## Journal of Stochastic Analysis

Volume 2 | Number 1

Article 6

March 2021

# Linear Decomposition and Anticipating Integral for Certain Random Variables

Ching-Tang Wu National Taitung University, No 369, University Road, Sec. 2, Taitung, Taiwan, ctwu@nttu.edu.tw

Ju-Yi Yen University of Cincinnati, Cincinnati, Ohio 45221-0025, US, ju-yi.yen@uc.edu

Follow this and additional works at: https://digitalcommons.lsu.edu/josa

Part of the Analysis Commons, and the Other Mathematics Commons

### **Recommended Citation**

Wu, Ching-Tang and Yen, Ju-Yi (2021) "Linear Decomposition and Anticipating Integral for Certain Random Variables," *Journal of Stochastic Analysis*: Vol. 2 : No. 1 , Article 6. DOI: 10.31390/josa.2.1.06 Available at: https://digitalcommons.lsu.edu/josa/vol2/iss1/6



## LINEAR DECOMPOSITION AND ANTICIPATING INTEGRAL FOR CERTAIN RANDOM VARIABLES

CHING-TANG WU AND JU-YI YEN\*

ABSTRACT. In this paper, we construct an anticipating stochastic integral by linearly decompose a class of non  $\mathcal{F}_t$ -measurable random variables. The result is applied to the derivation of the Itô formula.

#### 1. Introduction

Throughout the paper,  $B_t$  denotes the one-dimensional standard Brownian motion and  $\mathcal{F}_t$  is the  $\sigma$ -field generated by  $\{B_s; 0 \leq s \leq t\}$ .

The Itô stochastic integral:

$$\int_0^\infty f(t)dB_t \tag{1.1}$$

is established for any  $\mathcal{F}_t$ -adapted function f(t) such that  $\int_0^\infty |f(t)|^2 dt < \infty$ . A well-known example to extend the scope of the Itô integral was given in 1976 by Itô as follows:

$$\int_0^t B_1 dB_s, \quad 0 \le t \le 1. \tag{1.2}$$

Here, for  $0 \leq t < 1$ ,  $B_1$  is not adapted to  $\mathcal{F}_t$ . The concept of the *enlargement* of filtration was then introduced to solve this problem; the basic idea is to build a  $\sigma$ -field  $\mathcal{G}_t$  generated by  $\mathcal{F}_t$  and  $B_1$ , in a way that the Brownian motion  $B_t$  is a semimartingale with respect to  $\mathcal{G}_t$ . Thus, (1.2) can be defined in the sense of the original Itô stochastic integral. However, the enlargement of filtration cannot be easily generalized, since for different enlarged filtrations,  $B_t$  may no longer be a semimartingale. See e.g. [4, 5, 6] for detailed studies in this direction.

About a decade ago, an alternate technique was introduced by Ayed and Kuo [1, 2] in order to solve stochastic integrals with non-adapted integrands such as (1.2). Specifically, Ayed and Kuo decompose the integrand into two components, i.e., the adapted parts and the anticipating part. In this work, we follow the method developed in [1, 2] to define the anticipating stochastic integral

$$(G)\int_0^t f(L)dB_s,\tag{1.3}$$

Received 2020-7-2; Accepted 2021-2-12; Communicated by the editors.

<sup>2010</sup> Mathematics Subject Classification. Primary 60G20, 60H05; Secondary 60G48.

Key words and phrases. Brownian motion, adapted stochastic process, Itô Integral, instantly independent stochastic process, anticipating integral.

<sup>\*</sup> Corresponding author.

where L is not necessarily  $\mathcal{F}_t$ -measurable, and f is a  $C^2$ -function with bounded second derivative.

This paper is organized as follows. In Section 2, we first introduce some necessary notions for our analysis. We then show that if the integrand in (1.3) can be decomposed into two parts as in [1, 2] (namely, the adapted parts and the anticipating part), then such decomposition is unique with certain properties (to be specified in Section 2). The stochastic integral (1.3) is properly defined in Section 3. In particular, we use the left-end point to compute the adapted part, and the right-end point for the anticipating part. In Section 4, we give the Itô formula for (1.3) to complete our study.

