

Stochastic Calculus For Finance - volume 2

Section 4.2 Itô's Integral for Simple Integrands

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Section 4.2 Itô's Integral for Simple Integrands

投資組合標的物價格

$$\int_0^T \Delta(t) dW(t)$$

- 1. Brownian motion
- 2. Together with a filtration $\mathcal{F}(t)$

- 1. Adapted to $\mathcal{F}(t) \rightarrow \Delta(t)$ is $\mathcal{F}(t)$ -measurable
- 2. $\Delta(t)$ is independent of future Brownian increments
- 3. At time t the randomness of $\Delta(t)$ has been resolved

$$\int_0^T \Delta(t) dg(t) = \int_0^T \Delta(t) g'(t) dt$$

Problem :

Brownian motion paths cannot be differentiated with respect to time.

4.2.1 Construction of the Integral

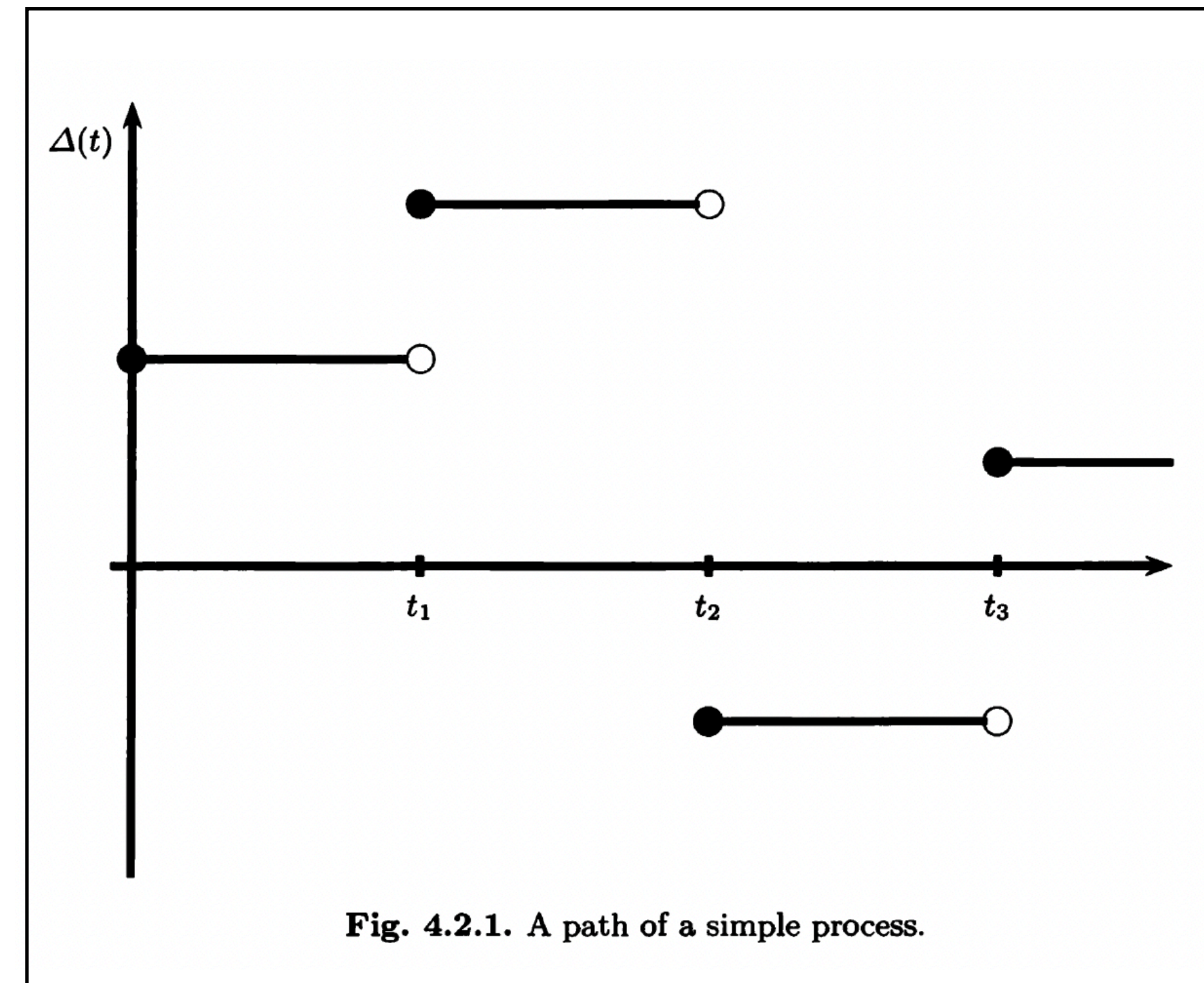
Define the Itô integral for simple integrands $\Delta(t)$

