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## A Stochastic Integral by a Near-Martingale

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## A STOCHASTIC INTEGRAL BY A NEAR-MARTINGALE

SHINYA HIBINO, HUI-HSIUNG KUO, AND KIMIYAKI SAITÔ\*

ABSTRACT. In this paper we discuss the new stochastic integral in [1] in terms of the Itô isometry. We prove the Doob-Meyer decomposition theorem for near-submartingales in the classes  $(D)$  and  $(DL)$ . Moreover, we introduce a stochastic integral by a near-martingale as an application of the decomposition theorem.

### 1. Introduction

A new stochastic integral was introduced in [1]. The Itô isometry based on the new integral for special processes was discussed in [10]. The Doob-Meyer decomposition theorem for continuous near-submartingales was also discussed in [3]. This stochastic integral has been studied from different points of view [2, 4, 7, 8, 9] and references cited therein.

Let  $(\Omega, \mathcal{F}, P; \mathcal{F}_t)_{a \leq t \leq b}$  be a basic probability space with a filtration  $\{\mathcal{F}_t\}_{a \leq t \leq b}$ , and  $B = \{B(t); a \leq t \leq b\}$  a  $\{\mathcal{F}_t\}$ -Brownian motion on  $(\Omega, \mathcal{F}, P)$ . A stochastic process  $g = \{g(t); a \leq t \leq b\}$  is called to be instantly independent of  $\{\mathcal{F}_t\}$  if  $g(t)$  is independent of  $\mathcal{F}_t$  for all  $t \in [a, b]$ . A stochastic process  $g = \{g(t); a \leq t \leq b\}$  is called to be in  $L_{\text{ind}}^2([a, b] \times \Omega)$  if the process satisfies the following conditions:

- (1)  $g = \{g(t); a \leq t \leq b\}$  is instantly independent of  $\{\mathcal{F}_t\}$ .
- (2)  $\int_a^b E[|g(t)|^2] dt < \infty$ .
- (3)  $g$  is right-continuous in  $t$ .

A stochastic process  $g = \{g(t); a \leq t \leq b\}$  is called to be in  $\mathcal{L}_{\text{ind}}(\Omega, L^2[a, b])$  if the process satisfies the following conditions:

- (1)  $g = \{g(t); a \leq t \leq b\}$  is instantly independent of  $\{\mathcal{F}_t\}$ .
- (2)  $\int_a^b |g(t)|^2 dt < \infty$ , a. e.

In this article we discuss the new stochastic integral through the Itô isometry. In Section 2 we discuss the stochastic integral by the Brownian motion  $B$  for processes in  $L_{\text{ind}}^2([a, b] \times \Omega)$  through the Itô isometry with its properties. In Section 3 we extend the stochastic integral to that on a class  $\mathcal{L}_{\text{ind}}(\Omega, L^2[a, b])$  which is larger than the space in Section 2. In Section 4 we give the proof of the Doob-Meyer decomposition theorem for near-submartingales in the classes  $(D)$  and  $(DL)$ . This theorem is important to discuss the new integral in [1] for its extension. In the last

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section we introduce a stochastic integral by a near-martingale as an application of the decomposition theorem. This is a formulation of the new integral in [1] from the point of view of the stochastic integral by the near-martingale.

## 2. Stochastic Integrals on $L_{\text{ind}}^2([a, b] \times \Omega)$

Let  $g$  be in  $L_{\text{ind}}^2([a, b] \times \Omega)$ . Then  $g$  is called to be an instantly independent step process if there exist a partition  $a = t_0 < t_1 < \dots < t_n = b$  and instantly independent random variables  $\eta_i$ ,  $i = 1, 2, \dots, n$  with  $E[\eta_i^2] < \infty$  such that

$$g(t, \omega) = \sum_{i=1}^n \eta_i(\omega) 1_{[t_{i-1}, t_i)}(t), \quad \omega \in \Omega, t \in [a, b]. \quad (2.1)$$

We denote the set of all instantly independent step processes by  $\text{Step}_{\text{ind}}([a, b] \times \Omega)$ .

For any  $g \in \text{Step}_{\text{ind}}([a, b] \times \Omega)$  given as (2.1), we define  $\mathcal{J}(g)$  by

$$\mathcal{J}(g) := \sum_{i=1}^n \eta_i(B(t_i) - B(t_{i-1})).$$

Then we have the following.

**Lemma 2.1.** *For any  $g, h \in \text{Step}_{\text{ind}}([a, b] \times \Omega)$  and  $a, b \in \mathbb{R}$ , it holds that*

$$\mathcal{J}(ag + bh) = a\mathcal{J}(g) + b\mathcal{J}(h).$$

**Lemma 2.2.** *For any  $g \in \text{Step}_{\text{ind}}([a, b] \times \Omega)$ , the following equalities hold.*

- (1)  $E[\mathcal{J}(g)] = 0$ .
- (2)  $E[|\mathcal{J}(g)|^2] = \int_a^b E[|g(t)|^2] dt$ .

*Proof.* Let  $g$  be a function in  $\text{Step}_{\text{ind}}([a, b] \times \Omega)$  given as (2.1).

(1): Since, for any  $1 \leq i \leq n$ ,  $\eta_i$  is independent to  $B(t_i) - B(t_{i-1})$ , we have

$$E[\eta_i(B(t_i) - B(t_{i-1}))] = E[\eta_i]E[B(t_i) - B(t_{i-1})] = 0.$$

Therefore,  $E[\mathcal{J}(g)] = 0$ .

(2): If  $i < j$ , we have

$$\begin{aligned} & E[\eta_i \eta_j (B(t_i) - B(t_{i-1}))(B(t_j) - B(t_{j-1}))] \\ &= E[(B(t_i) - B(t_{i-1}))] E[\eta_i \eta_j (B(t_j) - B(t_{j-1}))] = 0. \end{aligned}$$

If  $i = j$ , we have

$$\begin{aligned} E[\eta_i^2 (B(t_i) - B(t_{i-1}))^2] &= E[(B(t_i) - B(t_{i-1}))^2] E[\eta_i^2] \\ &= (t_i - t_{i-1}) E[\eta_i^2]. \end{aligned}$$

Therefore, we obtain

$$E[|\mathcal{J}(g)|^2] = \sum_{i,j=1}^n E[\eta_i \eta_j (B(t_i) - B(t_{i-1}))(B(t_j) - B(t_{j-1}))] = \int_a^b E[|g(t)|^2] dt.$$

□

**Lemma 2.3.** For any  $g \in L^2_{\text{ind}}([a, b] \times \Omega)$ , there exists  $\{g_n\}_{n=1}^{\infty} \subset \text{Step}_{\text{ind}}([a, b] \times \Omega)$  such that

$$\lim_{n \rightarrow \infty} \int_a^b E[|g(t) - g_n(t)|^2] dt = 0$$

holds.