#### 2. Decomposition of L

Before we define the stochastic integral (1.3), we shall first introduce two terminologies: *instant independence* and *near-martingale*. Ayed and Kuo [1] introduced the concept of instantly independent for certain anticipating stochastic processes. We recall the definition of *instant independence*.

**Definition 2.1.** A stochastic process  $(Y_t)_{a \leq t \leq b}$ , is said to be *instantly independent* with respect to the filtration  $(\mathcal{F}_t)$ , if  $\sigma(Y_t)$  and  $\mathcal{F}_t$  are independent for each t.

Moreover, in [8], the notion of *near-martingale* is defined:

**Definition 2.2.** Let  $\mathbb{E}|X_t| < \infty$  for all t. We say that a stochastic process  $(X_t)$  is a *near-martingale* with respect to the filtration  $(\mathcal{F}_t)$  if

$$\mathbb{E}[X_t - X_s | \mathcal{F}_s] = 0, \quad \forall s \le t.$$

Now, we decompose a random variable  $\overline{L}$  as

$$L = C_t + I_t,$$

where  $C_t$  takes the class of instantly independent stochastic process (called the counterpart), and  $I_t$  is an  $(\mathcal{F}_t)$ -adapted process (called the Itô part). Lemma 2.3 below shows the uniqueness of such decomposition if such decomposition exists.

**Lemma 2.3.** Given a filtration  $(\mathcal{F}_t)$ , if a random variable  $\overline{L}$  can be decomposed into

$$\bar{L} = C_t + I_t \tag{2.1}$$

for all t, where  $(C_t)$  takes the class of instantly independent stochastic process with  $\mathbb{E}[C_t] = 0$  for all t; and  $(I_t)$  is an  $(\mathcal{F}_t)$ -adapted process, then the linear decomposition in (2.1) is unique. Moreover,  $(I_t)$  is a martingale, and  $(C_t)$  is a near-martingale with respect to  $(\mathcal{F}_t)$ .

*Proof.* We first show the uniqueness of the decomposition. Suppose  $\overline{L}$  has two decompositions:

$$\bar{L} = C_t + I_t = D_t + J_t,$$

where  $C_t$  and  $D_t$  denote the counterpart, and  $I_t$  and  $J_t$  denote the Itô part of the random variable  $\bar{L}$ , respectively. Then

$$(C_t - D_t) + (I_t - J_t) = 0.$$

Taking the conditional expectation with respect to  $\mathcal{F}_t$  on both sides, we get

$$\mathbb{E}[(C_t - D_t)|\mathcal{F}_t] + \mathbb{E}[(I_t - J_t)|\mathcal{F}_t] = 0.$$

Since  $I_t - J_t$  are  $\mathcal{F}_t$ -measurable, we have

$$\mathbb{E}[(C_t - D_t)|\mathcal{F}_t] + (I_t - J_t) = 0.$$

Moreover, since  $C_t$  and  $D_t$  are instantly independent with respect to  $\mathcal{F}_t$  and  $\mathbb{E}[C_t] = \mathbb{E}[D_t] = 0$ , we get

$$\mathbb{E}[(C_t - D_t)|\mathcal{F}_t] = \mathbb{E}[C_t - D_t] = \mathbb{E}[C_t] - \mathbb{E}[D_t] = 0$$

Thus,  $(I_t - J_t) = 0$ . This implies that  $I_t = J_t$  and  $C_t = D_t$  for all t.

