

CONSTRUCTIVE STOCHASTIC INTEGRAL REPRESENTATION OF SOME
PATH-DEPENDENT BROWNIAN FUNCTIONAL *

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Abstract. Here we consider some path-dependent Brownian functionals and derive constructive formulas for the stochastic integral representation.

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1 Introduction. The question of representing Brownian functionals as a stochastic Ito integral with an explicit form of the integrand is investigated. In general, the representation of one martingale as a stochastic integral with respect to another martingale, when the martingale is adapted with respect to the natural flow of the σ -algebras of the last martingale, is called a martingale representation.

On the other hand, it is known that the Ito stochastic integral as a process under certain conditions on the integrand is a martingale with respect to the natural flow of σ -algebras of Brownian motion. The question naturally arises whether the opposite is true: can any martingale with respect to the natural flow of σ -algebras of Brownian motion be represented as a stochastic Ito integral? The well-known theorem of Clark (1970) gives a positive answer to this question (which is a particular case of solving the martingale representation problem), although an implicit answer can also be found in Ito (1951).

Formally, the task looks like this: assume that $(X_t, \mathfrak{F}_t^X := \sigma\{X_s : 0 \leq s \leq t\}), t \in [0, T]$ and $(M_t, \mathfrak{F}_t^X), t \in [0, T]$ martingales are given. The question is, can M_t be represented as a stochastic integral with respect to martingale X_t ? That is, is there a process $\varphi_t, t \in [0, T]$ adapted to the flow of σ -algebras \mathfrak{F}_t^X such that $M_t = \int_0^t \varphi_s dX_s$? It turned out that we have a positive answer to this question (Clark, 1970) when X_t is a standard Wiener process¹. But in general this is not so.

Consider two independent standard Wiener processes $w = (w_t, t \geq 0)$ and $v = (v_t, t \geq 0)$ (Kallianpur, 1987). Suppose that $X_t = \int_0^t w_s dv_s$ and $\mathfrak{F}_t^X := \sigma\{X_s : 0 \leq s \leq t\}$. Then $\langle X \rangle_t$ is \mathfrak{F}_t^X -measurable, and since $\langle X \rangle_t = \int_0^t w_s^2 ds$, from here it follows that w_t^2 is also \mathfrak{F}_t^X -measurable. Hence, process $M = (M_t, t \geq 0)$, where $M_t = w_t^2 - t$, is a square integrable martingale adapted with filtration $(\mathfrak{F}_t^X, t \geq 0)$, but it cannot be represented as an integral with respect to the martingale X_t .

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¹The Wiener process and the Brownian motion are the same.

It should be noted that finding the explicit expression for φ_t is a very difficult problem. In this direction, one general result is known, called Clark-Ocone formula (1984), according to which $\varphi_t = E[D_t F | \mathfrak{S}_t]$, where D_t is the so called Malliavin stochastic derivative. Glonti and Purtukhia (2017) generalized the Clark-Ocone formula to the case when the functional is not stochastically smooth, but its conditional mathematical expectation is stochastically differentiable, and proposed a method for finding the integrand.

Here we consider some path-dependent Brownian functionals and derive constructive formulas for the stochastic integral representation. The considered class of functionals is not stochastically smooth, and both the well-known Clark-Ocone formula (1984) and its generalization, the Glonti-Purtukhia formula ([1]), are not applicable to them. Here we use the earlier Jaoshvili-Purtukhia ([2]) generalization of the Clark-Ocone formula, according to which the following result is valid.

Theorem 1. *Let f and its generalized derivative ∂f be square-integrable functions with the weight function $\exp\{-\frac{x^2}{2}\}$ and $\xi \in D_{1,2}$, then the following stochastic integral representation is valid:*

$$f(\xi) = Ef(\xi) + \int_0^T E[\partial f(\xi) D_t \xi | \mathfrak{S}_t^B] dB_t \quad (P - a.s.)$$

In addition, we needed two technical results about ordinary integrals, which are obviously moments of the standard normal distribution in the case of a complete space ($a = -\infty$), which can be easily verified using the method of mathematical induction.

Proposition 1. *For any real number a and non-negative integer $n \geq 0$ we have*

$$\begin{aligned} & \frac{1}{\sqrt{2\pi}} \int_a^{+\infty} x^{2n} \exp\{-\frac{x^2}{2}\} dx = \\ & = \varphi(a) \sum_{k=1}^n \frac{(2n-1)!!}{(2k-1)!!} a^{2k-1} + (2n-1)!! (1 - \Phi(a)), \end{aligned}$$

where Φ is the standard normal distribution function and φ is its density function, $(2n-1)!! := 1 \cdot 3 \cdot \dots \cdot (2n-1)$, $(-1)!! := 1$.

Proposition 2. *For any real number a and natural n we have*

$$\frac{1}{\sqrt{2\pi}} \int_a^{+\infty} x^{2n-1} \exp\{-\frac{x^2}{2}\} dx = \varphi(a) \sum_{k=0}^{n-1} \frac{(2n-2)!!}{(2k)!!} a^{2k},$$

where $(2n)!! := 2 \cdot 4 \cdot \dots \cdot (2n)$, $0!! := 1$.