Let $\Pi = \{t_0, t_1, \dots, t_n\}$ be a partition of $[0, T]$;

i.e., $0 = t_0 \leq t_1 \leq \dots \leq t_n = T$.

Assume that $\Delta(t)$ is constant in t on each subinterval $[t_j, t_{j+1})$.

Such a process $\Delta(t)$ is a simple process.



Explain from the trading behavior of financial markets

$W(t)$: the price per share of an asset at time t .

$\Delta(t)$: as the position taken in the asset at each trading date and held to the next trading date.

t_0, t_1, \dots, t_{n-1} : as the trading dates in the asset.

$I(t)$: the gain from trading at each time t .

$$I(t) = \Delta(t_0)[W(t) - W(t_0)] = \Delta(0)W(t), \quad 0 \leq t \leq t_1,$$

$$I(t) = \Delta(0)W(t_1) + \Delta(t_1)[W(t) - W(t_1)], \quad t_1 \leq t \leq t_2,$$

$$I(t) = \Delta(0)W(t_1) + \Delta(t_1)[W(t_2) - W(t_1)] + \Delta(t_2)[W(t) - W(t_2)], \quad t_2 \leq t \leq t_3,$$

In general, if $t_k \leq t \leq t_{k+1}$

$$I(t) = \sum_{j=0}^{k-1} \Delta(t_j)[W(t_{j+1}) - W(t_j)] + \Delta(t_k)[W(t) - W(t_k)] \rightarrow I(t) = \int_0^T \Delta(u) dW(u)$$

discrete

continuous

Theorem 4.2.1. *The Itô integral defined by (4.2.2) is a **martingale**.*

$$\mathbb{E}[I(t) \mid \mathcal{F}(s)] = I(s) \quad \forall s \leq t$$

proof:

Let $0 \leq s \leq t \leq T$ be given. We shall assume that s and t are in different subintervals of the partition Π (i.e., there are partition points t_ℓ and t_k such that $t_\ell < t_k$, $s \in [t_\ell, t_{\ell+1})$ and $t \in [t_k, t_{k+1})$).

If s and t are in the same subinterval, the following proof simplifies. Equation (4.2.2) may be rewritten as

$$\begin{aligned}
 I(t) = & \underbrace{\sum_{j=0}^{\ell-1} \Delta(t_j)[W(t_{j+1}) - W(t_j)]}_{\textcircled{1}} + \underbrace{\Delta(t_\ell)[W(t_{\ell+1}) - W(t_\ell)]}_{\textcircled{2}} \quad \text{0~s 的部分} \\
 & + \underbrace{\sum_{j=\ell+1}^{k-1} \Delta(t_j)[W(t_{j+1}) - W(t_j)]}_{\textcircled{3}} + \underbrace{\Delta(t_k)[W(t) - W(t_k)]}_{\textcircled{4}} \quad \text{s~t 的部分}
 \end{aligned}$$

proof :

The latest time appearing in this sum is t_ℓ and $t_\ell \leq s$, so every random variable in the first sum is $\mathcal{F}(s)$ – *measurable*. Therefore,

$$\textcircled{1} \quad \mathbb{E} \left[\sum_{j=0}^{\ell-1} \Delta(t_j) [W(t_{j+1}) - W(t_j)] \middle| \mathcal{F}(s) \right] = \sum_{j=0}^{\ell-1} \Delta(t_j) [W(t_{j+1}) - W(t_j)]$$

For the second term, we “take out what is known” and use the martingale property of W to write

