

# Day 17 : Itô's Formula

Dain Kim

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# Today's Reading

[L] Section 5.1.4. Convergence of Stochastic Integrals  
Section 5.2. Itô's Formula

## Review : Continuous Semimartingales

A process  $X = (X_t)$ ,  $t \geq 0$ , is a *continuous semimartingale* if there is a decomposition

$$X_t = M_t + A_t$$

so that  $M$  is a continuous local martingale and  $A$  is a finite variation process.

Such decomposition is unique and called the *canonical decomposition*.

For two  $X = M + A$ ,  $Y = M' + A'$  continuous semimartingales with canonical decompositions, the *bracket* is the finite variation process defined by

$$\langle X, Y \rangle_t = \langle M, M' \rangle_t.$$

## Review : Stochastic Integrals for Semimartingales

A progressive process  $H$  is *locally bounded* if

$$\forall t \geq 0, \quad \sup_{s \leq t} |H_s| < \infty, \quad \text{a.s.}$$

Note that if  $H$  is progressive and locally bounded, then for any finite variation process  $V$ , we have

$$\forall t \geq 0, \quad \int_0^t |H_s| |dV_s| < \infty, \quad \text{a.s.}$$

and similarly  $H \in L^2_{\text{loc}}(M)$  for every continuous local martingale  $M$ .

Hence, we may define  $H \cdot X$  for a continuous semimartingale  $X$  with canonical decomposition  $X = M + V$  by

$$H \cdot X = H \cdot M + H \cdot V.$$

We write  $(H \cdot X)_t = \int_0^t H_s dX_s$ .

# Recall : Lebesgue Dominated Convergence Theorem

## Theorem

Given a measure  $\mu$  on  $\mathbb{R}$ , for a sequence of measurable function  $f_n: \mathbb{R} \rightarrow [-\infty, \infty]$  with pointwise limit  $f(x) = \lim_{n \rightarrow \infty} f_n(x)$ , if there exists an integrable function  $g$ , i.e.,  $\int g d\mu < +\infty$ , such that  $|f_n(x)| \leq g(x)$  for all  $n$  and  $x$ , then  $f_n$  and  $f$  are integrable and

$$\lim_{n \rightarrow \infty} \int_{\mathbb{R}} f_n d\mu = \int_{\mathbb{R}} \lim_{n \rightarrow \infty} f_n d\mu = \int_{\mathbb{R}} f d\mu.$$

# Dominated Convergence Theorem for Stochastic Integrals

## Proposition

Given a semimartingale  $X$ , let  $X = M + V$  be the canonical decomposition of  $X$ . Let  $t > 0$ . Let  $(H^n)_{n \geq 1}$  and  $H$  be locally bounded progressive processes and  $K$  a nonnegative progressive process. Assume the following properties hold a.s.:

- (i)  $H^n \rightarrow H$  as  $n \rightarrow \infty$ , for every  $s \in [0, t]$ ,
- (ii)  $|H^n_s| \leq K_s$ , for every  $n \geq 1$  and  $s \in [0, t]$ ,
- (iii)  $\int_0^t (K_s)^2 d\langle M, M \rangle_s < \infty$  and  $\int_0^t K_s |dV_s| < \infty$ .

Then

$$\int_0^t H^n_s dX_s \xrightarrow{n \rightarrow \infty} \int_0^t H_s dX_s$$

in probability.

*Remark.* The assumption (iii) automatically holds if  $K$  is locally bounded.

# An Approximation of Continuous Integrands

## Proposition

Let  $X$  be a continuous semimartingale and  $H$  an adapted process with continuous sample paths. Then for every  $t > 0$  and subdivisions

$$0 = t_0^n < \cdots < t_{p_n}^n = t,$$

$$\lim_{\sup_i |t_{i+1} - t_i| \rightarrow 0} \sum_{i=0}^{p_n-1} H_{t_i^n} (X_{t_{i+1}^n} - X_{t_i^n}) = \int_0^t H_s dX_s.$$

*Proof Sketch.* Define a process  $H^n$  by

$$H_s^n = \begin{cases} H_{t_i^n} & \text{if } s \in (t_i^n, t_{i+1}^n] \\ H_0 & \text{if } s = 0 \\ 0 & \text{if } s > t \end{cases}$$

and  $K_s = \max_{0 \leq r \leq s} |H_s|$ . Then apply the Dominated Convergence Theorem. □

