

# Stochastic Differential Equations with Anticipating Initial Conditions

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# **SECTION 1**

## **INTRODUCTION AND MOTIVATION**

# Review and notations

- ▷ Let  $T \in (0, \infty)$ ,  $t \in [0, T]$ , and  $B_t$  is a Brownian motion.
- ▷ The **Itô integral** for an **adapted** process  $X_t$  w.r.t.  $B_t$  is defined as

$$\int_0^t X_s \, dB_s = \mathbb{P} \lim_{\|\Delta_n\| \rightarrow 0} \sum_{j=0}^{n-1} X_{t_j} \Delta B_j, \text{ where } \Delta_n \text{ is a partition of } [0, t] \text{ and } \Delta B_j = B_{t_{j+1}} - B_{t_j}.$$

- ▷ The process  $Y_t = \int_0^t X_s \, dB_s$  is a martingale.
- ▷ An **Itô process** is an **adapted** process of the form  $X_t = X_0 + \int_0^t m_s \, ds + \int_0^t \sigma_s \, dB_s$ .  
Equivalently expressed as  $dX_t = m_t \, dt + \sigma_t \, dB_t$ .
- ▷ **Itô formula**: If  $X_t$  is an Itô process and  $f(t, x) \in C^{1,2}(\mathbb{R} \times \mathbb{R})$ , then  $Y_t = f(t, X_t)$  is also an Itô process given by

$$dY_t = df(t, X_t) = \partial_t f(t, X_t) \, dt + \partial_x f(t, X_t) \, dX_t + \frac{1}{2} \partial_x^2 f(t, X_t) (dX_t)^2.$$

# Remarks on the Itô integral

▷ The integrand must be an **adapted** stochastic process.

▷ Iterated integrals:  $\int_0^T \int_0^T dB_u dB_v = \int_0^T B_T dB_v = ?$

▷ What about anticipating stochastic differential equations like

$$\begin{cases} dX_t = X_t dB_t \\ X_0 = B_T \end{cases} \quad \text{or} \quad \begin{cases} dY_t = B_T dB_t \\ Y_0 = 1 \end{cases} \quad ?$$

▷ Problem: We want to define  $\int_0^T Z_t dB_t$ , where  $Z_t$  is not (necessarily) adapted.

▷ Some approaches

- Enlargement of filtration [Itô78]
- Malliavin calculus
- White noise theory
- Numerous others

# **SECTION 2**

## **THE GENERALIZED INTEGRAL**

# Definition of the integral [AK08; AK10]

▷ A process  $Y^t$  and filtration  $\mathcal{F}_t$  are called **instantly independent** if  $Y^t$  and  $\mathcal{F}_t$  are independent  $\forall t$ .  
Example: The process  $B_T - B_t$  is instantly independent of the filtration generated by  $B_t$ .

▷ Idea

1. Decompose the integrand into **adapted** and **instantly independent** parts.
2. Evaluate the **adapted** and the **instantly independent** parts at the **left** and **right** endpoints.

▷ Consider two continuous stochastic processes,  $X_t$  **adapted** and  $Y^t$  **instantly independent** w.r.t.  $\mathcal{F}_t$ . Then the integral  $\int_0^T X_t Y^t dB_t$  is **defined** as

$$\int_0^T X_t Y^t dB_t \triangleq \mathbb{P} \lim_{\|\Delta_n\| \rightarrow 0} \sum_{j=0}^{n-1} X_{t_j} Y^{t_{j+1}} \Delta B_j.$$

▷ Now, for any stochastic process  $Z(t) = \sum_{k=1}^n X_t^{(k)} Y_{(k)}^t$  we extend the definition by linearity. This is well-defined [HKS+16].

# A simple example

$$\begin{aligned}\int_0^t B_T \, dB_s &= \int_0^t (B_s + (B_T - B_s)) \, dB_s = \int_0^t B_s \, dB_s + \int_0^t (B_T - B_s) \, dB_s \\&= L^2 \lim_{\|\Delta_n\| \rightarrow 0} \sum_{j=0}^{n-1} B_{t_j} \Delta B_j + L^2 \lim_{\|\Delta_n\| \rightarrow 0} \sum_{j=0}^{n-1} (B_T - B_{t_{j+1}}) \Delta B_j \\&= L^2 \lim_{\|\Delta_n\| \rightarrow 0} \sum_{j=0}^{n-1} (B_T - \Delta B_j) \Delta B_j \\&= B_T \cdot L^2 \lim_{\|\Delta_n\| \rightarrow 0} \sum_{j=0}^{n-1} \Delta B_j - L^2 \lim_{\|\Delta_n\| \rightarrow 0} \sum_{j=0}^{n-1} (\Delta B_j)^2 \\&= B_T B_t - t\end{aligned}$$