Furthermore, since  $C_t$  is instantly independent with respect to  $(\mathcal{F}_t)$ ,  $C_t$  is independent of  $\mathcal{F}_s$  for all  $s \leq t$ . Thus,

$$\mathbb{E}[C_t - C_s | \mathcal{F}_s] = \mathbb{E}[C_t - C_s] = 0.$$

We have  $(C_t)$  is a near-martingale. Moreover, due to  $I_t - I_s = C_s - C_t$  for s < t,

$$\mathbb{E}[I_t - I_s | \mathcal{F}_s] = \mathbb{E}[C_s - C_t | \mathcal{F}_s] = \mathbb{E}[C_s - C_t] = 0, \qquad s < t.$$

Since  $(I_t)$  is adapted, we get clearly that  $\mathbb{E}[I_t|\mathcal{F}_s] = I_s$ , i.e.,  $(I_t)$  is a martingale.  $\Box$ 

**Example 2.4.** Clearly, the decomposition in (2.1) exists if  $(\bar{L}, B_s)_{0 \le s \le t}$  is Gaussian. In particular, we can write:

$$C_t = \bar{L} - \mathbb{E}[\bar{L}|\mathcal{F}_t]$$
$$I_t = \mathbb{E}[\bar{L}|\mathcal{F}_t].$$

Under this assumption,  $C_t$  and  $\mathcal{F}_t$  are independent since  $\mathbb{E}[C_t B_s] = 0$  for all  $s \leq t$ .

- (a) Let  $\overline{L} = B_1$ , then  $C_t = B_1 B_t$ , and  $I_t = B_t$ , for  $0 \le t \le 1$ .
- (b) Suppose  $\bar{L} = \int_0^1 \zeta(u) dB_u + \int_0^1 \eta(u) du$  for some deterministic functions  $\zeta$  and  $\eta$ , then

$$C_t = \int_t^1 \zeta(u) dB_u, \quad \text{and} \quad I_t = \int_0^t \zeta(u) dB_u + \int_0^1 \eta(u) du,$$
  
for  $0 \le t \le 1$ .

However, if  $(\bar{L}, B_s)_{0 \le s \le t}$  is not Gaussian, then the existence of  $(C_t)$  and  $(I_t)$  is not guaranteed. For example, when  $\bar{L} = B_1^2$ ,  $C_t = \bar{L} - \mathbb{E}[\bar{L}|\mathcal{F}_t]$  and  $B_s$  are not independent for  $s \le t$ , this implies that  $C_t$  is not instantly independent with respect to  $\mathcal{F}_t$ . Although, in this case,  $\bar{L}$  cannot be decomposed into  $C_t + I_t$ ,  $\bar{L}$  can be written as

$$\bar{L} = B_1^2 = ((B_1 - B_t) + B_t)^2 = (C_t + I_t)^2.$$

Throughout this paper, we investigate the case where  $\overline{L}$  can be written as f(L), where  $(L, B_s)_{0 \le s \le t}$  is Gaussian.

#### 3. The Extension of Itô Integral

We would like to study the properties of the stochastic integral defined in (1.3) under the construction of (2.1). Different from the definition of the Itô integral, we define the anticipating integral  $(G) \int_0^t f(L) dB_s$  as follows:

$$(G) \int_{0}^{t} f(L) dB_{s} = (G) \int_{0}^{t} f(C_{s} + I_{s}) dB_{s}$$
$$= \lim_{\|\Delta\| \to 0} \sum_{i=1}^{n} f(C_{t_{i}} + I_{t_{i-1}}) (B_{t_{i}} - B_{t_{i-1}}), \qquad (3.1)$$

where  $\Delta = \{t_0, t_1, ..., t_n\}$  is a partition of [0, t]. Notice that for the counter part, we take the left-end point; whereas for the Itô part, we take the right-end point as the usual Itô integral. Our goal is to analyze if the above limit exists.

In the sequel, we assume that  $I_t$  is a continuous process.