By Theorem 3.3.4. Brownian motion is a martingale

$$\begin{aligned} \textcircled{2} \quad \mathbb{E} \left[\Delta(t_\ell) (W(t_{\ell+1}) - W(t_\ell)) \middle| \mathcal{F}(s) \right] &= \Delta(t_\ell) \left(\mathbb{E} \left[W(t_{\ell+1}) \middle| \mathcal{F}(s) \right] - W(t_\ell) \right) \\ &= \Delta(t_\ell) (W(s) - W(t_\ell)) \end{aligned}$$

proof:

The summands in the third term are of the form $\Delta(t_j)[W(t_{j+1}) - W(t_j)]$, where $t_j \geq t_{\ell+1} \geq s$. This permits us to use the following **iterated conditioning trick**, which is based on properties (iii) iterated conditioning and (ii) taking out what is known

$$\begin{aligned} \textcircled{3} \quad \mathbb{E} \left\{ \Delta(t_j) \left(W(t_{j+1}) - W(t_j) \right) \middle| \mathcal{F}(s) \right\} &= \mathbb{E} \left\{ \mathbb{E} \left[\Delta(t_j) \left(W(t_{j+1}) - W(t_j) \right) \middle| \mathcal{F}(t_j) \right] \middle| \mathcal{F}(s) \right\} \\ &= \mathbb{E} \left\{ \Delta(t_j) \left(\mathbb{E} \left[W(t_{j+1}) \middle| \mathcal{F}(t_j) \right] - W(t_j) \right) \middle| \mathcal{F}(s) \right\} \\ &= \mathbb{E} \left\{ \Delta(t_j) \left(W(t_j) - W(t_j) \right) \middle| \mathcal{F}(s) \right\} = 0 \end{aligned}$$

✱ **iterated conditioning:**
 $\forall I_1 \subseteq I_2, \mathbb{E}[\mathbb{E}(Y|I_2)|I_1] = \mathbb{E}(Y|I_1)$

$$\Rightarrow \mathbb{E} \left[\sum_{j=\ell+1}^{k-1} \Delta(t_j) [W(t_{j+1}) - W(t_j)] \middle| \mathcal{F}(s) \right] = 0$$

proof :

The fourth term

$$\begin{aligned} \textcircled{4} \mathbb{E} \left\{ \Delta(t_k) (W(t) - W(t_k)) \middle| \mathcal{F}(s) \right\} &= \mathbb{E} \left\{ \mathbb{E} \left[\Delta(t_k) (W(t) - W(t_k)) \middle| \mathcal{F}(t_k) \right] \middle| \mathcal{F}(s) \right\} \\ &= \mathbb{E} \left\{ \Delta(t_k) \left(\mathbb{E} [W(t) \middle| \mathcal{F}(t_k)] - W(t_k) \right) \middle| \mathcal{F}(s) \right\} \\ &= \mathbb{E} \left\{ \Delta(t_k) (W(t_k) - W(t_k)) \middle| \mathcal{F}(s) \right\} = 0 \end{aligned}$$

$$\begin{aligned} \mathbb{E} [I(t) \middle| \mathcal{F}(s)] &= \sum_{j=0}^{\ell-1} \Delta(t_j) [W(t_{j+1}) - W(t_j)] + \Delta(t_\ell) [W(t_{\ell+1}) - W(t_\ell)] + \sum_{j=\ell+1}^{k-1} \Delta(t_j) [W(t_{j+1}) - W(t_j)] + \Delta(t_k) [W(t) - W(t_k)] \\ &= \sum_{j=0}^{\ell-1} \Delta(t_j) [W(t_{j+1}) - W(t_j)] + \Delta(t_\ell) (W(s) - W(t_\ell)) = I(s) \end{aligned}$$

proof :

Because $I(t)$ is a martingale and $I(0) = 0$,

$$\mathbb{E} \left[I(t) \mid \mathcal{F}(s) \right] = I_0 \Rightarrow \mathbb{E} \left\{ \mathbb{E} \left[I(t) \mid \mathcal{F}(0) \right] \right\} = \mathbb{E}(I_0)$$

$$\Rightarrow \mathbb{E} I(t) = \mathbb{E}(I_0) = \mathbb{E}(0) = 0$$

$$\Rightarrow \text{Var } I(t) = \mathbb{E} I^2(t) - \left[\mathbb{E} I(t) \right]^2 = \mathbb{E} I^2(t)$$