Let  $g \in L^2_{\text{ind}}([a, b] \times \Omega)$ . Then there exists  $\{g_n\}_{n=1}^{\infty} \subset \text{Step}_{\text{ind}}([a, b] \times \Omega)$  such that

$$\lim_{n \rightarrow \infty} \int_a^b E[|g(t) - g_n(t)|^2] dt = 0.$$

By Lemmas 2.1, 2.2 and 2.3, we have

$$E[|\mathcal{J}(g_n) - \mathcal{J}(g_m)|^2] = \int_a^b E[|g_n(t) - g_m(t)|^2] dt \xrightarrow{n, m \rightarrow \infty} 0.$$

Therefore,  $\{\mathcal{J}(g_n)\}_{n=1}^{\infty}$  is a Cauchy sequence in  $L^2(\Omega)$ . By the completeness of  $L^2(\Omega)$ , there exists  $\mathcal{J}(g) \in L^2(\Omega)$  such that

$$\mathcal{J}(g) = \lim_{n \rightarrow \infty} \mathcal{J}(g_n), \quad \text{in } L^2(\Omega).$$

Thus we can define the stochastic integral  $\int_a^b g(t) dB(t)$  by

$$\int_a^b g(t) dB(t) := \mathcal{J}(g)$$

as an element of  $L^2(\Omega)$ . This is well-defined. In fact, assume that there exist  $\{g_n(t)\}_{n=0}^{\infty}, \{h_n(t)\}_{n=0}^{\infty} \subset \text{Step}_{\text{ind}}([a, b] \times \Omega)$  such that

$$\lim_{n \rightarrow \infty} \int_a^b E[|g(t) - g_n(t)|^2] dt = 0, \quad \lim_{n \rightarrow \infty} \int_a^b E[|g(t) - h_n(t)|^2] dt = 0.$$

Then we can see that

$$\begin{aligned} E[|\mathcal{J}(g_n) - \mathcal{J}(h_n)|^2] &= \int_a^b E[|g_n(t) - h_n(t)|^2] dt \\ &= \int_a^b E[|g(t) - g_n(t)|^2] dt + \int_a^b E[|g(t) - h_n(t)|^2] dt \\ &\xrightarrow{n \rightarrow \infty} 0. \end{aligned}$$

Therefore,  $\lim_{n \rightarrow \infty} \mathcal{J}(g_n) = \lim_{n \rightarrow \infty} \mathcal{J}(h_n)$  in  $L^2(\Omega)$ .

**Theorem 2.4.** For any  $g \in L^2_{\text{ind}}([a, b] \times \Omega)$ ,  $\mathcal{J}(g)$  has the following properties:

- (1)  $E[\mathcal{J}(g)] = 0$ .
- (2)  $E[|\mathcal{J}(g)|^2] = \int_a^b E[|g(t)|^2] dt$ .

*Proof.* (1) follows from  $E[\mathcal{J}(g)] = \lim_{n \rightarrow \infty} E[\mathcal{J}(g_n)] = 0$ . (2) follows from

$$E[|\mathcal{J}(g)|^2] = \lim_{n \rightarrow \infty} E[|\mathcal{J}(g_n)|^2] = \lim_{n \rightarrow \infty} \int_a^b E[|g_n(t)|^2] dt = \int_a^b E[|g(t)|^2] dt.$$

□

**Corollary 2.5.** For any  $g, h \in L^2_{\text{ind}}([a, b] \times \Omega)$ , the equality

$$E \left[ \int_a^b g(t)dB(t) \int_a^b h(t)dB(t) \right] = \int_a^b E[g(t)h(t)]dt$$

holds.

*Proof.* By Theorem 2.4, we have

$$E \left[ \left| \int_a^b g(t)dB(t) + \int_a^b h(t)dB(t) \right|^2 \right] = \int_a^b E[|g(t) + h(t)|^2]dt.$$

Then we can see that

$$\begin{aligned} & E \left[ \left| \int_a^b g(t)dB(t) + \int_a^b h(t)dB(t) \right|^2 \right] \\ &= E \left[ \left( \int_a^b g(t)dB(t) \right)^2 \right. \\ &\quad \left. + 2 \left( \int_a^b g(t)dB(t) \right) \left( \int_a^b h(t)dB(t) \right) + \left( \int_a^b h(t)dB(t) \right)^2 \right] \\ &= \int_a^b E[|g(t)|^2]dB(t) \\ &\quad + 2E \left[ \int_a^b g(t)dB(t) \int_a^b h(t)dB(t) \right] + \int_a^b E[|h(t)|^2]dB(t). \end{aligned}$$

On the other hand, we get

$$\begin{aligned} & \int_a^b E[|g(t) + h(t)|^2]dt \\ &= \int_a^b E[|g(t)|^2]dB(t) + 2 \int_a^b E[g(t)h(t)]dt + \int_a^b E[|h(t)|^2]dB(t). \end{aligned}$$

Consequently, we obtain

$$E \left[ \int_a^b g(t)dB(t) \int_a^b h(t)dB(t) \right] = \int_a^b E[g(t)h(t)]dt.$$

□

**Example 2.6.** For any  $g \in L^2_{\text{ind}}([a, b] \times \Omega)$ , the stochastic process

$$\left\{ \int_t^b g(s)dB(s); a \leq t \leq b \right\}$$

is an instantly independent process of  $\{\mathcal{F}_t\}$ .

### 3. Stochastic Integrals on $\mathcal{L}_{\text{ind}}(\Omega, L^2[a, b])$

**Lemma 3.1.** *For any  $g \in \mathcal{L}_{\text{ind}}(\Omega, L^2[a, b])$ , there exists a sequence  $\{g_n\}_{n=0}^\infty \subset L^2_{\text{ind}}([a, b] \times \Omega)$  such that*

$$\lim_{n \rightarrow \infty} \int_a^b |g_n(t) - g(t)|^2 dt = 0, \quad \text{a. e.}$$

*Proof.* For any  $n \in \mathbb{N}$ , we set

$$g_n(t, \omega) = \begin{cases} g(t, \omega), & \int_t^b |g(s, \omega)|^2 ds \leq n, \\ 0, & \int_t^b |g(s, \omega)|^2 ds > n. \end{cases}$$

