

# Lecture 3. Itô's formula

— with 3 tables.

• Goal: The "chain rule" for stochastic calculus.

• Fundamental calculus: we apply.

$$\int f(x) dx = \int \frac{d}{dx} F(x) dx = F(x).$$

• Does it work with Itô?  $df(B_t) \stackrel{?}{=} \left( \frac{dB_t}{dt}, f'(B_t) \right) dt$ . → doesn't make sense!

• But! BMs has rough paths, nowhere differentiable!

• Instead we look at smt that at least makes sense:

$$df = f'(B_t) dB_t + \frac{1}{2} f''(B_t) (dB_t)^2 + \frac{1}{6} f'''(B_t) (dB_t)^3 + \dots$$

→ How many terms do we need?

Difficult part: BM. If we solve the problem for a BM, we can solve it for many other processes.

(Step 1. BM.)

Thm 3.1 (Itô's formula for BM).

Let  $f \in C^2(\mathbb{R})$  (cont. twice differentiable) and  $B_t$  be a standard BM

For any  $t > 0$ .

$$df(B_t) = f'(B_t) dB_t + \frac{1}{2} f''(B_t) dt$$

→ an SDE.

Or in integral form,

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

→ s.p.

Prf of thm 3.1 w.l.o.g. assume  $f, f', f''$  bdd.

(On top of this we can approximate any fn and pass to the limit)

Take a partition with  $N$  equal intervals:

$$f(B_t) - f(0) \stackrel{(*)}{=} \sum_{k=1}^N \underbrace{(f(B_{t_k}) - f(B_{t_{k-1}}))}_{\text{expand}}$$

By Taylor,

$$f(B_{t_k}) - f(B_{t_{k-1}}) = f'(B_{t_{k-1}})(B_{t_k} - B_{t_{k-1}}) + \frac{1}{2} f''(B_{t_{k-1}})(B_{t_k} - B_{t_{k-1}})^2.$$

Then we can write  $(*)$  as

$$f(B_t) - f(0) = \sum_{k=1}^N f'(B_{t_{k-1}})(B_{t_k} - B_{t_{k-1}}) + \sum_{k=1}^N \frac{1}{2} f''(B_{t_{k-1}})(B_{t_k} - B_{t_{k-1}})^2.$$

By the construction of Itô's integral,

$$\sum_{k=1}^N f'(B_{t_{k-1}})(B_{t_k} - B_{t_{k-1}}) \xrightarrow{L^2} \int_0^t f'(B_s) dB_s.$$

WTS:

$$\sum_{k=1}^N f''(B_{t_{k-1}})(B_{t_k} - B_{t_{k-1}})^2 \xrightarrow{L^2} \int_0^t f''(B_s) ds.$$

$$\Leftrightarrow \sum_{k=1}^N f''(B_{t_{k-1}}) \left[ (B_{t_k} - B_{t_{k-1}})^2 - (t_k - t_{k-1}) \right] \xrightarrow{L^2} 0$$

als. continuous trajectories.

Idea:  $(dB_t)^2 = dt$

$$\int_0^T (dB_t)^2 = \lim_{N \rightarrow \infty} \sum_{k=1}^N (\Delta B_{t_{k-1}})^2 \stackrel{L^2}{=} T = \int_0^T dt$$

$$\mathbb{E} \left[ \sum_{k=1}^N (\Delta B_{t_{k-1}})^2 \right] \stackrel{\text{ind.}}{=} \sum_{k=1}^N \mathbb{E} [(\Delta B_{t_{k-1}})^2] = \sum_{k=1}^N \Delta t_{k-1} = T.$$

$$\begin{aligned} \mathbb{E} \left[ \left( \sum_{k=1}^N (\Delta B_{t_{k-1}})^2 - T \right)^2 \right] &= \text{Var} \left( \sum_{k=1}^N (\Delta B_{t_{k-1}})^2 - T \right) \\ &= \sum_{k=1}^N \text{Var} \left[ (\Delta B_{t_{k-1}})^2 \right] \\ &= \sum_{k=1}^N \underbrace{\mathbb{E} [(\Delta B_{t_{k-1}})^4]}_{3(\Delta t)^2} - \underbrace{\mathbb{E} [(\Delta B_{t_{k-1}})^2]^2}_{(\Delta t)^2} \\ &= 2 \left( \frac{T}{N} \right)^2 \cdot N = \frac{2T^2}{N} \rightarrow 0. \quad \square \end{aligned}$$

Corol 3.2:  $f(t, x) \in C^{1,2}([0, \infty) \times \mathbb{R})$ . Let  $Y_t = f(t, B_t)$ , then

$$dY_t = \left( f_t + \frac{1}{2} f_{xx} \right) dt + f_x dB_t$$

Remark:  $(dB_t)^2$  is computed according to the following table:

	$dt$	$dB_t$
$dt$	0	0
$dB_t$	0	$dt$

(Drop terms smaller than  $dt$ .)

