

Several Exercises on Itô Integral Section

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Exercise 4.2.13.

Study the equality

$$\int_a^b B_t dB_t = \frac{1}{2}(B_b^2 - B_a^2 - (b - a)).$$

The exercise above is based from ([2], pg. 36).

The results below are mainly based on ([1], **Example 4.2**). For $n \in \mathbb{N}$, consider a partition \mathcal{P}_n defined by $a = t_{n,0} < t_{n,1} < t_{n,2} < \dots < t_{n,n} = b$, and let

$$X_{n,t} = \sum_{i=0}^{n-1} B_{t_{n,i}} \mathbf{1}_{(t_{n,i}, t_{n,i+1}]}(t)$$

with $(X_{n,t})_{t \in [a,b]}$ is an adapted elementary process for all n and the equation above is just the same equation as Equation (4.2.2) ([2], pg. 34). Note that $B_{t_{n,i}}$ is $\mathcal{F}_{t_{n,i}}$ -measurable because Brownian process is a stochastic process. By the continuity of B_t , one has $\lim_{n \rightarrow \infty} X_{n,t} = B_t$ a.s. since one has $\sup_i (t_{n,i+1} - t_{n,i}) \rightarrow 0$. Thus, according to **Definition 4.2.6** ([2], pg. 34) the Itô Integral of $X_{n,t}$ from a to b is given by

$$\int_a^b X_{n,t} dB_t = \sum_{i=0}^{n-1} B_{t_{n,i}} (B_{t_{n,i+1}} - B_{t_{n,i}}).$$

Observe that one has the following equality

$$B_{t_{n,i}} (B_{t_{n,i+1}} - B_{t_{n,i}}) = \frac{1}{2} (B_{t_{n,i+1}}^2 - B_{t_{n,i}}^2 - (B_{t_{n,i+1}} - B_{t_{n,i}})^2).$$

Hence, the Itô Integral becomes in the following form

$$\int_a^b X_{n,t} dB_t = \frac{1}{2} \sum_{i=0}^{n-1} (B_{t_{n,i+1}}^2 - B_{t_{n,i}}^2) - \frac{1}{2} \sum_{i=0}^{n-1} (B_{t_{n,i+1}} - B_{t_{n,i}})^2.$$

The first summation on the right-hand side can be easily computed as follows

$$\sum_{i=0}^{n-1} (B_{t_{n,i+1}}^2 - B_{t_{n,i}}^2) = B_{t_{n,n}}^2 - B_{t_{n,n-1}}^2 + B_{t_{n,n-1}}^2 - \dots + B_{t_{n,1}}^2 - B_{t_{n,0}}^2 = B_b^2 - B_a^2.$$

The summation computed above is telescopic. To compute the second summation on the right-hand side, one cannot use the same trick used in computing the first summation. Let us recall **Theorem 2.4.8.** ([2], pg. 19).

Theorem: Properties of 1-dimensional Brownian motion

Let $(\Omega, \mathcal{F}, \mathbb{P}, (\mathcal{F}_t)_{t \in \mathbb{R}^+}, (B_t)_{t \in \mathbb{R}^+})$ be a 1-dimensional Brownian motion.

1. Almost every path has an infinite variation on any finite interval, namely for any $a, b \in \mathbb{R}^+$,

$$\mathbb{P} \left(\left\{ \omega \in \Omega \mid \text{var}_{[a,b]}(t \mapsto B_t(\omega)) = \infty \right\} \right) = 1.$$

2. The quadratic variation of the Brownian motion converges in the L^2 -sense, namely

$$\lim_{|\mathcal{P}_\ell| \rightarrow 0} \mathbb{E} \left(\left[\sum_{j=0}^{n_\ell-1} (B_{t_{\ell,j+1}} - B_{t_{\ell,j}})^2 - (b-a) \right]^2 \right) = 0.$$

where $|\mathcal{P}_\ell| := \max_{j \in \{0,1,2,\dots,n_\ell-1\}} |t_{\ell,j+1} - t_{\ell,j}|$.