Theorem 4.2.2. (Itô isometry). The Itô integral defined by (4.2.2) satisfies

$$\mathbb{E} I^2(t) = \mathbb{E} \int_0^t \Delta^2(u) du$$

proof :

Let $D_j = W(t_{j+1}) - W(t_j)$ for $j = 0, \dots, k-1$ and $D_k = W(t) - W(t_k)$

so that $I(t) = \sum_{j=0}^{k-1} \Delta(t_j)[W(t_{j+1}) - W(t_j)] + \Delta(t_k)[W(t) - W(t_k)] = \sum_{j=0}^k \Delta(t_j) D_j$

$$I^2(t) = \left(\sum_{j=0}^k \Delta(t_j) D_j \right)^2 = \overset{\textcircled{1}}{\sum_{j=0}^k \Delta^2(t_j) D_j^2} + \overset{\textcircled{2}}{2 \sum_{0 \leq i < j \leq k} \Delta(t_i) \Delta(t_j) D_i D_j}$$

proof :

show that the expected value of each of the cross terms is zero. For $i < j$, the random variable $\Delta(t_i)\Delta(t_j)D_i$ is $\mathcal{F}(t_j)$ – *measurable*, while the Brownian increment D_j is independent of $\mathcal{F}(t_j)$. Furthermore, $\mathbb{E}D_j = 0$. Therefore

$$\mathbb{E}[\Delta(t_i)\Delta(t_j)D_iD_j] = \mathbb{E}[\Delta(t_i)\Delta(t_j)D_i] \cdot \mathbb{E}D_j = \mathbb{E}[\Delta(t_i)\Delta(t_j)D_i] \cdot 0 = 0.$$

※ D_j is independent of $\mathcal{F}(t_j)$

$$D_j = W(t_{j+1}) - W(t_j)$$

$\Rightarrow D_j$ 不是 $\mathcal{F}(t_j)$ – *measurable*

$\Rightarrow D_j$ is independent of $\mathcal{F}(t_j)$

※ $\mathbb{E}D_j = 0$

D_j : 布朗運動的增量

$$D_j = W(t_{j+1}) - W(t_j) \sim N(0, t_{j+1} - t_j)$$

$$\mathbb{E}D_j = 0, \text{Var}(D_j) = \mathbb{E}D_j^2 = t_{j+1} - t_j$$

proof :

consider the square terms $\Delta^2(t_j) D_j^2$. The random variable $\Delta^2(t_j)$ is $\mathcal{F}(t_j)$ – *measurable* , and the squared Brownian increment D_j^2 is independent of $\mathcal{F}(t_j)$. Furthermore, $\mathbb{E}D_j^2 = t_{j+1} - t_j$ for $j = 0, \dots, k-1$ and $\mathbb{E}D_k^2 = t - t_k$. Therefore,

$$\begin{aligned}\mathbb{E}I^2(t) &= \sum_{j=0}^k \mathbb{E}[\Delta^2(t_j) D_j^2] = \sum_{j=1}^k \mathbb{E}\Delta^2(t_j) \cdot \mathbb{E}D_j^2 \\ &= \sum_{j=1}^{k-1} \mathbb{E}\Delta^2(t_j) (t_{j+1} - t_j) + \mathbb{E}\Delta^2(t_k)(t - t_k) .\end{aligned}$$

proof :

But $\Delta(t_j)$ is constant on the interval $[t_j, t_{j+1})$, and hence $\Delta^2(t_j)(t_{j+1} - t_j) = \int_{t_j}^{t_{j+1}} \Delta^2(u) du$.