Why can we do this? Well, let  $p > 0$ .

$$\int_0^T (dt)^p = \lim_{N \rightarrow \infty} \sum_{k=1}^N (\Delta t)^p$$

$$= \lim_{N \rightarrow \infty} N \cdot \left(\frac{T}{N}\right)^p$$

$$\xrightarrow{N \rightarrow \infty} 0 \quad \text{if } p > 1.$$

• Choose ansatz.

• Apply Ito

• Integrate

Some examples with BM: Solve Ito integrals

E.X. 3.2  $I = \int_0^t B_s dB_s$  (Exercise 2.2)

Let  $f(t, x) = \frac{1}{2} x^2$  (why? from calculus).

$$Y_t := f(t, B_t) = \frac{1}{2} B_t^2, \quad Y_0 = 0$$

By Ito:

$$dY_t = B_t dB_t + \frac{1}{2} dt$$

$$\frac{1}{2} B_t^2 = Y_t = Y_0 + \int_0^t B_s dB_s + \int_0^t \frac{1}{2} ds$$

$$= \int_0^t B_s dB_s + \frac{1}{2} t$$

$$\int_0^t B_s dB_s = \frac{1}{2} (B_t^2 - t).$$



Ex. 3.3  $I = \int_0^t s dB_s$ . (Exercise 2.4)

Let  $f(t, x) = tx$ . (Why?)

$$Y_t = t B_t, \quad Y_0 = 0$$

$$dY_t = B_t dt + t dB_t + 0 \cdot (dB_t)^2$$

$$t B_t = \int_0^t B_s ds + \int_0^t s dB_s$$

$$\Rightarrow \int_0^t s dB_s = t B_t - \int_0^t B_s ds$$

what's this guy?  
r.v.

↓  
matches integration by parts.

Remark.  $\int_0^t s dB_s \sim N(0, \frac{1}{3} t^3)$ . (by exercise.)

$\int_0^t B_s ds$  is also Gaussian.

$$E\left[\int_0^t B_s ds\right] = \int_0^t E[B_s] ds = 0$$

$$\begin{aligned} \text{Var}\left[\int_0^t B_s ds\right] &= E\left[\left(\int_0^t B_s ds\right)^2\right] \\ &= E\left[\int_0^t \int_0^t B_s B_u ds du\right] \\ &= \int_0^t \int_0^t E[B_s B_u] ds du \\ &= \int_0^t \int_0^t \min(s, u) ds du \\ &= \int_0^t \left(\int_0^u s ds + \int_u^t u ds\right) du \end{aligned}$$

$$= \frac{1}{3} t^3.$$

$$\int_0^t B_s ds \sim N\left(0, \frac{1}{3} t^3\right)$$

Exercise 1.  $\int_0^t B_s^2 dB_s$ .

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(Step 2: 1d Ito process).

Thm 3.4 Let  $X_t$  be an Ito process with

$$dX_t = \mu_t dt + \sigma_t dB_t,$$

Let  $f(t, x) \in C^{1,2}([0, \infty), \mathbb{R})$ , Define  $Y_t = f(t, X_t)$ .

Then.

$$\begin{aligned} dY_t &= \underbrace{f_t(t, X_t)} dt + f_x(t, X_t) dX_t + \frac{1}{2} f_{xx}(t, X_t) (dX_t)^2 \\ &= f_t dt + f_x (\mu_t dt + \sigma_t dB_t) + \frac{1}{2} f_{xx} (\mu_t dt + \sigma_t dB_t)^2 \end{aligned}$$

plug in  $dX_t$  →

$$= (f_t + f_x \mu_t + \frac{1}{2} f_{xx} \sigma_t^2) dt + f_x \sigma_t dB_t.$$

## E.x. 3.5 · ( Financial models )

i)  $dX = \mu dt + \sigma dB_t$  (Bachelier, 1900)

"normal model". e.g. In 2020, oil price.

ii)  $dX = \mu X dt + \sigma X dB_t$  (Geometric BM, Black-Scholes-Merton, 1973)

"lognormal". modern approach.

iii)  $dX = \beta X dt + \sigma dB_t$  (Ornstein-Uhlenbeck).

$\downarrow$   
 $< 0$

"mean reversion". e.g. Vasicek model.

Solution to ii) with Ito:

Let  $Y_t = \log(X_t)$ .

$$dY_t = \frac{1}{X_t} dX_t + \frac{1}{2} \left(-\frac{1}{X_t^2}\right) (dX_t)^2$$

$$= \mu dt + \sigma dB_t + \left(-\frac{1}{2}\sigma^2 dt\right)$$

$$= \left(\mu - \frac{1}{2}\sigma^2\right) dt + \sigma dB_t$$

$$Y_t = Y_0 + \int_0^t \left(\mu - \frac{1}{2}\sigma^2\right) ds + \int_0^t \sigma dB_s$$

$$= \log(X_0) + \left(\mu - \frac{1}{2}\sigma^2\right)t + \sigma B_t$$

$$X_t = \exp(Y_t) = X_0 \exp\left\{\left(\mu - \frac{1}{2}\sigma^2\right)t + \sigma B_t\right\}.$$

Exercise = solve iii)

Ex. 3.6. Calculate  $\mathbb{E}[e^{\alpha W_t}]$ ,  $\alpha \in \mathbb{R}$ .