**Proposition 3.1.** Suppose f is a  $C^2$ -function with bounded second derivative and the cross-variation of I and B, [I, B], exists, and is defined as

$$[I,B]_t = \lim_{\|\Delta\| \to 0} \sum_{i=1}^n (B_{t_i} - B_{t_{i-1}}) (I_{t_i} - I_{t_{i-1}}).$$

Then the limit

$$\lim_{\|\Delta\| \to 0} \sum_{i=1}^{n} f(C_{t_i} + I_{t_{i-1}}) (B_{t_i} - B_{t_{i-1}})$$

converges in probability. Moreover, this new integral can be written as

$$(G)\int_0^t f(L)dB_s = f(L)B_t - f'(L)[I,B]_t.$$
(3.2)

*Proof.* Due to (2.1), we have  $C_{t_{i-1}} + I_{t_{i-1}} = L$ . Consider

$$\sum_{i=1}^{n} f(C_{t_i} + I_{t_{i-1}})(B_{t_i} - B_{t_{i-1}})$$
  
= 
$$\sum_{i=1}^{n} \left( f(C_{t_i} + I_{t_{i-1}}) - f(C_{t_{i-1}} + I_{t_{i-1}}) \right) (B_{t_i} - B_{t_{i-1}})$$
  
+ 
$$\sum_{i=1}^{n} f(L)(B_{t_i} - B_{t_{i-1}})$$

Clearly,

$$\sum_{i=1}^{n} f(L)(B_{t_i} - B_{t_{i-1}}) = f(L)B_t.$$

By Taylor expansion, we get

$$\sum_{i=1}^{n} (f(C_{t_i} + I_{t_{i-1}}) - f(C_{t_{i-1}} + I_{t_{i-1}}))(B_{t_i} - B_{t_{i-1}})$$
  
= 
$$\sum_{i=1}^{n} \left( f'(C_{t_{i-1}} + I_{t_{i-1}})(C_{t_i} - C_{t_{i-1}}) + \frac{1}{2} f''(\xi_i + I_{t_{i-1}})(C_{t_i} - C_{t_{i-1}})^2 \right) (B_{t_i} - B_{t_{i-1}})$$

for some  $\xi_i$  between  $C_{t_{i-1}}$  and  $C_{t_i}.$  The first order term of the above equation is equal to

$$\sum_{i=1}^{n} f'(L)(I_{t_{i-1}} - I_{t_i})(B_{t_i} - B_{t_{i-1}}),$$

which tends to  $-f'(L)[I, B]_t$  in probability as  $\|\Delta\| \to 0$ . Since f'' is bounded by a constant, say  $\alpha$ , we have

$$\left|\sum_{i=1}^{n} \frac{1}{2} f''(\xi_{i} + I_{t_{i-1}})(C_{t_{i}} - C_{t_{i-1}})^{2}(B_{t_{i}} - B_{t_{i-1}})\right|^{2}$$

$$\leq \frac{1}{2} \alpha^{2} \sum_{i=1}^{n} |C_{t_{i}} - C_{t_{i-1}}|^{4} |B_{t_{i}} - B_{t_{i-1}}|^{2}$$

$$= \frac{1}{2} \alpha^{2} \sum_{i=1}^{n} |I_{t_{i}} - I_{t_{i-1}}|^{4} |B_{t_{i}} - B_{t_{i-1}}|^{2}$$

$$\leq \frac{1}{2} \alpha^{2} \max_{j} |I_{t_{j}} - I_{t_{j-1}}|^{4} \sum_{i=1}^{n} |B_{t_{i}} - B_{t_{i-1}}|^{2}$$

which tends to 0 as  $\|\Delta\| \to 0$ .

Remark 3.2. The definition of (3.1) is well-defined. Suppose that f(L) can be also written as  $g(\tilde{L})$ , where both  $(L, B_s)_{0 \le s \le t}$  and  $(\tilde{L}, B_s)_{0 \le s \le t}$  are Gaussian. Without loss of generality, we may assume that  $L, \tilde{L} \sim \mathcal{N}(0, 1)$ . Since  $f(L) = g(\tilde{L})$ , the only possibilities are that  $\tilde{L} = \pm L$ .