Then  $\{g_n(t); a \leq t \leq b\}$  is instantly independent of  $\{\mathcal{F}_t\}$  and

$$\int_a^b |g_n(t, \omega)|^2 dt = \int_{\tau_n(\omega)}^b |g(t, \omega)|^2 dt, \quad \text{a. e. } \omega$$

holds, where  $\tau_n(\omega) = \inf \left\{ t; \int_t^b |g(s, \omega)|^2 ds \leq n \right\}$ . Therefore, we have

$$\int_a^b |g_n(t)|^2 dt \leq n, \quad \text{a. e. } \omega.$$

Since  $\int_a^b E[|g_n(t)|^2] dt \leq n$  and  $g \in \mathcal{L}_{\text{ind}}(\Omega, L^2[a, b])$ , it holds that

$$\int_a^b |g(t, \omega)|^2 dt \leq n, \quad \text{a. e. } \omega \in \Omega$$

for a large  $n$ . Then we have  $g(t, \omega) = g_n(t, \omega)$  for all  $t \in [a, b]$ . Consequently, we obtain

$$\lim_{n \rightarrow \infty} \int_a^b |g_n(t, \omega) - g(t, \omega)|^2 dt = 0, \quad \text{a. e. } \omega.$$

□

**Lemma 3.2.** *Let  $g \in \text{Step}_{\text{ind}}([a, b] \times \Omega)$ . Then for any  $\epsilon > 0$ , there exists  $c > 0$  such that*

$$P \left( \left| \int_a^b g(t) dB(t) \right| > \epsilon \right) \leq \frac{c}{\epsilon^2} + P \left( \int_a^b |g(t)|^2 dt > c \right).$$

*Proof.* For any  $c > 0$ , we define  $g_c(t, \omega)$  by

$$g_c(t, \omega) = \begin{cases} g(t, \omega), & \int_t^b |g(s, \omega)|^2 ds \leq c, \\ 0, & \int_t^b |g(s, \omega)|^2 ds > c. \end{cases}$$

Since

$$\begin{aligned} & \left\{ \left| \int_a^b g(t) dB(t) \right| > \epsilon \right\} \\ & \subset \left\{ \left| \int_a^b g_c(t) dB(t) \right| > \epsilon \right\} \cup \left\{ \int_a^b g(t) dB(t) \neq \int_a^b g_c(t) dB(t) \right\}, \end{aligned}$$

for any  $\epsilon > 0$  and  $c > 0$ , we have

$$\begin{aligned} & P \left( \left| \int_a^b g(t) dB(t) \right| > \epsilon \right) \\ & \leq P \left( \left| \int_a^b g_c(t) dB(t) \right| > \epsilon \right) + P \left( \int_a^b g(t) dB(t) \neq \int_a^b g_c(t) dB(t) \right). \end{aligned}$$

Then since  $g \in \text{Step}_{\text{ind}}([a, b] \times \Omega)$ , we have

$$\left\{ \int_a^b g(t) dB(t) \neq \int_a^b g_c(t) dB(t) \right\} \subset \left\{ \int_a^b |g(t)|^2 dt > c \right\}$$

Therefore,

$$P \left( \left| \int_a^b g(t) dB(t) \right| > \epsilon \right) \leq P \left( \left| \int_a^b g_c(t) dB(t) \right| > \epsilon \right) + P \left( \int_a^b |g(t)|^2 dt > c \right).$$

By the Chebyshev inequality, we obtain

$$\begin{aligned} & P \left( \left| \int_a^b g(t) dB(t) \right| > \epsilon \right) \leq \frac{1}{\epsilon^2} E \left[ \left| \int_a^b g_c(t) dB(t) \right|^2 \right] + P \left( \int_a^b |g(t)|^2 dt > c \right) \\ & = \frac{1}{\epsilon^2} \int_a^b E[|g_c(t)|^2] dt + P \left( \int_a^b |g(t)|^2 dt > c \right) \\ & \leq \frac{c}{\epsilon^2} + P \left( \int_a^b |g(t)|^2 dt > c \right). \end{aligned}$$

□

**Lemma 3.3.** *For any  $g \in \mathcal{L}_{\text{ind}}(\Omega, L^2[a, b])$ , there exists a sequence  $\{g_n\}_{n=1}^{\infty} \subset \text{Step}_{\text{ind}}([a, b] \times \Omega)$  such that*

$$\lim_{n \rightarrow \infty} \int_a^b |g_n(t) - g(t)|^2 dt = 0, \quad \text{in probability.}$$

*Proof.* By Lemma 3.1, for any  $g \in \mathcal{L}_{\text{ind}}(\Omega, L^2[a, b])$ , we can take  $\{h_n\}_{n=1}^{\infty} \subset L^2_{\text{ind}}([a, b] \times \Omega)$  such that

$$\lim_{n \rightarrow \infty} \int_a^b |h_n(t) - g(t)|^2 dt = 0, \quad \text{in probability.}$$

For any  $n$ , applying Lemma 2.3 to  $h_n$ , there exists  $\{g_n\}_{n=1}^{\infty} \subset \text{Step}_{\text{ind}}([a, b] \times \Omega)$  such that

$$E \left[ \int_a^b |g_n(t) - h_n(t)|^2 dt \right] < \frac{1}{n}.$$



Then we have

$$\begin{aligned} & \left\{ \int_a^b |g_n(t) - g(t)|^2 dt > \varepsilon \right\} \\ & \subset \left\{ \int_a^b |g_n(t) - h_n(t)|^2 dt > \frac{\varepsilon}{4} \right\} \cup \left\{ \int_a^b |h_n(t) - g(t)|^2 dt > \frac{\varepsilon}{4} \right\} \end{aligned}$$

for all  $\varepsilon > 0$ . Hence, for all  $\varepsilon > 0$ ,

$$\begin{aligned} & P \left( \int_a^b |g_n(t) - g(t)|^2 dt > \varepsilon \right) \\ & \leq P \left( \int_a^b |g_n(t) - h_n(t)|^2 dt > \frac{\varepsilon}{4} \right) + P \left( \int_a^b |h_n(t) - g(t)|^2 dt > \frac{\varepsilon}{4} \right). \end{aligned}$$

Therefore, by the Chebyshev inequality,

$$\begin{aligned} & P \left( \int_a^b |g_n(t) - g(t)|^2 dt > \varepsilon \right) \\ & \leq \frac{4}{\varepsilon} E \left[ \int_a^b |g_n(t) - h_n(t)|^2 dt \right] + P \left( \int_a^b |h_n(t) - g(t)|^2 dt > \frac{\varepsilon}{4} \right) \\ & \leq \frac{4}{n\varepsilon} + P \left( \int_a^b |h_n(t) - g(t)|^2 dt > \frac{\varepsilon}{4} \right) \end{aligned}$$

for all  $\varepsilon > 0$ . Consequently, we obtain

$$\lim_{n \rightarrow \infty} P \left( \int_a^b |g_n(t) - g(t)|^2 dt > \varepsilon \right) = 0$$

for all  $\varepsilon > 0$ . This means the assertion:

$$\lim_{n \rightarrow \infty} \int_a^b |g_n(t) - g(t)|^2 dt = 0, \quad \text{in probability.}$$