# An Approximation of Continuous Integrands

What if we take the right end of the interval  $(t_i^n, t_{i+1}^n]$ ? Consider a special case  $H_t = X_t$ , then by the proposition,

$$\lim_{\sup_i |t_{i+1} - t_i| \rightarrow 0} \sum_{i=0}^{p_n-1} X_{t_i^n} (X_{t_{i+1}^n} - X_{t_i^n}) = \int_0^t X_s dX_s \quad (1)$$

in probability. On the other hand, note that

$$\sum_{i=0}^{p_n-1} X_{t_{i+1}^n} (X_{t_{i+1}^n} - X_{t_i^n}) = \sum_{i=0}^{p_n-1} X_{t_i^n} (X_{t_{i+1}^n} - X_{t_i^n}) + \sum_{i=0}^{p_n-1} (X_{t_{i+1}^n} - X_{t_i^n})^2,$$

which gives

$$\lim_{\sup_i |t_{i+1} - t_i| \rightarrow 0} \sum_{i=0}^{p_n-1} X_{t_{i+1}^n} (X_{t_{i+1}^n} - X_{t_i^n}) = \int_0^t X_s dX_s + \langle X, X \rangle_t \quad (2)$$

in probability.



# An Approximation of Continuous Integrands

Adding (1) and (2), we get

$$(X_t)^2 - (X_0)^2 = 2 \int_0^t X_s dX_s + \langle X, X \rangle_t.$$

This is a special case of Itô's formula.

# Itô's Formula

## Theorem (1-dimensional Itô's formula)

Let  $X$  be a continuous semimartingale and  $F: \mathbb{R} \rightarrow \mathbb{R}$  a twice continuously differentiable real function. Then for every  $t \geq 0$ ,

$$F(X_t) = F(X_0) + \int_0^t F'(X_s) dX_s + \frac{1}{2} \int_0^t F''(X_s) d\langle X, X \rangle_s.$$

## Theorem ( $p$ -dimensional Itô's formula)

Let  $X^1, \dots, X^p$  be  $p$  continuous semimartingales and  $F: \mathbb{R}^p \rightarrow \mathbb{R}$  a twice continuously differentiable real function. Then for every  $t \geq 0$ ,

$$\begin{aligned} F(X_t^1, \dots, X_t^p) &= F(X_0^1, \dots, X_0^p) + \sum_{i=1}^p \int_0^t \frac{\partial F}{\partial x^i}(X_s^1, \dots, X_s^p) dX_s^i \\ &\quad + \frac{1}{2} \sum_{i,j=1}^p \int_0^t \frac{\partial^2 F}{\partial x^i \partial x^j}(X_s^1, \dots, X_s^p) d\langle X^i, X^j \rangle_s. \end{aligned}$$

## Special Cases of Itô's formula

- Taking  $p = 2$  and  $F(x, y) = xy$ , we get

$$X_t Y_t = X_0 Y_0 + \int_0^t X_s dY_s + \int_0^t Y_s dX_s + \langle X, Y \rangle_t.$$

This is often called the *formula of integration by parts*.

- If  $X = Y$  in the above example, we get

$$X_t^2 = X_0^2 + 2 \int_0^t X_s dX_s + \langle X, X \rangle_t.$$

# Special Cases of Itô's formula - Brownian motions

Let  $B$  be an  $(\mathcal{F}_t)$ -real Brownian motion.

- Since  $\langle B, B \rangle_t = t$ , Itô's formula gives

$$F(B_t) = F(B_0) + \int_0^t F'(B_s) dB_s + \frac{1}{2} \int_0^t F''(B_s) ds.$$

- Let  $p = 2$  and  $X_t^1 = t, X_t^2 = B_t$ . For every twice continuously differentiable function  $F(t, x): \mathbb{R}_{\geq 0} \times \mathbb{R} \rightarrow \mathbb{R}$ , Itô's formula gives

$$F(t, B_t) = F(0, B_0) + \int_0^t \frac{\partial F}{\partial x}(s, B_s) dB_s + \int_0^t \left( \frac{\partial F}{\partial t} + \frac{1}{2} \frac{\partial^2 F}{\partial x^2} \right)(s, B_s) ds.$$

## Special Cases of Itô's formula - Brownian motions

Let  $B_t = (B_t^1, \dots, B_t^d)$  be a  $d$ -dimensional  $(\mathcal{F}_t)$ -Brownian motion. Since  $\langle B^i, B^j \rangle_t = 0$  if  $i \neq j$ , Itô's formula gives

$$\begin{aligned} F(B_t^1, \dots, B_t^d) &= F(B_0^1, \dots, B_0^d) + \sum_{i=1}^d \int_0^t \frac{\partial F}{\partial x_i}(B_s^1, \dots, B_s^d) dB_s^i \\ &\quad + \frac{1}{2} \int_0^t \Delta F(B_s^1, \dots, B_s^d) ds. \end{aligned}$$

The formula is often written as

$$F(B_t) = F(B_0) + \int_0^t \nabla F(B_s) \cdot dB_s + \frac{1}{2} \int_0^t \Delta F(B_s) ds.$$