Assume that  $\tilde{L} = -L$ , then f(x) = g(-x). Suppose that  $\tilde{L}$  can be decomposed to  $\tilde{L} = -C_t - \tilde{I}_t$  for all t, then

$$\sum f(C_{t_i} + I_{t_{i-1}})(B_{t_i} - B_{t_{i-1}}) - \sum g(-C_{t_i} - I_{t_{i-1}})(B_{t_i} - B_{t_{i-1}})$$
  
= 
$$\sum \left( f(C_{t_i} + I_{i_{i-1}}) - f(C_{t_{i-1}} + I_{i_{i-1}}) \right) (B_{t_i} - B_{t_{i-1}})$$
  
$$- \sum \left( g(-C_{t_i} - I_{t_{i-1}}) - g(-C_{t_{i-1}} - I_{t_{i-1}}) \right) (B_{t_i} - B_{t_{i-1}}),$$

which approaches to  $f'(L)[I, B]_t + g'(-L)[I, B]_t$  as  $||\Delta||$  goes to 0. Since f'(x) = -g'(-x), which results (G)  $\int_0^t f(L) dB_u = (G) \int_0^t g(\tilde{L}) dB_u$ .

Remark 3.3. In [1], the authors consider the stochastic integral of the form

$$\int h(t)\varphi(t)dB_t,\tag{3.3}$$

where h(t) is an adapted process, and  $\varphi(t)$  is an instantly independent process. In this paper, we consider the stochastic integral of the form

$$\int f(C_t + I_t) dB_t. \tag{3.4}$$

Notice that if  $f(x + y) = \sum_{n} h_n(x)\varphi_n(x)$ , for example, when f is a polynomial, sine function, cosine function, or exponential function, then the two integrals (3.3) and (3.4) obtain the same results.

*Remark* 3.4. Using a similar argument as in the proof of Proposition 3.1, we get the following generalization.

(a) Suppose that  $f : \mathbb{R}^2 \longrightarrow \mathbb{R}$  is twice differentiable with bounded second partial derivatives and  $(L^1, L^2, B_s)_{s \leq t}$  is Gaussian. Assume that  $I^1$  and  $I^2$  are the Itô part of  $L^1$  and  $L^2$ , respectively, and the cross-variation  $[I^1, B]$  and  $[I^2, B]$  exist. Then

$$(G)\int_0^t f(L^1, L^2)dB_s = f(L^1, L^2)B_t - \frac{\partial}{\partial x}f(L^1, L^2)[I^1, B]_t - \frac{\partial}{\partial y}f(L^1, L^2)[I^2, B]_t$$

(b) Suppose  $f : [0,T] \times \mathbb{R}^2 \longrightarrow \mathbb{R}$  is a  $C^{1,2,2}$  function and the Itô integral  $\int_0^t f(s,L,B_s) dB_s$  exists in the sense of enlargement of filtration<sup>1</sup>, then

$$(G)\int_0^t f(s,L,B_s)dB_s = \int_0^t f(s,L,B_s)dB_s - \int_0^t \frac{\partial}{\partial x} f(s,L,B_s)d[I,B]_s.$$

In particular, if  $f(t, x, y) = f_1(x)f_2(y)$ , then

$$(G)\int_0^t f_1(L)f_2(B_s)dB_s = f_1(L)\int_0^t f_2(B_s)dB_s - f_1'(L)\int_0^t f_2(B_s)d[I,B]_s.$$

(c) The above discussion holds also for general semimartingale X.

**Example 3.5.** (a) Let  $L = B_1$ , f(x) = x, then

$$C_t = \begin{cases} B_1 - B_t, & 0 \le t \le 1, \\ 0, & t > 1, \end{cases}$$

and

$$I_t = \begin{cases} B_t, & 0 \le t \le 1, \\ B_1, & t > 1, \end{cases}$$

Thus,

$$(G) \int_0^t B_1 dB_s = \begin{cases} B_1 B_t - 1 \cdot [B, B]_t = B_1 B_t - t, & 0 \le t < 1, \\ B_1 B_t - 1 \cdot [B, B]_1 = B_1 B_t - 1, & t \ge 1. \end{cases}$$

(b) Let 
$$L = B_1$$
,  $f(x) = x^2$ , then

$$(G) \int_0^t B_1^2 dB_s = \begin{cases} B_1^2 B_t - 2tB_1, & 0 \le t < 1, \\ B_1^2 B_t - 2B_1, & t \ge 1. \end{cases}$$

<sup>&</sup>lt;sup>1</sup>The condition on L such that the integral exists is shown in [6].