□

By Lemma 3.3, for any  $g \in \mathcal{L}_{\text{ind}}(\Omega, L^2[a, b])$ , there exists  $\{g_n\}_{n=1}^\infty \subset \text{Step}_{\text{ind}}([a, b] \times \Omega)$  such that

$$\lim_{n \rightarrow \infty} \int_a^b |g_n(t) - g(t)|^2 dt = 0, \quad \text{in probability.}$$

Then by Lemma 3.2, for any  $\varepsilon > 0$ ,

$$P(|\mathcal{J}(g_n) - \mathcal{J}(g_m)| > \varepsilon) \leq \frac{\varepsilon}{2} + P \left( \int_a^b |g_n(t) - g_m(t)|^2 dt > \frac{\varepsilon^3}{2} \right).$$

Since

$$\begin{aligned} & \left\{ \int_a^b |g_n(t) - g_m(t)|^2 dt > \frac{\varepsilon^3}{2} \right\} \\ & \subset \left\{ \int_a^b |g_n(t) - g(t)|^2 dt > \frac{\varepsilon^3}{8} \right\} \cup \left\{ \int_a^b |g_m(t) - g(t)|^2 dt > \frac{\varepsilon^3}{8} \right\}, \end{aligned}$$

we have

$$\begin{aligned} & P \left( \int_a^b |g_n(t) - g_m(t)|^2 dt > \frac{\varepsilon^3}{2} \right) \\ & \leq P \left( \int_a^b |g_n(t) - g(t)|^2 dt > \frac{\varepsilon^3}{8} \right) + P \left( \int_a^b |g_m(t) - g(t)|^2 dt > \frac{\varepsilon^3}{8} \right). \end{aligned}$$

Hence, since there exists  $N \in \mathbb{N}$  such that

$$P \left( \int_a^b |g_n(t) - g(t)|^2 dt > \frac{\varepsilon^3}{8} \right) < \frac{\varepsilon}{4}$$

for all  $n \geq N$  by Lemma 3.3, it holds that

$$P \left( \int_a^b |g_n(t) - g_m(t)|^2 dt > \frac{\varepsilon^3}{2} \right) < \frac{\varepsilon}{2}$$

for all  $n, m \geq N$ . Consequently, for any  $\varepsilon > 0$  there exists  $N \in \mathbb{N}$  such that

$$P(|\mathcal{J}(g_n) - \mathcal{J}(g_m)| > \varepsilon) < \varepsilon$$

for all  $n, m \geq N$ . This implies that  $\{\mathcal{J}(g_n)\}$  converges in probability. Thus we define the stochastic integral  $\int_a^b g(t)dB(t)$  by

$$\int_a^b g(t)dB(t) = \lim_{n \rightarrow \infty} \mathcal{J}(g_n), \quad \text{in probability.}$$

This is well-defined. In fact, suppose that there exist sequences  $\{g_n(t)\}_{n=0}^\infty$  and  $\{h_n(t)\}_{n=0}^\infty \subset \text{Step}_{\text{ind}}([a, b] \times \Omega)$  such that

$$\lim_{n \rightarrow \infty} \int_a^b |g(t) - g_n(t)|^2 dt = 0, \quad \lim_{n \rightarrow \infty} \int_a^b |g(t) - h_n(t)|^2 dt = 0 \quad \text{in probability.}$$

Then by Lemma 3.2, for any  $\varepsilon > 0$ , we have

$$P(|\mathcal{J}(g_n) - \mathcal{J}(h_n)| > \varepsilon) \leq \frac{\varepsilon}{2} + P \left( \int_a^b |g_n(t) - h_n(t)|^2 dt > \frac{\varepsilon^3}{2} \right).$$

Since

$$\begin{aligned} & \left\{ \int_a^b |g_n(t) - h_n(t)|^2 dt > \frac{\varepsilon^3}{2} \right\} \\ & \subset \left\{ \int_a^b |g_n(t) - g(t)|^2 dt > \frac{\varepsilon^3}{8} \right\} \cup \left\{ \int_a^b |h_n(t) - g(t)|^2 dt > \frac{\varepsilon^3}{8} \right\}, \end{aligned}$$

we have

$$\begin{aligned} & P\left(\int_a^b |g_n(t) - h_n(t)|^2 dt > \frac{\varepsilon^3}{2}\right) \\ & \leq P\left(\int_a^b |g_n(t) - g(t)|^2 dt > \frac{\varepsilon^3}{8}\right) + P\left(\int_a^b |h_n(t) - g(t)|^2 dt > \frac{\varepsilon^3}{8}\right). \end{aligned}$$

By Lemma 3.3, for any  $\varepsilon > 0$ , there exist  $N_1 \in \mathbb{N}$  and  $N_2 \in \mathbb{N}$  such that

$$P\left(\int_a^b |g_n(t) - g(t)|^2 dt > \frac{\varepsilon^3}{8}\right) < \frac{\varepsilon}{4}, \quad \text{for all } n \geq N_1,$$

and

$$P\left(\int_a^b |h_n(t) - g(t)|^2 dt > \frac{\varepsilon^3}{8}\right) < \frac{\varepsilon}{4}, \quad \text{for all } n \geq N_2.$$

Therefore, putting  $N = \max\{N_1, N_2\}$ , we have

$$P\left(\int_a^b |g_n(t) - h_n(t)|^2 dt > \frac{\varepsilon^3}{2}\right) < \frac{\varepsilon}{2}$$

for all  $n, m \geq N$ . Consequently,

$$P(|\mathcal{J}(g_n) - \mathcal{J}(h_n)| > \varepsilon) < \varepsilon$$

holds for all  $n \geq N$ . Thus, we obtain  $\lim_{n \rightarrow \infty} \mathcal{J}(g_n) = \lim_{n \rightarrow \infty} \mathcal{J}(h_n)$  in probability.

#### 4. The Doob-Meyer Decomposition by the Near-martingale

Let  $(\Omega, \mathcal{F}, P; \mathcal{F}_t)_{a \leq t \leq b}$  be a basic probability space with a filtration  $\{\mathcal{F}_t\}_{a \leq t \leq b}$ . A stochastic process  $\{X(t); a \leq t \leq b\}$  is called to be the near-martingale with filtration  $\{\mathcal{F}_t\}_{a \leq t \leq b}$  if it satisfies the following conditions:

- (1)  $E[|X(t)|] < \infty$  for all  $a \leq t \leq b$ ,
- (2)  $E[X(t)|\mathcal{F}_s] = E[X(s)|\mathcal{F}_s]$  for all  $s < t$ .

If the condition

- (3)  $E[X(t)|\mathcal{F}_s] \geq E[X(s)|\mathcal{F}_s]$  for all  $s < t$

holds instead of the condition (2), the stochastic process  $\{X(t); a \leq t \leq b\}$  is called to be the near-submartingale with the filtration  $\{\mathcal{F}_t\}_{a \leq t \leq b}$ .