3. Almost every path is nowhere differentiable, namely

$$\mathbb{P} \left(\left\{ \omega \in \Omega \mid t \rightarrow B_t(\omega) \text{ is nowhere differentiable} \right\} \right) = 1.$$

Consider the second point of the theorem above. If one takes the limit $|\mathcal{P}_\ell| \rightarrow 0$, then it corresponds to taking the limit $n_\ell \rightarrow \infty$. Then, by using the second point of the theorem above (The quadratic variation of the Brownian motion converges in the L^2 -sense), one has

$$\int_a^b B_t dB_t = \lim_{n \rightarrow \infty} \int_a^b X_{n,t} dB_t = \frac{1}{2}(B_b^2 - B_a^2 - (b-a)).$$

Theorem 4.2.14. ([2], pg. 36)

Let $(\Omega, \mathcal{F}, \mathbb{P}, (\mathcal{F}_t)_{t \geq 0}, (B_t)_{t \geq 0})$ be the standard 1-dim. Brownian motion, and let $(X_t)_{t \in [0,T]}, (Y_t)_{t \in [0,T]}$ be adapted stochastic processes belonging to $M^2([0,T])$. Then

$$\mathbb{E} \left(\left(\int_0^T X_t dB_t \right) \left(\int_0^T Y_t dB_t \right) \right) = \int_0^T \mathbb{E}(X_t Y_t) dt.$$

Let $I_1 := \int_0^T X_t dB_t$ and $I_2 := \int_0^T Y_t dB_t$. Observe that $I_1 I_2$ can be expressed as follows

$$I_1 I_2 = \frac{(I_1 + I_2)^2}{2} - \frac{I_1^2}{2} - \frac{I_2^2}{2} = \frac{1}{2} \left(\int_0^T X_t dB_t + \int_0^T Y_t dB_t \right)^2 - \frac{1}{2} \left(\int_0^T X_t dB_t \right)^2 - \frac{1}{2} \left(\int_0^T Y_t dB_t \right)^2.$$

Then, by using the isometry property of Itô Integral written in **Proposition 4.2.7** ([2], pg. 35), namely

$$\mathbb{E} \left(\left(\int_a^b X_t dB_t \right)^2 \middle| \mathcal{F}_a \right) = \mathbb{E} \left(\int_a^b X_t^2 dB_t \middle| \mathcal{F}_a \right).$$

By using the equation written above for $a = 0$, $b = T$ and by the linearity of the expectation, one has

$$\begin{aligned}
& \mathbb{E}\left(\left(\int_0^T X_t dB_t\right)\left(\int_0^T Y_t dB_t\right)\right) \\
&= \frac{1}{2}\mathbb{E}\left(\left(\int_0^T X_t dB_t + \int_0^T Y_t dB_t\right)^2\right) - \frac{1}{2}\mathbb{E}\left(\left(\int_0^T X_t dB_t\right)^2\right) - \frac{1}{2}\mathbb{E}\left(\left(\int_0^T Y_t dB_t\right)^2\right) \\
&= \frac{1}{2}\mathbb{E}\left(\left(\int_0^T (X_t + Y_t) dB_t\right)^2\right) - \frac{1}{2}\mathbb{E}\left(\left(\int_0^T X_t dB_t\right)^2\right) - \frac{1}{2}\mathbb{E}\left(\left(\int_0^T Y_t dB_t\right)^2\right) \\
&= \frac{1}{2}\int_0^T \mathbb{E}\left((X_t + Y_t)^2\right) dt - \frac{1}{2}\int_0^T \mathbb{E}(X_t^2) dt - \frac{1}{2}\int_0^T \mathbb{E}(Y_t^2) dt \\
&= \frac{1}{2}\int_0^T \mathbb{E}(X_t^2 + Y_t^2 + 2X_t Y_t) dt - \frac{1}{2}\int_0^T \mathbb{E}(X_t^2) dt - \frac{1}{2}\int_0^T \mathbb{E}(Y_t^2) dt \\
&= \frac{1}{2}\int_0^T (\mathbb{E}(X_t^2) + \mathbb{E}(Y_t^2) + 2\mathbb{E}(X_t Y_t)) dt - \frac{1}{2}\int_0^T \mathbb{E}(X_t^2) dt - \frac{1}{2}\int_0^T \mathbb{E}(Y_t^2) dt \\
&= \int_0^T \mathbb{E}(X_t Y_t) dt.
\end{aligned}$$

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References

- [1] Fima C Klebaner. *Introduction to stochastic calculus with applications*. World Scientific Publishing Company, 2012.
- [2] Serge Richard. *Special Mathematics Lecture: Introduction to Stochastic Calculus*. 2023.