(c) Let  $L = B_1$ ,  $f(x) = e^x$ , then for  $0 \le t \le 1$ ,  $(G) \int_0^t e^{B_1} dB_s = e^{B_1} (B_t - t).$ 

(d) Let  $L = B_1$ ,  $f(x) = \tan^{-1}(x)$ , then for  $0 \le t \le 1$ ,

(G) 
$$\int_0^t \tan^{-1}(B_1) dB_u = B_t \tan^{-1}(B_1) - \frac{t}{1+B_1^2}.$$
  
(e) Let  $L = B_1, f(t, x, y) = \frac{y}{1+x^2}$ , then for  $0 \le t \le 1$ ,

$$G\int_{0}^{t} \frac{B_{u}}{1+B_{1}^{2}} dB_{u} = \frac{1}{1+B_{1}^{2}} \int_{0}^{t} B_{u} dB_{u} + \frac{2B_{1}}{(1+B_{1}^{2})^{2}} \int_{0}^{t} B_{u} du$$
$$= \frac{B_{t}^{2}-t}{2(1+B_{1}^{2})} + \frac{2B_{1}}{(1+B_{1}^{2})^{2}} \int_{0}^{t} B_{u} du$$

(f) Let 
$$L = \int_0^1 B_u du = B_1 - \int_0^1 u dB_u$$
,  $f(x) = x$ , then  
 $I_s = \mathbb{E}[L|\mathcal{F}_s] = \mathbb{E}[B_1 - \int_0^1 u dB_u|\mathcal{F}_s] = B_s - \int_0^s u dB_u$ ,

Thus

$$\int_{0}^{1} \left( \int_{0}^{1} B_{u} du \right) dB_{s} = \int_{0}^{1} L dB_{u}$$
  
=  $LB_{1} - [I, B]_{1}$   
=  $B_{1} \int_{0}^{1} B_{u} du - \frac{1}{2},$ 

since for 
$$t < 1$$
,  $dI_t = (1-t)dB_t$ ,  $[I, B]_t = \int_0^t (1-u)du = t - \frac{1}{2}t^2$ 

**Proposition 3.6.** Let  $f : [0,T] \times \mathbb{R}^2 \longrightarrow \mathbb{R}$  be a  $C^{1,2,2}$ -function with bounded second partial derivatives, then for  $0 \le t \le T$ ,

$$M_t = (G) \int_0^t f(u, L, B_u) dB_u$$

is a near-martingale.

*Proof.* Let  $\Delta = \{s = t_0 < t_1 < \dots < t_n = t\}$  be a partition of [s, t]. Then

$$\begin{split} & \mathbb{E}[M_{t} - M_{s}|\mathcal{F}_{s}] \\ &= \mathbb{E}\left[ (G) \int_{s}^{t} f(u, L, B_{u}) dB_{u} \middle| \mathcal{F}_{s} \right] \\ &= \lim_{\|\Delta\| \to 0} \mathbb{E}\left[ \sum_{i=1}^{n} f(t_{i-1}, C_{t_{i}} + L_{t_{i-1}}, B_{t_{i-1}}) (B_{t_{i}} - B_{t_{i-1}}) \middle| \mathcal{F}_{s} \right] \\ &= \lim_{\|\Delta\| \to 0} \sum_{i=1}^{n} \mathbb{E}\left[ \mathbb{E}\left[ f(t_{i-1}, C_{t_{i}} + L_{t_{i-1}}, B_{t_{i-1}}) (B_{t_{i}} - B_{t_{i-1}}) \middle| \mathcal{F}_{t_{i-1}} \right] \middle| \mathcal{F}_{s} \right] \\ &= \lim_{\|\Delta\| \to 0} \sum_{i=1}^{n} \mathbb{E}\left[ \mathbb{E}\left[ f(t_{i-1}, C_{t_{i}} + L_{t_{i-1}}, B_{t_{i-1}}) \middle| \mathcal{F}_{t_{i-1}} \right] \mathbb{E}[B_{t_{i}} - B_{t_{i-1}}] \middle| \mathcal{F}_{s} \right] \end{split}$$