**Theorem 4.1.** ([5]) *Let  $X = \{X(t); n \in \mathbb{N}\}$  be a near-submartingale. Then, there exist a near-martingale  $N = \{N(n); n \in \mathbb{N}\}$  and an increasing process  $A = \{A(n); n \in \mathbb{N}\}$  such that*

$$X(n) = N(n) + A(n), n \in \mathbb{N},$$

where  $A$  is called to be the increasing process if it satisfies the following conditions:

- (1)  $A(1) = 0$ ,
- (2) for each  $n \geq 2$ ,  $A(n)$  is  $\mathcal{F}_{n-1}$ -measurable,
- (3) for any  $m \leq n$ ,  $A(m) \leq A(n)$ , a. e.

**Theorem 4.2.** *Let  $X(t) = \int_t^b g(s)dB(s)$  for any  $a \leq t \leq b$  and  $g \in L_{\text{ind}}^2([a, b] \times \Omega)$ . Then the stochastic process  $\{X(t); a \leq t \leq b\}$  is a near-martingale with  $\{\mathcal{F}_t\}_{a \leq t \leq b}$ .*

*Proof.* Let  $g \in \text{Step}_{\text{ind}}([a, b] \times \Omega)$ . Then  $g$  has the form

$$g(u, \omega) = \sum_{i=1}^n \eta_i(\omega) 1_{[t_{i-1}, t_i)}(u), \quad s = t_0 < t_1 < \cdots < t_j = t < \cdots < t_n = b,$$

where  $\eta_i, i = 0, 1, 2, \dots, n$ , are random variables which independent to  $\mathcal{F}_{t_i}$  satisfying  $E[\eta_i^2] < \infty$ . Then we obtain

$$\begin{aligned} & E \left[ \int_t^b g(u) dB(u) \middle| \mathcal{F}_s \right] \\ &= E \left[ \sum_{i=j+1}^n \eta_i (B(t_i) - B(t_{i-1})) \middle| \mathcal{F}_s \right] \\ &= E \left[ \sum_{i=j+1}^n \eta_i (B(t_i) - B(t_{i-1})) \middle| \mathcal{F}_s \right] + \sum_{i=1}^j E[\eta_i] E[(B(t_i) - B(t_{i-1}))] \\ &= E \left[ \sum_{i=j+1}^n \eta_i (B(t_i) - B(t_{i-1})) \middle| \mathcal{F}_s \right] + E \left[ \sum_{i=1}^j \eta_i (B(t_i) - B(t_{i-1})) \middle| \mathcal{F}_s \right] \\ &= E \left[ \int_s^b g(u) dB(u) \middle| \mathcal{F}_s \right]. \end{aligned}$$

Next we prove the theorem in the case of  $g \in L^2_{\text{ind}}([a, b] \times \Omega)$ . By Lemma 2.3, there exists  $\{g_n\}_{n=1}^\infty \subset \text{Step}_{\text{ind}}([a, b] \times \Omega)$  such that  $\lim_{n \rightarrow \infty} \int_a^b E[|g(t) - g_n(t)|^2] dt = 0$ . Let

$$X^{(n)}(t) = \int_t^b g_n(u) dB(u), \quad n \in \mathbb{N}.$$

Then  $\{X^{(n)}(t); a \leq t \leq b\}$  is a near-martingale for each  $n \in \mathbb{N}$  from above argument. For any  $s < t$ , we have

$$E[X(t) - X(s) | \mathcal{F}_s] = E[X(t) - X^{(n)}(t) | \mathcal{F}_s] + E[X^{(n)}(s) - X(s) | \mathcal{F}_s].$$

Since

$$\begin{aligned} E[|E[X(t) - X^{(n)}(t) | \mathcal{F}_s]|^2] &\leq E[E[|X(t) - X^{(n)}(t)|^2 | \mathcal{F}_s]] \\ &= E[|X(t) - X^{(n)}(t)|^2] \\ &= \int_t^b E[|g(u) - g_n(u)|^2] du \\ &\leq \int_a^b E[|g(u) - g_n(u)|^2] du \xrightarrow{n \rightarrow \infty} 0, \end{aligned}$$

and by taking subsequence of  $\{X^{(n)}(t)\}$ , we get

$$E[X(t) - X^{(n)}(t) | \mathcal{F}_s] \xrightarrow{n \rightarrow \infty} 0, \quad \text{a. e.}$$

Similarly, we have

$$E[X(s) - X^{(n)}(s)|\mathcal{F}_s] \xrightarrow[n \rightarrow \infty]{} 0, \quad \text{a. e.}$$

Consequently, we obtain

$$E[X(t) - X(s)|\mathcal{F}_s] = 0, \quad \text{a. e.}$$

This implies

$$E[X(t)|\mathcal{F}_s] = E[X(s)|\mathcal{F}_s], \quad \text{a. e.}$$

□

From now on, we assume that the submartingale and the near-submartingale are right-continuous. Let  $\{\mathcal{F}_t; t \geq 0\}$  be a right-continuous filtration and set

$$\mathcal{F}_\infty = \bigvee_{t \geq 0} \mathcal{F}_t.$$

The Doob decomposition theorem for the near-submartingale is proved in [11]. In [3] the Doob-Meyer decomposition theorem is proved for the continuous near-submartingale. In this section we prove the Doob-Meyer decomposition theorem for the right-continuous near-submartingale.

**Definition 4.3.** Let  $X = \{X(t), t \in \mathbb{R}_+\}$  be a near-submartingale (respectively, near-martingale). Suppose there exists an  $\mathcal{F}_\infty$ -measurable and integrable random variable  $X(\infty)$  such that

$$E[X(t)|\mathcal{F}_t] \leq E[X(\infty)|\mathcal{F}_t], \quad (\text{respectively, } E[X(t)|\mathcal{F}_t] = E[X(\infty)|\mathcal{F}_t])$$

for all  $t \in \mathbb{R}_+$  ( $\equiv [0, \infty)$ ). Then we call  $\{X(t), t \in \overline{\mathbb{R}}_+ (\equiv [0, \infty])\}$  a *closed near-submartingale* (respectively, *closed near-martingale*).

**Definition 4.4.** An  $(\mathcal{F}_t)$ -adapted right-continuous process  $A = \{A(t); t \in \mathbb{R}_+\}$  is called an *increasing process* if  $A(t)$  is an increasing function in  $t$  and  $A(0) = 0$  almost surely.

**Definition 4.5.** An integrable increasing process  $A$  is called a *natural increasing process* if it satisfies the equality

$$E \left[ \int_0^t X(s) dA(s) \right] = E \left[ \int_0^t X(s-) dA(s) \right], \quad \forall t \in \mathbb{R}_+$$

for all bounded martingales  $X$ .