Similarly, $\Delta^2(t_k)(t - t_k) = \int_{t_k}^t \Delta^2(u) du$.

$$\mathbb{E}I^2(t) = \sum_{j=0}^k \mathbb{E} \left[\Delta^2(t_j) \right] \times \mathbb{E}D_j^2 = \sum_{j=0}^{k-1} \mathbb{E} \left[\Delta^2(t_j) \right] \times (t_{j+1} - t_j) + \mathbb{E} \left[\Delta^2(t_k) \right] \times (t - t_k)$$

$$= \sum_{j=0}^{k-1} \mathbb{E} \left[\int_{t_j}^{t_{j+1}} \Delta^2(u) du \right] + \mathbb{E} \left[\int_{t_k}^t \Delta^2(u) du \right] = \mathbb{E} \left[\sum_{j=0}^{k-1} \int_{t_j}^{t_{j+1}} \Delta^2(u) du + \int_{t_k}^t \Delta^2(u) du \right] = \mathbb{E} \int_0^t \Delta^2(u) du.$$

Theorem 4.2.3. The quadratic variation accumulated up to time by the Itô integral

$$[I, I](t) = \int_0^t \Delta^2(u) du .$$

Recall : Theorem 3.4.3 Let W be a Brownian motion. Then $[W, W](T) = T$ for all $T \geq 0$

proof :

We first compute the quadratic variation accumulated by the Itô integral on one of the subintervals $[t_j, t_{j+1}]$ on which $\Delta(u)$ is constant. Choose partition points

$t_j = s_0 < s_1 < \dots < s_m = t_{j+1}$, and consider

$$\sum_{i=0}^{m-1} [I(s_{i+1}) - I(s_i)]^2 = \sum_{i=0}^{m-1} \left[\Delta(t_j) (W(s_{i+1}) - W(s_i)) \right]^2 = \Delta^2(t_j) \sum_{i=0}^{m-1} (W(s_{i+1}) - W(s_i))^2$$

$$* I(s_{i+1}) - I(s_i) = \int_{s_i}^{s_{i+1}} \Delta(t_j) dW(u) = \Delta(t_j) [W(s_{i+1}) - W(s_i)] \quad \text{布朗運動的增量性質}$$

proof :

As $m \rightarrow \infty$ and the step size $\max_{i=0, \dots, m-1} (s_{i+1} - s_i)$ approaches zero

$$\sum_{i=0}^{m-1} \left[(W(s_{i+1}) - W(s_i)) \right]^2 \rightarrow \sum_{i=0}^{m-1} (s_{i+1} - s_i) = t_{j+1} - t_j$$

$$\Delta^2(t_j) \sum_{i=0}^{m-1} (W(s_{i+1}) - W(s_i))^2 = \Delta^2(t_j)(t_{j+1} - t_j) = \int_{t_j}^{t_{j+1}} \Delta^2(u) du$$

We can get the quadratic variation accumulated by the Itô integral on intervals $[0, t]$

$$[I, I](t) = \int_0^t \Delta^2(u) du .$$

Quadratic variation:

1. Computed path-by-path, the result can depend on the path.
2. Depends on the size of the positions we take

We choose large positions $\Delta(u)$, the Itô integral will have a large quadratic variation.

Choose small positions $\Delta(u)$ and the Itô integral would have a small quadratic variation.

3. Regarded as a measure of risk.

The variance of $I(t)$

1. Average over all possible paths of the quadratic variation.
2. Cannot be random.
3. More theoretical concept than quadratic variation.

Empirical variance

1. Computed from a realized path and is an estimator of the theoretical variance.

Remark 4.2.4. (on notation). The notations

$I(t) = \int_0^t \Delta(u) dW(u) \quad (4.2.11)$	integral form
$I(t) = I(0) + \int_0^t \Delta(u) dW(u) \quad (4.2.13)$	
$dI(t) = \Delta(t)dW(t) \quad (4.2.12)$	differential form

Recall : $dW(t) dW(t) = dt \Leftrightarrow [W, W](t) = t, t \geq 0$

$$dI(t) dI(t) = \Delta(t)dW(t) \times \Delta(t)dW(t) = \Delta^2(t)[dW(t)]^2 = \Delta^2(t)dt$$

$dI(t) dI(t) = [I, I](t) = \Delta^2(t)dt$ is another way of the result of Theorem 4.2.3

※ Theorem4.2.3

$$[I, I](t) = \int_0^t \Delta^2(u) du$$