since  $B_{t_i} - B_{t_{i-1}}$  is independent of  $\mathcal{F}_{t_{i-1}}$  and  $C_{t_i}$ ,  $L_{t_{i-1}}$  and  $B_{t_{i-1}}$  are  $\mathcal{F}_{t_{i-1}}$ -measurable. Due to  $\mathbb{E}[B_{t_i} - B_{t_{i-1}}] = 0$ , we get  $(M_t)$  is a near-martingale.  $\Box$ 

#### 4. Itô Formula

In [3, 7], the authors discuss the Itô formula of the form (3.3); here, we derive the Itô formula in the sense of (3.1).

**Theorem 4.1.** Suppose  $f \in C^{1,2,2}$ . Then, the Itô formula for the anticipating integral defined in (3.1) is:

$$f(T, L, B_T) = f(0, L, 0) + \int_0^T \frac{\partial}{\partial t} f(u, L, B_u) du + (G) \int_0^T \frac{\partial}{\partial y} f(u, L, B_u) dB_u$$
$$\frac{1}{2} \int_0^T \frac{\partial^2}{\partial y^2} f(u, L, B_u) du + \int_0^T \frac{\partial^2}{\partial x \partial y} f(u, L, B_u) d[I, B]_u;$$

and the Itô formula for the anticipating integral defined in the sense of the enlargement filtration is:

$$f(T,L,B_T) - f(0,L,0) = \int_0^T \frac{\partial}{\partial t} f(u,L,B_u) du + \int_0^T \frac{\partial}{\partial y} f(u,L,B_u) dB_u + \frac{1}{2} \int_0^T \frac{\partial^2}{\partial y^2} f(u,L,B_u) du.$$

*Proof.* Suppose that  $0 = t_0 < t_1 < t_2 < \cdots < t_n = T$ . Then

$$f(T, L, B_T) - f(0, L, 0) = \sum_{i=1}^{n} (f(t_i, L, B_{t_i}) - f(t_{i-1}, L, B_{t_{i-1}})).$$

We have

$$f(t_i, L, B_{t_i}) - f(t_{i-1}, L, B_{t_{i-1}}) = (f(t_i, L, B_{t_i}) - f(t_{i-1}, L, B_{t_i})) + (f(t_{i-1}, L, B_{t_i}) - f(t_{i-1}, L, B_{t_{i-1}}))$$

(i) By Taylor formula, there exists  $t_i^*$  between  $t_{i-1}$  and  $t_i$ , such that

$$f(t_i, L, B_{t_i}) - f(t_{i-1}, L, B_{t_i}) = \frac{\partial}{\partial t} f(t_i^*, L, B_{t_i})(t_i - t_{i-1}).$$

Since f is continuously differentiable in t,

$$\lim_{\|\Delta\|\to 0} \sum_{i=1}^{n} (f(t_i, L, B_{t_i}) - f(t_{i-1}, L, B_{t_i}))$$
$$= \lim_{\|\Delta\|\to 0} \sum_{i=1}^{n} \frac{\partial}{\partial t} f(t_i, L, B_{t_i})(t_i - t_{i-1})$$
$$= \int_0^T \frac{\partial}{\partial t} f(u, L, B_u) du.$$