Let  $X = \{X(\lambda); \lambda \in \Lambda\}$  be a system of integrable random variables on a probability space  $(\Omega, \mathcal{F}, P)$ . If  $X$  satisfies

$$\sup_{\lambda \in \Lambda} \int_{|X(\lambda)| > c} |X(\lambda)| dP \xrightarrow[c \rightarrow \infty]{} 0,$$

then  $X$  is called to be uniformly integrable. A near-submartingale  $X = \{X(t), t \in \mathbb{R}_+\}$  is called to have the Doob-Meyer decomposition if  $X$  is expressed in the form

$$X(t) = N(t) + A(t), \quad \forall t \in \mathbb{R}_+$$

for some near-martingale  $N$  and natural increasing process  $A$ .

**Lemma 4.6.** *Let  $A, B$  be natural increasing processes. Then, if  $A - B$  is a near-martingale, for any bounded  $(\mathcal{F}_t)$ -adapted process  $f = \{f(t); t \geq 0\}$ , the equality*

$$E \left[ \int_0^t f(s) dA(s) \right] = E \left[ \int_0^t f(s) dB(s) \right]$$

holds.

*Proof.* Let  $N(t) = A(t) - B(t)$  for all  $t \in \mathbb{R}_+$ . Take a partition of  $[0, t]$ :

$$\delta := \{0 = t_0 < \cdots < t_n = t\}.$$

Then since  $N$  is a near-martingale, we get

$$\begin{aligned} & E \left[ \sum_{k=1}^n f(t_{k-1})(N(t_k) - N(t_{k-1})) \right] \\ &= E \left[ \sum_{k=1}^n E[f(t_{k-1})(N(t_k) - N(t_{k-1})) | \mathcal{F}_{t_{k-1}}] \right] \\ &= E \left[ \sum_{k=1}^n f(t_{k-1})(E[N(t_k) | \mathcal{F}_{t_{k-1}}] - E[N(t_{k-1}) | \mathcal{F}_{t_{k-1}}]) \right] = 0. \end{aligned}$$

Therefore,

$$E \left[ \sum_{k=1}^n f(t_{k-1})(A(t_k) - A(t_{k-1})) \right] = E \left[ \sum_{k=1}^n f(t_{k-1})(B(t_k) - B(t_{k-1})) \right]$$

holds. Here, setting  $f^\delta(s) = f(t_k), t_k < s \leq t_{k+1}; k = 0, 1, \dots, n-1$ , we have

$$E \left[ \int_0^t f^\delta(s) dA(s) \right] = E \left[ \int_0^t f^\delta(s) dB(s) \right].$$

Consequently, by  $|\delta| \rightarrow 0$  and the left-continuity, we obtain

$$E \left[ \int_0^t f(s) dA(s) \right] = E \left[ \int_0^t f(s) dB(s) \right].$$

□

**Lemma 4.7.** (cf.[5]) *Let  $A$  be an integrable increasing process. Then  $A$  is natural if and only if*

$$E[X(t)A(t)] = E \left[ \int_0^t X(s-) dA(s) \right]$$

holds for any bounded martingale  $X$ .

**Theorem 4.8.** *The Doob-Meyer decomposition of a near-submartingale is uniquely determined if it exists.*

*Proof.* Let  $X = \{X(t), t \in \mathbb{R}_+\}$  be a near-submartingale. Suppose that both of  $X = M + A$  and  $X = N + B$  are the Doob-Meyer decompositions. Then since  $A - B$  is a near-martingale and by Lemma 4.6, for any bounded martingale  $\{Y(t); t \in \mathbb{R}_+\}$ , we have

$$E \left[ \int_0^t Y(s-) dA(s) \right] = E \left[ \int_0^t Y(s-) dB(s) \right].$$

Since  $A, B$  is natural increasing and by Lemma 4.7, we have

$$E[Y(t)A(t)] = E[Y(t)B(t)].$$

For any bounded random variable  $Y$ , we define  $\mathbf{Y} = \{Y(t); t \in \mathbb{R}_+\}$  by  $Y(t) := E[Y|\mathcal{F}_t]$  for all  $t \in \mathbb{R}_+$ . Then,  $\mathbf{Y}$  is a  $(\mathcal{F}_t)_{t \in \mathbb{R}_+}$ -martingale, and therefore, we have

$$\begin{aligned} E[YA(t)] &= E[E[YA(t)|\mathcal{F}_t]] = E[Y(t)A(t)] \\ &= E[Y(t)B(t)] = E[E[YB(t)|\mathcal{F}_t]] = E[YB(t)]. \end{aligned}$$

Consequently, putting  $Y = 1_\Lambda$  for all  $\Lambda \in \mathcal{F}$ , we obtain  $P(A(t) = B(t)) = 1$  for each  $t \in \mathbb{R}_+$ . This implies

$$P(\forall t \in \mathbb{R}_+; A(t) = B(t)) = 1$$

by the right-continuity of  $A(t)$  and  $B(t)$ . □

Let  $\mathcal{T}$  be the set of stopping times and set  $\mathcal{T}_a := \{\tau \in \mathcal{T}; \tau(\omega) \leq a, \forall \omega \in \Omega\}$ . A closed near-submartingale  $X = \{X(t), t \in \overline{\mathbb{R}}_+\}$  is called to be in the class  $(D)$  if  $X(\tau)$  is uniformly integrable for any  $\tau \in \mathcal{T}$ . A near-submartingale  $X = \{X(t), t \in \mathbb{R}_+\}$  is called to be in the class  $(DL)$  if  $X(\tau)$  is uniformly integrable for any  $a > 0$  and  $\tau \in \mathcal{T}_a$ .

**Lemma 4.9.** (cf. [5])  $\{A_\infty^n; n \in \mathbb{N}\}$  is uniformly integrable.

**Theorem 4.10.** Let  $X$  be a near-submartingale in the class  $(DL)$ . If  $X(t) \rightarrow X(\infty)$  a. e. and there exists an integrable random variable  $Y$  such that  $|X_t| \leq Y$  for all  $t \geq 0$ , then  $X$  has the Doob-Meyer decomposition  $X = N + A$ . Moreover, if  $X$  is in the class  $(D)$ , then  $N$  and  $A$  in the decomposition of  $X$  are uniformly integrable.

*Proof.* It is enough to prove the theorem in the case of a near-submartingale  $X = \{X(t), t \in \mathbb{R}_+\}$  in the class  $(D)$ . Let  $Y(t)$  be  $Y(t) = X(t) - E[X(\infty)|\mathcal{F}_t]$  for all  $t \in \mathbb{R}_+$ . Then,  $\{Y(t), t \in \mathbb{R}_+\}$  is a near-submartingale, and hence  $\lim_{t \rightarrow \infty} Y(t) = 0$ , a. e. Let  $\{X(t), t \in \mathbb{R}_+\}$  be a near-submartingale satisfying  $\lim_{t \rightarrow \infty} X(t) = 0$ , a. e.

