

$$df(t, B_t) = \boxed{f'(t, B_t) dB_t} + \left\{ f + \frac{1}{2} f'' \right\}(t, B_t) dt$$

Extending the stochastic Integral

We've seen how Itô's formula works for  $C^3$  bounded functions of  $(t, B_t)$ , but we really only need  $f$  to be  $C^{1,2}$ . Then the formula makes sense. But there's a catch...

The term  $\int_0^t f'(s, B_s) dB_s$  is generally not a martingale only a local martingale.

A process  $M$  is a local martingale if there exists

stopping times  $T_n \uparrow \infty$  such that  $M_t^{T_n} = M_{t \wedge T_n}$  is a mart.

For example,  $X_t \equiv \int_0^t \exp(B_s^4) dB_s$  is a local martingale, but not a martingale; we could take

$$T_n = n \wedge \min \{ t : |B_t| \geq n \}$$

Then when we take  $X(t \wedge T_n) = \int_0^{t \wedge T_n} \exp(B_s^4) dB_s$

the integrand  $\exp(B_s^4) \mathbb{I}_{\{s \leq T_n\}}$  is bounded, and zero off the time interval  $[0, n]$ .

So the integrand is  $e^{n^4} \bar{f}$ , and the stochastic integral is an  $L^2$  bounded martingale.

But without localisation

$$\mathbb{E} \left[ \int_0^\infty e^{2B_s^4} ds \right] = +\infty$$

So the original process has no nice integrability properties.

This frees us from the need to constantly check the integrability of the integrand.

Another extension is to multidimensional B.M.

$$B_t = (B_t^1, \dots, B_t^n)$$

where the  $B^j$  are independent standard Brownian Motions.

The rules of the calculus are

$$dt^2 = 0 \quad dt dB_t = 0 \quad dB_t^i dB_t^j = \delta_{ij} dt$$

and now

$$d(f(t, B_t)) = f dt + \sum_{i=1}^n (D_i f)(t, B_t) dB_t^i + \frac{1}{2} \sum_{i,j=1}^n D_{ij} f(t, B_t) dB_t^i dB_t^j$$

$$= f dt + (D_i f)(t, B_t) dB_t^i + \frac{1}{2} \Delta f(t, B_t) dt$$

Summation convention

Next extension

If  $X_t = x + \int_0^t H_s dB_s + \int_0^t N_s ds$  then

~~$$d(f(t, X_t)) = f dt + f'(t, X_t) dx_t + \frac{1}{2} f''(t, X_t) dx_t dx_t$$~~

$$= f dt + f'(t, X_t) (H_t dB_t + N_t dt) + \frac{1}{2} f''(t, X_t) H_t^2 dt$$

Multivariate

$$dx^i = H^{ij} dB^j + N^i dt$$

→ locally in  $\mathbb{R}^n(\bar{\Omega}, \text{Leb} \times \mathbb{R})$

→ adapted and pathwise integrable  
 $t \mapsto N_t(\omega)$

then

$$d(f(t, X_t)) = f dt + (D_i f)(t, X_t) dx_t^i + \frac{1}{2} (D_{ij} f)(t, X_t) dx_t^i dx_t^j$$

dropping & subscripts

$$= f dt + D_i f(t, X) \{ H^{ik} dB^k + N^i dt \} + \frac{1}{2} (D_{ij} f)(t, X_t)$$

$$[dx^i dx^j = H^{jl} dB^l H^{ik} dB^k = H^{jl} H^{ik} dt]$$

Special case  $f(x,y) = xy$

(Integration by parts formula)

$$\text{Suppose } dX = H dB + N dt$$

$$dY = G dB + v dt$$

$$d(XY) = Y dX + X dY + 1 \cdot dx dy$$

$$= Y dX + X dY + HG dt$$

e.g.

$$dX^2 = 2X dX + dX dX$$

Stochastic differential Equations

What could we do with S.d.e.'s;

$$dX_t^i = \sigma^i(t, X_t) dB_t^j + N^i(t, X_t) dt$$

$$X_t = x_0$$

"Lipshitz + Picard Iteration"

Just as in classical ODE theory, this SDE will have a unique solution if for, each  $T > 0$  there exists  $C_T < \infty$  such that

$$|\sigma(t, x) - \sigma(t, y)| + |N(t, x) - N(t, y)| \leq C_T |x - y|$$

$$|\sigma(t, 0)| + |N(t, 0)| \leq C_T \quad \forall x, y \in \mathbb{R}^n$$

$$\forall t \in [0, T]$$

The proof is constructive; start with  $X_t^{(0)} \equiv x_0$

and define recursively

$$X_t^{(n+1)} = x_0 + \int_0^t \sigma(s, X_s^{(n)}) dB_s$$

$$+ \int_0^t N(s, X_s^{(n)}) ds$$

In one dimension, we can do better; we still require  $\mu$  to be Lipschitz, but we can find a unique solution

provided

$$|\sigma(t, x) - \sigma(t, y)| \leq C_T |x - y|^\alpha$$

for all  $x, y, \forall t \in [0, T]$ , where  $\alpha \in [\frac{1}{2}, 1]$  is fixed

(Yamada - Watanabe; Le Gall).