(ii) There exists  $B_{t_i}^*$  between  $B_{t_{i-1}}$  and  $B_{t_i}$  such that

$$f(t_{i-1}, L, B_{t_i}) - f(t_{i-1}, L, B_{t_{i-1}})$$
  
=  $\frac{\partial}{\partial y} f(t_{i-1}, L, B_{t_{i-1}}) (B_{t_i} - B_{t_{i-1}})$   
+  $\frac{1}{2} \frac{\partial^2}{\partial y^2} f(t_{i-1}, L, B_{t_i^*}) (B_{t_i} - B_{t_{i-1}})^2$ 

Summing across *i* from i = 1 to *n*, the last equality tends to

$$\int_0^t \frac{\partial}{\partial y} f(u, L, B_u) dB_u + \frac{1}{2} \int_0^t \frac{\partial^2}{\partial y^2} f(u, L, B_u) du$$

By Remark 3.4,

$$\int_0^t \frac{\partial}{\partial y} f(u, L, B_u) dB_u = (G) \int_0^t \frac{\partial}{\partial y} f(u, L, B_u) dB_u + \int_0^t \frac{\partial^2}{\partial x \partial y} f(u, L, B_u) d[I, B]_u.$$
  
Hence, from (i) and (ii), the result is obtained.

Hence, from (i) and (ii), the result is obtained.

Acknowledgments. C.-T. Wu is greatly indebted to the Ministry of Science and Technology, Taiwan for the research grant (MOST 108-2115-M-143-001-). J.-Y. Yen is grateful to the Academia Sinica, Institute of Mathematics (Taipei, Taiwan) for their hospitality and support during some extended visits.

#### References

- 1. Ayed, Wided; and Kuo, Hui-Hsiung: An extension of the Itô integral, Communications on Stochastic Analysis 2 (2008), no. 3, 323-333.
- 2. Ayed, Wided; Kuo, Hui-Hsiung: An extension of the Itô integral: toward a general theory of stochastic integration, Theory of Stochastic Processes 16 (2010), no. 1, 17–28.
- 3. Hwang, Chii-Ruey; Kuo, Hui-Hsiung; Saitô, Kimiaki; Zhai, Jiayu: A general Itô formula for adapted and instantly independent stochastic processes, Communications on Stochastic Analysis 10 (2016), no. 3, 341-362.
- 4. Jeulin, Thierry: Grossissement d'une filtration et applications, in: Séminaire de Probabilités XIII 721 (1979) 574–609, Springer, Berlin, Heidelberg.
- 5. Jeulin, Thierry; Yor, Marc: Grossissement dune filtration et semi-martingales: formules explicites, in: Séminaire de Probabilités XII 649 (1978) 78-97, Springer, Berlin, Heidelberg.
- 6. Jeulin, Thierry; Yor Marc (editors): Grossissements de filtrations: exemples et applications, Séminaire de Calcul Stochastique 1982/83 Université Paris VI, Vol. 1118, Springer, 1985.
- 7. Kuo, Hui-Hsiung; Sae-Tang, Anuwat; Szozda, Benedykt: The Itô formula for a new stochastic integral, Communications on Stochastic Analysis 6 (2012), no. 4, 603-614.
- 8. Kuo, Hui-Hsiung; Sae-Tang Anuwat; Szozda, Benedykt: A stochastic integral for adapted and instantly independent stochastic processes, in: Stochastic processes, finance and control 1 (2012) 53-71, Advances in Statistics, Probability and Actuarial Science, World Sci. Publ., Hackensack, NJ.

CHING-TANG WU: DEPARTMENT OF MATHEMATICS, NATIONAL TAITUNG UNIVERSITY, NO 369, UNIVERSITY ROAD, SEC. 2, TAITUNG, TAIWAN

E-mail address: ctwu@nttu.edu.tw

JU-YI YEN: DEPARTMENT OF MATHEMATICAL SCIENCES, UNIVERSITY OF CINCINNATI, CINCIN-NATI, OHIO 45221-0025, USA

E-mail address: ju-yi.yen@uc.edu