Take a sequence  $\delta_n = \{t_j^{(n)} = \frac{j}{2^n}, j \in \mathbb{N}\}$ ,  $n = 1, 2, 3, \dots$  of partitions of  $[0, \infty)$ . For an arbitrarily fixed  $\delta_n$ , we denote  $t_j^{(n)}$  by  $t_j$  simply. For each  $n$ , we define an increasing process  $A^n(t), t \in \delta_n$  by

$$A^n(t_k) = \sum_{i=1}^{k-1} \{E[X(t_{i+1})|\mathcal{F}_{t_i}] - E[X(t_i)|\mathcal{F}_{t_i}]\}, \quad t_j \in \delta_n.$$

Then by Lemma 4.9,  $A^n(\infty)$  is uniformly integrable. Therefore, there exist some subsequence  $A^{n_\ell}(\infty), \ell = 1, 2, \dots$  and an integrable random variable  $A(\infty)$  such that  $A^{n_\ell}(\infty) \rightarrow A(\infty)$  in  $L^1$ . For any  $t \in \mathbb{R}_+$ , we define  $A(t)$  by

$$A(t) = E[X(t)|\mathcal{F}_t] + E[A(\infty)|\mathcal{F}_t]. \tag{4.1}$$

Then  $A$  is a  $(\mathcal{F}_t)$ -adapted process. Since

$$\begin{aligned} E[A^{n_\ell}(\infty)|\mathcal{F}_0] &= \lim_{k \rightarrow \infty} E \left[ \sum_{j=0}^{k-1} \{E[X(t_{j+1})|\mathcal{F}_{t_j}] - E[X(t_j)|\mathcal{F}_{t_j}]\} \middle| \mathcal{F}_0 \right] \\ &= \lim_{k \rightarrow \infty} \{E[X(t_k)|\mathcal{F}_0] - E[X(0)|\mathcal{F}_0]\} \\ &= -E[X(0)|\mathcal{F}_0], \quad t_k \in \delta_{n_\ell} \end{aligned}$$

for any  $\ell = 1, 2, \dots$ , we have

$$A(0) = E[X(0)|\mathcal{F}_0] + \lim_{\ell \rightarrow \infty} E[A^{n_\ell}(\infty)|\mathcal{F}_0] = 0.$$

We next prove that  $A$  is a natural increasing process. Take  $s$  and  $t$  with  $s < t$  in  $\bigcup_n \delta_n$ . Then since  $s, t \in \delta_{n_\ell}$  for a large  $n_\ell \in \mathbb{N}$ , by Theorem 4.1, we have

$$E[X(s)|\mathcal{F}_s] + E[A^{n_\ell}(\infty)|\mathcal{F}_s] \leq E[X(t)|\mathcal{F}_t] + E[A^{n_\ell}(\infty)|\mathcal{F}_t].$$

Taking  $n_\ell \rightarrow \infty$ , we get

$$E[X(s)|\mathcal{F}_s] + E[A(\infty)|\mathcal{F}_s] \leq E[X(t)|\mathcal{F}_t] + E[A(\infty)|\mathcal{F}_t], \quad \text{a. e.}$$

Hence,  $A(s) \leq A(t)$ . Since  $\bigcup_n \delta_n$  is dense in  $\mathbb{R}_+$ , we obtain  $A(s) \leq A(t)$  for all  $s < t$ . This implies that  $A$  is an increasing process. For any bounded closed martingale  $Z$ , we can see that

$$\begin{aligned} E[Z(\infty)A^n(\infty)] &= \sum_k E[Z(\infty)(A^n(t_{k+1}) - A^n(t_k))] \\ &= \sum_k E[(A^n(t_{k+1}) - A^n(t_k))E[Z(\infty)|\mathcal{F}_{t_k}]] \\ &= \sum_k E[(A^n(t_{k+1}) - A^n(t_k))E[Z(t_k)|\mathcal{F}_{t_k}]] \\ &= \sum_k E[Z(t_k)(A^n(t_{k+1}) - A^n(t_k))], \quad t_k \in \delta_n. \end{aligned}$$

On the other hand, since

$$E[A(t) - A(s)|\mathcal{F}_s] = E[X(t) - X(s)|\mathcal{F}_s]$$

by taking conditional expectations under  $\mathcal{F}_s$  in (4.1), we have

$$\begin{aligned} &E[A(t_{k+1}) - A(t_k)|\mathcal{F}_{t_k}] \\ &= E[X(t_{k+1})|\mathcal{F}_{t_k}] - E[X(t_k)|\mathcal{F}_{t_k}] \\ &= A^n(t_{k+1}) - A^n(t_k). \end{aligned}$$

Therefore, it holds that

$$E[Z(\infty)A^n(\infty)] = \sum_k E[Z(t_k)(A(t_{k+1}) - A(t_k))].$$

Taking  $n \rightarrow \infty$ , we obtain

$$E[Z(\infty)A(\infty)] = E \left[ \int_0^\infty Z(s-) dA(s) \right].$$



This implies that  $A$  is natural. Since

$$\begin{aligned} E[X(t) - A(t)|\mathcal{F}_s] &= E[E[X(t) - A(t)|\mathcal{F}_t]|\mathcal{F}_s] \\ &= E[-E[A(\infty)|\mathcal{F}_t]|\mathcal{F}_s] \\ &= -E[A(\infty)|\mathcal{F}_s] \\ &= E[X(s) - A(s)|\mathcal{F}_s], \end{aligned}$$

the near-martingale part of  $X$  is given by  $X - A$ .  $\square$

### 5. A Stochastic Integral by a Near-martingale

Let  $0 \leq a < b$ . Let  $\mathcal{F}_t := \sigma(B(b) - B(s); t < s \leq b) \vee \mathcal{N}$  for any  $t \in [a, b]$ , and  $C([a, b])$  the Banach space of all continuous functions on  $[a, b]$  with norm  $\|\cdot\|_\infty$  given by  $\|f\|_\infty := \sup_{t \in [a, b]} |f(t)|$ ,  $f \in C([a, b])$ . Define  $\mathcal{B}(C([a, b]))$  by the smallest  $\sigma$ -field including the family of open sets in  $C([a, b])$ , which is called the *topological Borel field*. Denote by  $P_W$  the Wiener measure on  $\mathcal{B}(C([a, b]))$ . For any  $(\mathcal{F}_t)$ -adapted process  $g = \{g(t); a \leq t \leq b\}$  we consider

$$N(t) := \int_t^b g(u)dB(u), \quad t \in [a, b]. \quad (5.1)$$

Then,  $g$  is an instantly independent process of  $(\mathcal{F}_t)$  and  $N = \{N(t); a \leq t \leq b\}$  is a near-martingale and also an instantly independent process of  $(\mathcal{F}_t)$ . Since  $g(t)$  is  $\mathcal{F}_t$ -measurable for any  $t \in [a, b]$ , then  $g(t)$  can be expressed in the form

$$g(t) = G(B(b) - B(s); t < s \leq b)$$

for some  $\mathcal{B}(C([a, b]))$ -measurable function  $G$  for any  $t \in [a, b]$ .