# The near-martingale property

- ▷ Question: What is the analogues of the martingale property?
- ▷ Example:  $\mathbb{E}(B_T B_t - t \mid \mathcal{F}_s) = B_s^2 - s \neq B_T B_s - s$ . ☹  
But  $\mathbb{E}(B_T B_s - s \mid \mathcal{F}_s) = B_s^2 - s$ . ☺
- ▷ Let  $Z_t$  be a process such that  $\mathbb{E}|Z_t| < \infty \forall t$ , and  $0 \leq s \leq t \leq T$ . Then  $Z_t$  is called a **near-martingale** if  $\mathbb{E}(Z_t \mid \mathcal{F}_s) = \mathbb{E}(Z_s \mid \mathcal{F}_s)$ .

**Theorem 1** ([[KSS12b](#)]) *Let  $f$  and  $\phi$  be continuous functions on  $\mathbb{R}$ . Under integrability conditions, the processes  $X_t = \int_0^t f(B_t)\phi(B_T - B_t) dB_t$  and  $Y^t = \int_t^T f(B_t)\phi(B_T - B_t) dB_t$  are near-martingales.*

**Theorem 2** ([[HKS+17](#)]) *Let  $Z_t$  be a stochastic process bounded in  $L^1$ , and  $X_t = \mathbb{E}(Z_t \mid \mathcal{F}_t)$ . Then  $X_t$  is a martingale if and only if  $Z_t$  is a near-martingale.*

# The general Itô formula [HKS+16]

Process	Definition	Representation
Itô	$X_t = X_0 + \int_0^t m_s \, ds + \int_0^t \sigma_s \, dB_s$	$dX_t = m_t \, dt + \sigma_t \, dB_t$
instantly independent	$Y^t = Y^T + \int_t^T \eta^s \, ds + \int_t^T \zeta^s \, dB_s$	$dY^t = -\eta^t \, dt - \zeta^t \, dB_t$

Here  $\eta^t$  and  $\zeta^t$  are instantly independent such that  $Y^t$  is also instantly independent.

**Theorem 3** ([HKS+16]) Let  $dX_t = m_t \, dt + \sigma_t \, dB_t$  be an Itô process, and  $dY^t = -\eta^t \, dt - \zeta^t \, dB_t$  be a *instantly independent* process. If  $f(t, x, y) \in C^{1,2,2}(\mathbb{R} \times \mathbb{R} \times \mathbb{R})$ , then

$$\begin{aligned}
 df(t, X_t, Y^t) = & \partial_t f(t, X_t, Y^t) \, dt + \partial_x f(t, X_t, Y^t) \, dX_t + \frac{1}{2} \partial_x^2 f(t, X_t, Y^t) (dX_t)^2 \\
 & + \partial_y f(t, X_t, Y^t) \, dY^t - \frac{1}{2} \partial_y^2 f(t, X_t, Y^t) (dY^t)^2.
 \end{aligned}$$

# **SECTION 3**

## **CONDITIONAL EXPECTATION**

# Linear stochastic differential equations

**Definition 4** Define the *exponential process* with parameters  $\alpha$  and  $\beta$  by

$$\mathcal{E}_t^{(\alpha, \beta)} = \exp \left( \int_0^t \alpha_s dB_s + \int_0^t \left( \beta_s - \frac{1}{2} \alpha_s^2 \right) ds \right).$$

**Theorem 5** ([HKS+16]) The solution of the stochastic differential equation

$$\begin{cases} dX_t = \alpha_t X_t dB_t + \beta_t X_t dt \\ X_0 = \psi(B_T) \end{cases}$$

is given by  $X_t = \psi \left( B_T - \int_0^t \alpha_s ds \right) \mathcal{E}_t^{(\alpha, \beta)}$ .

# Motivating question

What can we say about the conditional expectation of the solution of the stochastic differential equation

$$\begin{cases} dX_t = \alpha_t X_t dB_t + \beta_t X_t dt \\ X_0 = \psi(B_T) \end{cases} \quad ?$$

In particular, if  $Y_t = \mathbb{E}(X_t \mid \mathcal{F}_t)$ , can we expect  $Y_t$  to be the solution of the stochastic differential equation

$$\begin{cases} dY_t = \alpha_t Y_t dB_t + \beta_t Y_t dt \\ Y_0 = \mathbb{E}\psi(B_T) \end{cases} \quad ?$$

# Unexpected behaviour

**Theorem 6** ([KSZ18]) Suppose  $\alpha_t \in L^2[0, T]$ ,  $\beta_t$  is adapted with  $\mathbb{E} \int_0^T |\beta_t|^2 dt < \infty$ , and  $\psi : \mathbb{R} \rightarrow \mathbb{R}$  has power series expansion at 0 with infinite radius of convergence, and  $\psi'$  denotes the derivative of  $\psi$ .

