

## Stochastic Calculus

## Ito's integrals

White Noise

$\{W(t), t \geq 0\}$  SBM/Weiner process,  $f$  having a cont. deriv. on  $[a, b]$

sto. integral

$$\int_a^b f dW(t) \equiv \lim_{\substack{n \rightarrow \infty \\ \max(t_i - t_{i-1}) \rightarrow 0}} \sum_{i=1}^n f(t_{i-1}) [W(t_i) - W(t_{i-1})] \rightarrow (4)$$

where  $a = t_0 < t_1 < \dots < t_n = b$  is a partition of  $[a, b]$

$$\sum_{i=1}^n f(t_{i-1}) (W(t_i) - W(t_{i-1})) = f(b)W(b) - f(a)W(a) - \sum_{i=1}^n W(t_i) (f(t_i) - f(t_{i-1}))$$

$$\begin{aligned} & \cancel{f(t_n)W(t_n)} - \cancel{f(t_0)W(t_0)} - W(t_1)(f(t_1) - f(t_0)) \\ & - W(t_2)(f(t_2) - f(t_1)) - \dots - W(t_n)(\cancel{f(t_n)} - f(t_{n-1})) \end{aligned}$$

defn

$$\boxed{\int_a^b f(t) dW(t) = f(b)W(b) - f(a)W(a) - \int_a^b W(t) df(t)} \quad \star$$

$$E\left(\int_a^b f(t) dW(t)\right) = 0$$

$$V\left(\int_a^b f(t) dW(t)\right) = E\left(\underbrace{\left(\int_a^b f(t) dW(t)\right)^2}_{\text{Ito's isometry}}\right) = \int_a^b f^2(t) dt,$$

Since

$$V\left(\sum_{i=1}^n f(t_{i-1}) (W(t_i) - W(t_{i-1}))\right) = \sum_{i=1}^n f^2(t_{i-1}) \underbrace{V(W(t_i) - W(t_{i-1}))}_{= t_i - t_{i-1}}$$

$l=1$

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$t_i - t_{i-1}$

$$= \sum_{i=1}^n f^2(t_{i-1}) \times (t_i - t_{i-1})$$

2) integrand  $f_t \equiv f(t)$  is non-random

$$\int_a^b f(t) dW(t) \sim N\left(0, \int_a^b f^2(t) dt\right)$$

Example (1) Find the distribution of  $X := \int_0^t s dW_s$

sol Since integrand is non-random

$$X \sim N\left(0, \int_0^t s^2 ds\right) \equiv N\left(0, \frac{t^3}{3}\right)$$

$$(2) \int_0^t W_s dW_s \simeq \sum_j W_{t_j} (W_{t_{j+1}} - W_{t_j})$$

$$\stackrel{*}{=} \frac{W_t^2}{2} - \frac{1}{2} \sum_j (W_{t_{j+1}} - W_{t_j})^2 \rightarrow \frac{W_t^2}{2} - \frac{t}{2}$$

$$\left( \begin{array}{l} E_x. \\ * \end{array} \right. \sum_{k=1}^n a_{k-1} (a_k - a_{k-1}) = \frac{1}{2} a_n^2 - \frac{1}{2} \sum_{k=1}^n (a_k - a_{k-1})^2$$

$$\sqrt{h} Z \stackrel{d}{=} W_{t+h} - W_t \sim N(0, h) \quad , \quad Z \sim N(0, 1)$$

$$\sum_{j=1}^n (W_{tj/n} - W_{t(j-1)/n})^2 \stackrel{d}{=} \sum_{j=1}^n \left( \sqrt{\frac{t}{n}} Z_j \right)^2$$

$$= t \underbrace{\frac{1}{n} \sum_{j=1}^n Z_j^2}_{\rightarrow E[Z^2] = 1} \rightarrow t$$

$$\rightarrow E[Z^2] = 1$$

$$\int_0^t W_s dW_s = \frac{W_t^2}{2} - \frac{t}{2}$$

$$W_t^2 = \int_0^t ds + 2 \int_0^t W_s dW_s$$

$$\text{Illy } W_t^3 = 3 \int_0^t W(s) ds + 3 \int_0^t W_s^2 dW_s$$

}

smooth h

 $x_T^2 = 2 \int_0^T x_s dx_s$ 
 $x_T^3 = 3 \int_0^T x_s^2 dx_s$

Ito formula:

$\{X_t\}_{t \in [0, T]}$  is an Ito process (on a filtered prob. space  $(\Omega, \mathcal{F}, \mathbb{F}, P)$  with a BM  $\{W_t\}_{t \in [0, T]}$  on it) if

$$X_t = X_0 + \int_0^t a_s ds + \int_0^t b_s dW_s, \quad t \in [0, T]$$

— ①

where  $\rightarrow X_0$  is  $\mathcal{F}_0$  mble

$\rightarrow \{a_t = a_t(\omega)\}_{t \in [0, T]}$  is an  $\mathbb{F}$ -adapted process with mble trajectory s.t.  $\int_0^T |a_s| ds < \infty$

$\rightarrow \{b_t = b_t(\omega)\}_{t \in [0, T]}$  is an  $\mathbb{F}$ -adapted process s.t.

$$\int_0^T E(b_s^2) ds < \infty$$

— x —

① has on  $[0, T]$  stochastic differentials

$$dX_t = a_t dt + b_t dW_t \quad \text{--- ②}$$

multiplication table

$$dt \cdot dt = (dt)^2 = 0$$

$$dt dW_t = dW_t dt = 0$$

$$\sum (\Delta t_k)^2 \rightarrow 0$$

$$\sum (\Delta t_k)(\Delta W_k) \rightarrow 0$$

$$dW_t dW_t = (dW_t)^2 = dt$$

$$\sum (\Delta W_k)^2 \rightarrow t$$

Let  $\{X_t\}$  Itô process with stoch differentials ②,  $f$  twice cont differentiable  
 , let  $Y_t = f(X_t)$ ,  $t \in [0, T]$

Using Taylor's formula

$$dY_t \equiv df(X_t) = f'(X_t) dX_t + \frac{1}{2} f''(X_t) \underbrace{(dX_t)^2}_{\substack{= \\ b_t^2 dt}}$$

From ②

$$dX_t = a_t dt + b_t dW_t$$

$$\begin{aligned} (dX_t)^2 &= (dX_t)(dX_t) = (a_t dt + b_t dW_t)(a_t dt + b_t dW_t) \\ &= \underbrace{a_t^2 (dt)^2}_{=0} + 2a_t b_t \underbrace{dW_t dt}_{=0} + \underbrace{b_t^2 (dW_t)^2}_{dt} \\ &= b_t^2 dt \end{aligned}$$

$$\begin{aligned} dY_t \equiv df(X_t) &= \left( a_t f'(X_t) + \frac{1}{2} f''(X_t) b_t^2 \right) dt \\ &\quad + b_t f'(X_t) dW_t \end{aligned}$$

Example ①  $dY_t$ ,  $Y_t = \frac{1}{2} W_t^2$

$$\begin{aligned} dY_t \equiv d\left(\frac{1}{2} W_t^2\right) &= f'(W_t) dW_t + \frac{1}{2} f''(W_t) (dW_t)^2 \\ &= d(f(W_t)) \end{aligned}$$

$$f(x) = \frac{1}{2} x^2, f'(x) = x, f''(x) = 1$$

$$= W_t dW_t + \frac{1}{2} dt$$

$$d\left(\frac{1}{2} W_t^2\right) = W_t dW_t + \frac{1}{2} dt$$

②  $d(e^{W_t})$

$$f(x) = e^x \quad f'(x) = e^x \quad f''(x) = e^x$$

$$f(w) = 0, f'(w) = 0, f''(w) = 0$$

$$df(w_t) = f'(w_t)dw_t + \frac{1}{2} f''(w_t)dw_t^2$$

$$d(e^{w_t}) = e^{w_t}dw_t + \frac{1}{2} e^{w_t}dt$$

Statement:

$f(t, x)$  has cont. partial derivatives  $\partial_t f$ ,  $\partial_x f$ , and be twice cont. differentiable in  $x$   $\partial_{xx} f$ ;

$$(X_t) \text{ Ito process } dX_t = a_t dt + b_t dw_t$$

Then  $Y_t := f(t, X_t)$  is also an Ito process with

$$dY_t = \partial_t f(t, X_t)dt + \partial_x f(t, X_t)dX_t + \frac{1}{2} \partial_{xx} f(t, X_t) \underbrace{(dX_t)^2}_{\stackrel{//}{b_t^2 dt}}$$

Taylor series  $f$  for  $(x, y)$  near  $(a, b)$

$$f(x, y) = f(a, b) + \underbrace{f_x(a, b)(x-a)} + f_y(a, b)(y-b) + \frac{f_{xx}(a, b)}{2}(x-a)^2 + f_{xy}(a, b)(x-a)(y-b) + \frac{f_{yy}(a, b)}{2}(y-b)^2$$

Example Geometric BM

$$Z_t = e^{\mu t + \sigma W_t}, \quad \mu, \sigma \in \mathbb{R}$$

$$= f(t, W_t)$$

$$\dots \dots \mu t + \sigma x$$

$$f(t, x) = e^x$$

$$\partial_t f = \mu f(t, x)$$

$$\partial_x f = \sigma f(t, x) \quad , \quad \partial_{xx} f = \sigma^2 f(t, x)$$

$$d(Z_t) = ?$$

$$d(Z_t) = d(f(t, W_t))$$

$$= \partial_t f(t, W_t) dt + \partial_x f(t, W_t) dW_t + \frac{1}{2} \partial_{xx} f(t, W_t) (dW_t)^2$$

$$= \mu f(t, W_t) dt + \sigma f(t, W_t) dW_t + \frac{1}{2} \sigma^2 f(t, W_t) dt$$

$$= \mu Z_t dt + \sigma Z_t dW_t + \frac{1}{2} \sigma^2 Z_t dt$$

$$= \left( \mu + \frac{\sigma^2}{2} \right) Z_t dt + \sigma Z_t dW_t .$$

Product rule

$X_t, Y_t$  Itô process on common filtered prob. space

$$Z_t := X_t Y_t$$

$$dZ_t \equiv d(X_t Y_t) = X_t dY_t + Y_t dX_t + dX_t dY_t$$

~~Quotient rule~~ —  $X$  —

$$dZ_t = d\left(\frac{X_t}{Y_t}\right)$$

$$= \frac{dX_t}{Y_t} - \frac{X_t}{Y_t^2} dY_t + \frac{X_t}{Y_t^3} (dY_t)^2$$

$$- \frac{1}{Y_t^2} dX_t dY_t$$

$$\frac{y^2}{x}$$

Example

$$Z_t := W_t e^{W_t} = X_t Y_t$$

$$X_t = W_t$$

$$Y_t = e^{W_t}$$

$$dZ_t = X_t dY_t + Y_t dX_t + dX_t dY_t$$

$$= W_t d(e^{W_t}) + e^{W_t} dW_t + dW_t d(e^{W_t})$$

$$= W_t \left( e^{W_t} dW_t + \frac{1}{2} e^{W_t} dt \right) + e^{W_t} dW_t + dW_t \left( e^{W_t} dW_t + \frac{1}{2} e^{W_t} dt \right)$$

$$\therefore d(e^{W_t})$$

$$= e^{W_t} dW_t + \frac{1}{2} e^{W_t} (dW_t)^2$$

$$= e^{W_t} dW_t + \frac{1}{2} e^{W_t} dt$$

$$f(t, x) = e^x$$

$$\partial_t f = 0, \partial_x f = e^x, \partial_{xx} f = e^x$$

$$= \left( \frac{1}{2} W_t e^{W_t} + e^{W_t} \right) dt + (W_t e^{W_t} + e^{W_t}) dW_t$$

$$= \left( 1 + \frac{1}{2} W_t \right) e^{W_t} dt + (1 + W_t) e^{W_t} dW_t$$