

Lecture 5: Ito's Calculus

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Now that we have constructed an integral with respect to Brownian trajectories, we need an associated differential calculus.

1 Ito's formula for the Brownian motion

If $(B_t)_{t \geq 0}$ is an $(\mathcal{F}_t)_{t \geq 0}$ -BM and $\Phi : \mathbb{R} \rightarrow \mathbb{R}$ is C^2 then for any $T > 0$,

$$\Phi(B_T) = \Phi(B_0) + \int_0^T \Phi'(B_t) dB_t + \frac{1}{2} \int_0^T \Phi''(B_t) dt, \text{ almost surely.}$$

We rewrite this under the short-hand notation

$$d\Phi(B_t) = \Phi'(B_t) dB_t + \frac{1}{2} \Phi''(B_t) dt.$$

We will call $d\Phi(B_t)$ the *stochastic differential* of $\Phi(B_t)$.

2 Ito process

An Ito process is a process $(X_t)_{t \geq 0}$ which writes under the form

$$\forall T > 0, \quad X_T = X_0 + \int_0^T K_t dt + \int_0^T H_t dB_t, \quad \text{almost surely,}$$

with:

- X_0 \mathcal{F}_0 -measurable;
- $(K_t)_{t \geq 0}$ progressively measurable and such that $\int_0^T |K_t| dt < \infty$, a.s., for all $T > 0$;
- $(H_t)_{t \geq 0}$ progressively measurable and such that $\int_0^T H_t^2 dt < \infty$, a.s., for all $T > 0$.

We use the stochastic differential notation $dX_t = K_t dt + H_t dB_t$.

An Ito process is adapted and a.s. continuous.

We want to prove that the decomposition of an Ito process into a stochastic integral and an absolutely continuous part is unique. The first step of the argument is given by the following statement.

Lemme 2.1 (Quadratic variation). *Let $(X_t)_{t \geq 0}$ be an Ito process. Then, for any $T > 0$,*

$$\lim_{\|\sigma\| \rightarrow 0} \sum_{i=0}^{n-1} (X_{t_i} - X_{t_{i-1}})^2 = \int_0^T H_t^2 dt, \quad \text{in probability,}$$

with $\sigma = (t_0, \dots, t_n)$ a subdivision of $[0, T]$.

Corollary 2.2 (Uniqueness of the decomposition of an Ito process). *If there are $X'_0, (K'_t)_{t \geq 0}$ and $(H'_t)_{t \geq 0}$ such that*

$$\forall T > 0, \quad X_0 + \int_0^T K_t dt + \int_0^T H_t dB_t = X'_0 + \int_0^T K'_t dt + \int_0^T H'_t dB_t, \quad \text{almost surely,}$$

then $X_0 = X'_0$, a.s., and $K_t = K'_t, H_t = H'_t, \mathbb{P} \otimes dt$ -a.e.

Proof. For X_0, X'_0 just take $T = 0$. Then the lemma above allows to identify H_t and H'_t , which finally yields the identification of K_t and K'_t . \square

From this corollary there is now no ambiguity in defining the quadratic variation of $(X_t)_{t \geq 0}$ as the process $(\langle X \rangle_t)_{t \geq 0}$ defined by

$$\forall T \geq 0, \quad \langle X \rangle_T = \int_0^T H_t^2 dt.$$

The BM $(B_t)_{t \geq 0}$ is an Ito process with quadratic variation $\langle B \rangle_t = t$. Any absolutely continuous process $X_t = X_0 + \int_0^t K_s ds$ is an Ito process with quadratic variation $\langle X \rangle_t = 0$.

Theorem 2.3 (Ito's formula for Ito processes). *For any C^2 function $\Phi : \mathbb{R} \rightarrow \mathbb{R}$, $d\Phi(X_t) = \Phi'(X_t)dX_t + \frac{1}{2}\Phi''(X_t)d\langle X \rangle_t$.*

Application: the Geometric Brownian motion.

Proposition 2.4 (Lévy's characterisation of the BM). *If $(H_t)_{t \geq 0}$ is progressively measurable and $H_t^2 = 1$, $\mathbb{P} \otimes dt$ -a.e., then $X_t := \int_0^t H_s dB_s$ is a Brownian motion.*

This is in fact a particular case of the following statement.

Theorem 2.5 (Dambis–Dubins–Schwarz). *If $(H_t)_{t \geq 0} \in \mathbf{\Lambda}_{\text{loc}}$ then there exists a Brownian motion $(\beta_r)_{r \geq 0}$ such that*

$$\forall T > 0, \quad X_t = \int_0^t H_s dB_s = \beta_{\langle X \rangle_t} = \beta_{\int_0^t H_s^2 ds}.$$

It is useful when you want to study trajectories of the stochastic integral: they are time-changes of Brownian trajectories.

3 Multidimensional Ito's formula

We now take d independent $(\mathcal{F}_t)_{t \geq 0}$ -BM B^1, \dots, B^d . An Itô process now writes

$$dX_t = K_t dt + \sum_{k=1}^d H_t^k dB_t^k.$$

The extension of Lemma 2.1 then writes

$$\lim_{\|\sigma\| \rightarrow 0} \sum_{i=0}^{n-1} (X_{t_i} - X_{t_{i-1}})^2 = \sum_{k=1}^d \int_0^T (H_t^k)^2 dt, \quad \text{in probability.}$$

This allows to show again that if $dX_t = K'_t dt + \sum_{k=1}^d H_t^k dB_t^k$ then $K_t = K'_t$ and for all k , $H_t^k = H_t'^k$, $\mathbb{P} \otimes dt$ -a.e., and also to define

$$\langle X \rangle_T = \sum_{k=1}^d \int_0^T (H_t^k)^2 dt.$$

By polarisation, for arbitrary Ito processes X, X' , we then have

$$\langle X, X' \rangle_T := \sum_{k=1}^d \int_0^T H_t^k H_t'^k dt = \lim_{\|\sigma\| \rightarrow 0} \sum_{i=0}^{n-1} (X_{t_i} - X_{t_{i-1}})(X'_{t_i} - X'_{t_{i-1}}), \quad \text{in probability.}$$

Theorem 3.1 (Multidimensional Ito's formula). *Let X^1, \dots, X^n be Itô processes and $\Phi : \mathbb{R}^n \rightarrow \mathbb{R}$ be C^2 . Then, with $X_t = (X_t^1, \dots, X_t^n)$,*

$$d\Phi(X_t) = \sum_{i=1}^n \frac{\partial \Phi}{\partial x_i}(X_t) dX_t^i + \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 \Phi}{\partial x_i \partial x_j}(X_t) d\langle X^i, X^j \rangle_t.$$

Particular case of time-dependent Φ .

Applications: integration by parts; recurrence and transience of the BM is dimension $d \geq 2$.