Consider the two stochastic differential equations

$$\begin{cases} dX_t = \alpha_t X_t dB_t + \beta_t X_t dt \\ X_0 = \psi(B_T) \end{cases} \quad \text{and} \quad \begin{cases} d\bar{X}_t = \alpha_t \bar{X}_t dB_t + \beta_t \bar{X}_t dt \\ \bar{X}_0 = \psi'(B_T) \end{cases}.$$

Denote  $Y_t = \mathbb{E}(X_t | \mathcal{F}_t)$  and  $\bar{Y}_t = \mathbb{E}(\bar{X}_t | \mathcal{F}_t)$ .

Then  $Y_t$  satisfies the stochastic differential equation

$$\begin{cases} dY_t = \alpha_t Y_t dB_t + \beta_t Y_t dt + \bar{Y}_t dB_t \\ Y_0 = \mathbb{E}\psi(B_T) \end{cases}.$$

## A brief detour: Hermite polynomials

- ▷ An Hermite polynomial of degree  $n$  with parameter  $\rho$  is given by

$$H_n(x; \rho) = (-\rho)^n e^{\frac{x^2}{2\rho}} \partial_x^n e^{-\frac{x^2}{2\rho}}.$$

- ▷ The first few Hermite polynomials are:  $1, x, x^2 - \rho, x^3 - 3\rho x, x^4 - 6\rho x^2 + 3\rho^2, \dots$
- ▷ Hermite polynomials form an orthonormal basis of  $L^2(\mathbb{R}, \gamma)$ , where  $\gamma$  is the Gaussian measure with mean 0 and variance  $\rho$ .
- ▷ For fixed  $n \in \mathbb{N}$ , the stochastic process  $X_t = H_n(B_t; t)$  is a martingale, and

$$dX_t = nH_{n-1}(B_t; t) dB_t.$$

# Initial condition: Hermite polynomials

**Theorem 7** ([KSZ18]) Suppose  $\alpha_t \in L^2[0, T]$ ,  $\beta_t$  is adapted with  $\mathbb{E} \int_0^T |\beta_t|^2 dt < \infty$ , and let  $n$  be a fixed natural number. Let  $X_t$  be the solution of

$$\begin{cases} dX_t = \alpha_t X_t dB_t + \beta_t X_t dt \\ X_0 = H_n(B_T; T), \end{cases}$$

and  $Y_t = \mathbb{E}(X_t | \mathcal{F}_t)$ .

Then  $Y_t$  satisfies the stochastic differential equation

$$\begin{cases} dY_t = \alpha_t Y_t dB_t + \beta_t Y_t dt + n H_{n-1}\left(B_t - \int_0^t \alpha_s ds; t\right) \mathcal{E}_t^{(\alpha, \beta)} dB_t \\ Y_0 = 0 \end{cases},$$

and is explicitly given by

$$Y_t = H_n\left(B_t - \int_0^t \alpha_s ds; t\right) \mathcal{E}_t^{(\alpha, \beta)}.$$



Initial condition: differentiable function in  $L^2(\mathbb{R}, \gamma)$

**Theorem 8** ([KSZ18]) Suppose  $\alpha_t \in L^2[0, T]$ ,  $\beta_t$  is adapted with  $\mathbb{E} \int_0^T |\beta_t|^2 dt < \infty$ .

Let  $\psi(x) = \sum_{n=0}^{\infty} c_n H_n(x; T)$  be a differentiable function in  $L^2(\mathbb{R}, \gamma)$ .

Consider the two stochastic differential equations

$$\begin{cases} dX_t = \alpha_t X_t dB_t + \beta_t X_t dt \\ X_0 = \psi(B_T) \end{cases} \quad \text{and} \quad \begin{cases} d\bar{X}_t = \alpha_t \bar{X}_t dB_t + \beta_t \bar{X}_t dt \\ \bar{X}_0 = \psi'(B_T) \end{cases} .$$

Denote  $Y_t = \mathbb{E}(X_t | \mathcal{F}_t)$  and  $\bar{Y}_t = \mathbb{E}(\bar{X}_t | \mathcal{F}_t)$ .