2 Main results. For any non-negative integer n we denote

$$G(2n+1) := \left(\int_0^T B_s ds \right)^{2n+1}$$

and consider the following path-dependent Brownian functional:

$$F(2n+1) := [G(2n+1)]^+.$$

Theorem 2. *For any non-negative integer n the following stochastic integral representation is valid:*

$$F(2n+1) = EF(2n+1) + (2n+1) \sum_{k=0}^{2n} \int_0^T (T-t) C_{2n}^k \sigma^k \eta^{2n-k} I_k(\sigma, \eta) \Big|_{\eta=\int_0^t (T-s)dB_s} dB_t,$$

where

$$\sigma^2 := (T-t)^3/3, \quad I_{2k-1}(\sigma, \eta) = \varphi\left(\frac{\eta}{\sigma}\right) \sum_{i=0}^{k-1} \frac{(2k-2)!!}{(2i)!!} \left(\frac{\eta}{\sigma}\right)^{2i},$$

$$I_{2k}(\sigma, \eta) = (2k-1)!! \phi\left(\frac{\eta}{\sigma}\right) + \varphi\left(\frac{\eta}{\sigma}\right) \sum_{i=1}^k \frac{(2k-1)!!}{(2i-1)!!} \left(-\frac{\eta}{\sigma}\right)^{2i-1}.$$

Proof. It is not difficult to see that

$$\int_t^T (T-s)dB_s \sim N(0, (T-t)^3/3) = N(0, \sigma^2).$$

According to Theorem 1, we have

$$\begin{aligned} \varphi(t, \omega) &= E\left[I_{\left\{\int_0^T B_s ds\right\}^{2n+1} > 0}\right] (2n+1) \left(\int_0^T B_s ds\right)^{2n} \cdot \int_0^T I_{[0,s]}(t) ds \Big| \mathfrak{S}_t^B \\ &= (2n+1)(T-t) E\left[I_{\left\{\int_0^T (T-s)dB_s\right\} > 0}\right] \left(\int_0^T (T-s)dB_s\right)^{2n} \Big| \mathfrak{S}_t^B = (2n+1)(T-t) \\ &\quad \times E\left[I_{\left\{\int_t^T (T-s)dB_s + \int_0^t (T-s)dB_s\right\} > 0}\right] \left(\int_t^T (T-s)dB_s + \int_0^t (T-s)dB_s\right)^{2n} \Big| \mathfrak{S}_t^B. \end{aligned}$$

Further, thanks to the Markov property of Brownian Motion, using the Newton binomial formula and the values of the moments of the normal distribution, it is not difficult to establish that

$$\begin{aligned} \varphi(t, y) &= (2n+1)(T-t) E\left[I_{\left\{\int_t^T (T-s)dB_s > -y\right\}} \left(\int_t^T (T-s)dB_s + y\right)^{2n}\right] \Big|_{y=\int_0^t (T-s)dB_s} \\ &= \frac{(2n+1)(T-t)}{\sqrt{2\pi}\sigma} \sum_{r=0}^{2n} C_{2n}^r y^{2n-r} \int_{-y}^{+\infty} x^r \exp\left\{-\frac{x^2}{2\sigma^2}\right\} dx \Big|_{y=\int_0^t (T-s)dB_s} \\ &= (2n+1)(T-t) \sum_{r=0}^{2n} C_{2n}^r \sigma^r y^{2n-r} I_{2n}(r, \sigma, y) \Big|_{y=\int_0^t (T-s)dB_s}, \end{aligned}$$

where

$$I_{2n}(r, \sigma, y) := \frac{1}{\sqrt{2\pi}} \int_{-y/\sigma}^{+\infty} x^r \exp\{-\frac{x^2}{2}\} dx.$$

On the other hand, from Proposition 1 and 2, for any natural r we have

$$\begin{aligned} I_{2n}(2r-1, \sigma, y) &:= \frac{1}{\sqrt{2\pi}} \int_{-y/\sigma}^{\infty} x^{2r-1} \exp\{-\frac{x^2}{2}\} dx \\ &= \varphi(-\frac{y}{\sigma}) \sum_{k=0}^{r-1} \frac{(2r-2)!!}{(2k)!!} (-\frac{y}{\sigma})^{2k} = \varphi(\frac{y}{\sigma}) \sum_{k=0}^{r-1} \frac{(2r-2)!!}{(2k)!!} (\frac{y}{\sigma})^{2k}. \end{aligned}$$

and

$$\begin{aligned} I_{2n}(2r, \sigma, y) &:= \frac{1}{\sqrt{2\pi}} \int_{-y/\sigma}^{\infty} x^{2r} \exp\{-\frac{x^2}{2}\} dx \\ &= \varphi(-\frac{y}{\sigma}) \sum_{k=1}^r \frac{(2r-1)!!}{(2k-1)!!} (-\frac{y}{\sigma})^{2k-1} + (2r-1)!! (1 - \phi(-y/\sigma)) \\ &= \varphi(\frac{y}{\sigma}) \sum_{k=1}^r \frac{(2r-1)!!}{(2k-1)!!} (-\frac{y}{\sigma})^{2k-1} + (2r-1)!! \phi(y/\sigma). \end{aligned}$$

□

Corollary. The following stochastic integral representation is valid:

$$\left(\int_0^T B_s ds \right)^+ = \sqrt{\frac{T^3}{6\pi}} + \int_0^T (T-t) \phi(\sigma^{-1} \int_0^t (T-s) dB_s) dB_t.$$

R E F E R E N C E S

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