By Theorem 4.10, there exists a unique natural increasing process  $A = \{A(t); a \leq t \leq b\}$  such that  $-N^2 - A$  is a near-martingale. We denote  $A$  by  $\langle N \rangle = \{\langle N \rangle(t); a \leq t \leq b\}$ . Here, we have

$$E[(N(t) - N(s))^2|\mathcal{F}_s] = E[\langle N \rangle(t) - \langle N \rangle(s)|\mathcal{F}_s]$$

for any  $s < t$ . Let

$$\mathcal{L}^2(\langle N \rangle) := \left\{ X; X \text{ is predictable and satisfies } E \left[ \int_a^t |X(t)|^2 d\langle N \rangle(t) \right] < \infty \forall t \right\}.$$

For any  $X$  in  $\mathcal{L}^2(\langle N \rangle)$ , we define semi-norms  $\|X\|_t(\langle N \rangle)$ ,  $a \leq t \leq b$ , by

$$\|X\|_t(\langle N \rangle) := E \left[ \int_a^t |X|^2 d\langle N \rangle(t) \right]^{1/2}.$$

Then  $\mathcal{L}^2(\langle N \rangle)$  is the complete metric space with semi-norms  $\|X\|_t(\langle N \rangle)$ ,  $a \leq t \leq b$ .

For any  $f \in C([a, b])$  and partition  $\Delta : a = t_0 < t_1 < \dots < t_n = b$ , we put

$$f_\Delta = \sum_{k=1}^n f(B(t_{k-1}))1_{[t_{k-1}, t_k)}$$

and define the stochastic integral  $\int_a^b f_\Delta(B(t))dN(t)$  by

$$\int_a^b f_\Delta(B(t))dN(t) := \sum_{k=1}^n f(B(t_{k-1}))(N(t_k) - N(t_{k-1})), \quad \text{in } L^2(\Omega).$$

Then we have the following:

**Proposition 5.1.** *For any  $f \in C([a, b])$  and partition*

$$\Delta : a = t_0 < t_1 < \cdots < t_n = b,$$

*the process  $\int_a^\cdot f_\Delta dN$  is an  $L^2$  near-martingale and satisfies*

$$\left\langle \int_a^\cdot f_\Delta(B(\cdot)) dN \right\rangle (t) = \int_a^t f_\Delta(B(t)) d\langle N \rangle (t), \quad (5.2)$$

$$E \left[ \left| \int_a^t f_\Delta(B(t)) dN(t) \right|^2 \right] = \|f_\Delta(B(\cdot))\|_t^2(\langle N \rangle)^2 \quad (5.3)$$

for all  $a \leq t \leq b$ .

*Proof.* Let  $t > s > a$  and  $f \in C([a, b])$ . Then for any partition

$$\Delta : s = t_0 < t_1 < \cdots < t_n = b,$$

we can see that

$$\begin{aligned} & E \left[ \left( \int_s^t f_\Delta(B(t)) dN(t) \right)^2 \middle| \mathcal{F}_s \right] \\ &= \sum_{k=1}^n E[E[f_{k-1}^2 (\Delta_k N(t))^2 | \mathcal{F}_{t_{k-1}}] | \mathcal{F}_s] \\ &\quad + 2 \sum_{k>\ell} E[E[f_{k-1} f_{\ell-1} \Delta_k N(t) \Delta_\ell N(t) | \mathcal{F}_{t_{\ell-1}}] | \mathcal{F}_s] \\ &= \sum_{k=1}^n E[f_{k-1}^2 E[(\Delta_k N(t))^2 | \mathcal{F}_{t_{k-1}}] | \mathcal{F}_s] \\ &\quad + 2 \sum_{k>\ell} E[E[f_{k-1} f_{\ell-1} E[(\Delta_k N(t))(\Delta_\ell N(t)) | \mathcal{F}_{t_{k-1}}] | \mathcal{F}_{t_{\ell-1}}] | \mathcal{F}_s], \end{aligned}$$

where  $f_{k-1} := f(B(t_{k-1}))$ , and  $\Delta_k N(t) := N(t_k) - N(t_{k-1})$  for  $k = 1, 2, \dots, n$ . By Corollary 2.5 and Theorem 2.6, we have

$$E[\Delta_k N(t) \Delta_\ell N(t) | \mathcal{F}_{t_{k-1}}] = 0.$$

Therefore, we get

$$\begin{aligned} & E \left[ \left( \int_s^t f_\Delta(B(u)) dN(u) \right)^2 \middle| \mathcal{F}_s \right] \\ &= \sum_{k=1}^n E[f(B(t_{k-1}))^2 E[\langle N \rangle(t_k) - \langle N \rangle(t_{k-1}) | \mathcal{F}_{t_{k-1}}] | \mathcal{F}_s] \\ &= E \left[ \sum_{k=1}^n f(B(t_{k-1}))^2 (\langle N \rangle(t_k) - \langle N \rangle(t_{k-1})) \middle| \mathcal{F}_s \right] \\ &= E \left[ \int_s^t f_\Delta(B(u))^2 d\langle N \rangle(u) \middle| \mathcal{F}_s \right]. \end{aligned}$$

This implies (5.2), and taking the expectation of the both sides of (5.2), we obtain (5.3).  $\square$

For any  $f \in C([a, b])$ , we have  $f_{\Delta}(B(t)) \rightarrow f(B(t))$  in  $\mathcal{L}^2(\langle N \rangle)$  as  $|\Delta| := \max\{t_k - t_{k-1}; k = 1, 2, \dots, n\} \rightarrow 0$ . Therefore by Proposition 5.1, we can define  $\int_a^b f(B(t))dN(t)$  by

$$\int_a^b f(B(t))dN(t) := \lim_{|\Delta| \rightarrow 0} \int_a^b f_{\Delta}(B(t))dN(t) \quad \text{in } L^2(\Omega).$$

The stochastic integral  $\int_a^b f(B(t))g(t)dB(t)$  with  $g(t)$  from (5.1) can be regarded as  $-\int_a^b f(B(t))dN(t)$ . This is a generalization of [10] and a formulation of the new integral in [1] from the point of view of the stochastic integral by the near-martingale.

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