Then  $Y_t$  satisfies the stochastic differential equation

$$\begin{cases} dY_t = \alpha_t Y_t dB_t + \beta_t Y_t dt + \bar{Y}_t dB_t \\ Y_0 = \mathbb{E}\psi(B_T) \end{cases} ,$$

and is explicitly given by

$$Y_t = \sum_{n=0}^{\infty} c_n H_n \left( B_t - \int_0^t \alpha_s ds; t \right) \mathcal{E}_t^{(\alpha, \beta)} .$$

# SECTION 4

## A LARGER CLASS OF INITIAL CONDITIONS

# Initial condition: Wiener integral

**Question:** Can we extend the class of initial conditions?

**Theorem 9** ([KSZ18]) *Let  $\alpha_t \in L^2[0, T], \beta_t \in L^1[0, T], h_t \in L^2[0, T], \psi(t) \in C^2(\mathbb{R})$ . Then the (unique) solution of the stochastic differential equation*

$$\begin{cases} dX_t = \alpha_t X_t dB_t + \beta_t X_t dt \\ X_0 = \psi\left(\int_0^T h_s dB_s\right) \end{cases}$$

*is given by*

$$X_t = \psi\left(\int_0^T h_s dB_s - \int_0^t \alpha_s h_s ds\right) \mathcal{E}_t^{(\alpha, \beta)}.$$

# An example

Consider the stochastic differential equation

$$\begin{cases} \mathrm{d}X_t = X_t \mathrm{d}B_t \\ X_0 = \psi\left(\int_0^T B_s \mathrm{d}s\right) \end{cases} \quad .$$

Using Itô lemma, we rewrite  $\int_0^T B_s \mathrm{d}s = \int_0^T (T - s) \mathrm{d}B_s$ .

Now using the previous theorem, we get

$$X_t = \psi\left(\int_0^T B_s \mathrm{d}s - \left(Tt - \frac{1}{2}t^2\right)\right) e^{B_t - \frac{1}{2}t}.$$

Thank you!

# APPENDIX

# Iterated integrals

**Theorem 10** ([[Itô51](#)]) *Let  $f \in L^2([0, T]^n)$  and  $\hat{f}$  be its symmetrization. Then*

$$\int_{[0, T]^n} f(t_1, \dots, t_n) dB_{t_1} \dots dB_{t_n} = n! \int_0^T \dots \int_0^{t_{n-2}} \left( \int_0^{t_{n-1}} \hat{f}(t_1, \dots, t_n) dB_{t_n} \right) dB_{t_{n-1}} \dots dB_{t_1}.$$

**Theorem 11** ([[AK10](#)]) *Let  $f \in L^2([0, T]^n)$ . Then*

$$\int_{[0, T]^n} f(t_1, \dots, t_n) dB_{t_1} \dots dB_{t_n} = \int_0^T \dots \int_0^T f(t_1, \dots, t_n) dB_{t_n} \dots dB_{t_1}.$$

Example[[HKS+16](#)]: For the new integral,  $\int_0^T \left( \int_0^T B_u du \right) dB_v = \int_0^T \left( \int_0^T B_u dB_v \right) du$ .

# A generalization of Itô isometry

**Theorem 12** ([[KSS12b](#)]) *Let  $\phi$  be an analytic function on  $\mathbb{R}$ . Then under integrability conditions and for each  $t$ ,*

$$\mathbb{E} \left[ \left( \int_0^t \phi(B_T - B_s) dB_s \right)^2 \right] = \int_0^t \mathbb{E} \left[ (\phi(B_T - B_s))^2 \right] ds$$

**Theorem 13** ([[KSS13](#)]) *Let  $f$  and  $\phi$  be  $C^1$  functions on  $\mathbb{R}$ . Then*

$$\begin{aligned} \mathbb{E} \left[ \left( \int_0^T f(B_t) \phi(B_T - B_t) dB_t \right)^2 \right] &= \int_0^T \mathbb{E} \left[ (f(B_t) \phi(B_T - B_t))^2 \right] dt \\ &\quad + 2 \int_0^T \int_0^{\textcolor{brown}{t}} \mathbb{E} [f(B_s) \phi'(B_T - B_s) f'(B_s) \phi(B_T - B_s)] ds dt. \end{aligned}$$



# A generalization of Girsanov theorem

**Theorem 14** ([KPS13]) *Let  $X_t$  and  $Y^t$  be continuous square-integrable stochastic processes such that  $X_t$  is adapted and  $Y^t$  is instantly independent.*  
*Then the translated stochastic process  $W_t = B_t - \int_0^t (X_t + Y^t) dt$  is a near-martingale under the probability measure  $\widetilde{\mathbb{P}}$  defined by the Radon-Nikodym derivative  $\frac{d\widetilde{\mathbb{P}}}{d\mathbb{P}} = \mathcal{E}_T^{(X+Y,0)}$ .*

Thank you!

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