. $n \rightarrow \infty$ by the uniform continuity of g'' over a compact interval. Since $\sum_{i=0}^{p_n-1} (X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}})^2$ converges in probability to 0, it follows that

$$\lim_{n \rightarrow \infty} \left| \sum_{i=0}^{p_n-1} g_{n,i} (X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}})^2 - \sum_{i=0}^{p_n-1} g''(X_{t_i^{(n)}}) (X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}})^2 \right| = 0$$

in probability. Therefore, it remains to show that

$$\lim_{n \rightarrow \infty} \sum_{i=0}^{p_n-1} g''(X_{t_i^{(n)}}) (X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}})^2 = \int_0^t g''(X_s) (dX_s)^2$$

Cont (2).

To this end, we note that

$$\sum_{i=0}^{p_n-1} g''(X_{t_i^{(n)}})(X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}})^2 = \int_{[0,t]} g''(X_s) \mu_n(ds)$$

where μ_n is the random measure on $[0, t]$ defined by

$$\mu_n(dr) := \sum_{i=0}^{p_n-1} (X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}})^2 \delta_{t_i^{(n)}}(dr)$$

This random measure converges a.s. to the measure $\mathbf{1}_{[0,t]}(dX_r)^2 = \mathbf{1}_{[0,t]} dr$ ². This completes the proof of the case $p = 1$.

²이건 지금은 그냥 받아들이자..

The Martingale Representation Theorem

The Martingale Representation Theorem

- In chapter 3, we proved that if $v \in \nu^n$ then the Itô integral

$$X_t = x_0 + \int_0^t v(s, w) dB_s \quad t \geq 0$$

is always a $\mathcal{F}_t^{(n)}$ -martingale ([Corollary 3.2.6](#)). In this section we will prove that the converse is also true: Any $\mathcal{F}_t^{(n)}$ -martingale can be represented as an Itô integral.

The Martingale Representation Theorem

Lemma 1 (Dense subset in L^2 (I))

Fix $T > 0$. The set of random variables

$$\{\phi(B_{t_1}, \dots, B_{t_n}) \mid t_i \in [0, T], \phi \in C_0^\infty(\mathbb{R}^n), n = 1, 2, \dots\} \quad (3.1)$$

is dense in $L^2(\mathcal{F}_T, \mathbb{P})$

Lemma 2 (Dense subset in L^2 (II))

The linear span of random variables of the type

$$\exp\left(\int_0^T h(t) \, dB_t(w) - \frac{1}{2} \int_0^T h^2(t) \, dt\right); \quad h \in L^2[0, T] \quad (3.2)$$

is dense in $L^2(\mathcal{F}_T, \mathbb{P})$.

The Martingale Representation Theorem

- Suppose $B(t) = (B_1(t), \dots, B_n(t))$ is n -dimensional. If $v(s, w) \in \nu^n(0, T)$ then the random variable

$$V(w) := \int_0^T v(t, w) dB(t) \quad (3.3)$$

is $\mathcal{F}_T^{(n)}$ -measurable and by the Itô isometry

$$\mathbb{E} V^2 = \int_0^T \mathbb{E} v^2(t, \cdot) dt < \infty, \implies V \in L^2(\mathcal{F}_T^{(n)}, \mathbb{P})$$

The next lemma shows that any $F \in L^2(\mathcal{F}_T^{(n)}, \mathbb{P})$ can be represented this way:

Lemma 3 (The Itô representation theorem)

Let $F \in L^2(\mathcal{F}_T, \mathbb{P})$. Then there exists a unique stochastic process $f(t, w) \in \nu(0, T)$ such that

$$F(w) = \mathbb{E} F + \int_0^T f(t, w) dB(t). \quad (3.4)$$

The Martingale Representation Theorem

Theorem 4 (The martingale representation theorem)

Let $B(t) = (B_1(t), \dots, B_n(t))$ be n -dimensional. Suppose M_t is an $\mathcal{F}_t^{(n)}$ -martingale and that $M_t \in L^2(P)$ for all $t \geq 0$. Then there exists a unique stochastic process $g(s, w)$ such that $g \in \nu^{(n)}(0, t)$ for all $t \geq 0$ and

$$M_t(w) = \mathbb{E} M_0 + \int_0^t g(s, w) dB(s) \quad \text{a.s. for all } t \geq 0. \quad (3.5)$$

The Martingale Representation Theorem

Theorem 4 (The martingale representation theorem)

Let $B(t) = (B_1(t), \dots, B_n(t))$ be n -dimensional. Suppose M_t is an $\mathcal{F}_t^{(n)}$ -martingale and that $M_t \in L^2(P)$ for all $t \geq 0$. Then there exists a unique stochastic process $g(s, w)$ such that $g \in \nu^{(n)}(0, t)$ for all $t \geq 0$ and

$$M_t(w) = \mathbb{E} M_0 + \int_0^t g(s, w) dB(s) \quad \text{a.s. for all } t \geq 0. \quad (3.5)$$

Proof. ($n = 1$). By Lemma 3 with $T = t, F = M_t$, we have that for all t there exists a unique $f^{(t)}(s, w) \in L^2(\mathcal{F}_t, \mathbb{P})$ such that

$$M_t(w) = \mathbb{E} M_t + \int_0^t f^{(t)}(s, w) dB_s(w) = \mathbb{E} M_0 + \int_0^t f^{(t)}(s, w) dB_s(w)$$

Now fix $0 \leq t_1 < t_2$. Then,

$$M_{t_1} = \mathbb{E} M_{t_2} | \mathcal{F}_{t_1} = \mathbb{E} M_0 + \mathbb{E} \int_0^{t_2} f^{(t_2)}(s, w) dB_s(w) | \mathcal{F}_{t_1} \quad (3.6)$$

$$= \mathbb{E} M_0 + \int_0^{t_1} f^{(t_2)}(s, w) dB_s(w) \quad (3.7)$$

Cont.

But we also have

$$M_{t_1} = \mathbb{E} M_0 + \int_0^{t_1} f^{(t_1)}(s, w) \, dB_s(w). \quad (3.8)$$

By comparing eq. (3.7) and eq. (3.8), we get

$$0 = \mathbb{E} \left(\int_0^{t_1} (f^{(t_2)} - f^{(t_1)}) \, dB \right)^2 = \int_0^{t_1} \mathbb{E} (f^{(t_2)} - f^{(t_1)})^2 \, ds \quad (3.9)$$

$$\implies f^{(t_1)}(s, w) = f^{(t_2)}(s, w) \text{ for a.a. } (s, w) \in [0, t_1] \times \Omega.$$

So we can define $f(s, w)$ for a.a. $s \in [0, t] \times \Omega$ by setting

$$f(s, w) = f^{(t)}(s, w) \quad \text{if } s \in [0, t] \quad (3.10)$$

then we get

$$M_t = \mathbb{E} M_0 + \int_0^t f^{(t)}(s, w) \, dB_s(w) = \mathbb{E} M_0 + \int_0^t f(s, w) \, dB_s(w) \quad \text{for all } t \geq 0 \quad (3.11)$$

The Martingale Representation Theorem

Problem 4.3. Let X_t and Y_t be Itô process in \mathbb{R} . Prove that

$$dX_t Y_t = X_t dY_t + Y_t dX_t + dX_t \cdot dY_t$$

Deduce the following general [integration by parts formula](#)

$$\int_0^t X_s dY_s = X_t Y_t - X_0 Y_0 - \int_0^t Y_s dX_s - \int_0^t dX_s \cdot dY_s \quad (3.12)$$

The Martingale Representation Theorem

Problem 4.4. Suppose $\theta(t, w) = (\theta_1(t, w), \dots, \theta_n(t, w)) \in \mathbb{R}^n$ with $\theta_k(t, w) \in \nu[0, T]$ for $k = 1, \dots, n$, where $T \leq \infty$. Define

$$Z_t = \exp \left(\int_0^t \theta(s, w) dB(s) - \frac{1}{2} \int_0^t \theta^2(s, w) ds \right); \quad 0 \leq t \leq T \quad (3.13)$$

where $B(s) \in \mathbb{R}^n$ and $\theta^2 = \theta \cdot \theta$ (dot product).

a) Use Itô formula to prove that

$$dZ_t = Z_t \theta(t, w) dB(t).$$

b) Deduce that Z_t is a martingale for $t \leq T$, provided that

$$Z_t \theta_k(t, w) \in \nu[0, T] \quad \text{for } 1 \leq k \leq n$$

Remark. A sufficient condition that Z_t be a martingale is the [Kazamaki condition](#)

$$\mathbb{E} \exp \left(\frac{1}{2} \int_0^t \theta(s, w) dB(s) \right) < \infty \quad \text{for all } t \leq T \quad (3.14)$$

This is implied by the following [Novikov condition](#)

$$\mathbb{E} \exp \left(\frac{1}{2} \int_0^T \theta^2(s, w) ds \right) < \infty \quad (3.15)$$

[Novikov condition](#) \implies [Kazamaki condition](#)