• By prob.,  $X \sim \text{lognormal}(\mu, \sigma^2)$ ,  $\mathbb{E}[X] = e^{\mu + \frac{1}{2}\sigma^2}$ .

$$\Rightarrow \mathbb{E}[e^{\alpha W_t}] = \exp\left(0 + \frac{1}{2}\alpha^2 t\right) = e^{\frac{1}{2}\alpha^2 t}.$$

• By Itô, let  $Y = e^{\alpha W}$ ,  $Y_0 = 1$

$$dY = \alpha e^{\alpha W} dW + \frac{1}{2}\alpha^2 e^{\alpha W} dt$$

$$= \alpha Y dW + \frac{1}{2}\alpha^2 Y dt$$

$$Y_t = Y_0 + \frac{1}{2}\alpha^2 \int_0^t Y_s ds + \alpha \int_0^t Y_s dW_s$$

$$M_t := \mathbb{E}[Y_t] = Y_0 + \frac{1}{2}\alpha^2 \int_0^t \mathbb{E}[Y_s] ds + 0$$

$$dM_t = \frac{1}{2}\alpha^2 M_t dt, \quad M_0 = Y_0 = 1$$

$$M_t = M_0 \exp\left(\frac{1}{2}\alpha^2 t\right) = \exp\left(\frac{1}{2}\alpha^2 t\right).$$

Exercise: Calculate  $\mathbb{E}[X_t]$  where  $X_t$  is a GBM.  $(\mathbb{E}[X_t] = x_0 e^{\mu t})$ .

(Step 3, n-dim Itô).

Recall:  $A = [a_{ij}]$ ,  $B = [b_{ij}]$ ,  $m \times n$  matrices. We define their

Frobenius product:

$$\langle A, B \rangle = \sum_{ij} a_{ij} b_{ij} = \text{Tr}(AB^T)$$

Thm 3.7 Let  $f(t, x) \in C^{1,2}$ , we write

$$f_x = \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{pmatrix}, \quad f_{xx} = \begin{bmatrix} \frac{\partial^2 f}{\partial x_i \partial x_j} \end{bmatrix}$$

Let  $X_t$  be an  $n$ -dim Itô process, i.e.

$$dX = \underbrace{\mu}_{n \times 1} dt + \underbrace{\sigma}_{n \times 1} dB_t$$

$\uparrow$   $\uparrow$   $\uparrow$   $\uparrow$   
 $n \times 1$   $n \times 1$   $n \times d$   $d \times 1, \text{ ind!}$

Let  $Y_t = f(t, X_t)$ ,

$$dY_t = f_t + \sum_{i=1}^n \frac{\partial f}{\partial x_i} dX_i + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j} dX_i dX_j$$

$$= \left( f_t + \langle \mu, f_x \rangle + \frac{1}{2} \langle \sigma \sigma^T, f_{xx} \rangle \right) dt + \langle f_x, \sigma dB \rangle.$$

Check!  $\rightarrow$

at  $(t, X_t)$ .

In other words:

	$dt$	$dB_i$	$dB_j$
$dt$	0	0	0
$dB_i$	0	$dt$	0
$dB_j$	0	0	$dt$

Ex. 3.8 (Important).

Denote the Euclidean norm by  $\|\cdot\|$ , and let  $B_t$  be an  $n$ -dim

BM starting from  $X_0 \in \mathbb{R}^n$ . Let  $f(x) = \|x\|^2$ .

$$f(B_t) = B_1^2(t) + \dots + B_n^2(t).$$

$$f_x(B_t) = 2B_t, \quad f_{xx} = 2I_n$$

$$\begin{aligned} df(B_t) &= (0 + 0 + \frac{1}{2} \langle I_n, 2I_n \rangle) dt + \langle 2B_t, dB_t \rangle \\ &= n dt + \langle 2B_t, dB_t \rangle. \end{aligned}$$

Or, in integral form,

$$\|B_t\|^2 = \|x_0\|^2 + nt + 2 \sum_{i=1}^n \int_0^t B_i(s) dB_i(s).$$

(Step 4.  $n$ -dim correlated  $I_{t_0}$ !)

$\hookrightarrow$  BM's are correlated.

Suppose the components of BMs are not independent, such that for  $s < t$ ,

$$\begin{aligned} \text{Corr}(B_i(t) - B_i(s), B_j(t) - B_j(s)) &= \frac{1}{t-s} \mathbb{E}[(B_i(t) - B_i(s))(B_j(t) - B_j(s))] \\ &= \rho_{ij} \end{aligned}$$

Denote  $P = [\rho_{ij}]$  as the correlation matrix.

Thm 3.9.  $I_{t_0}$ 's lemma still holds, but with  $\sigma\sigma^T$  changed to  $\sigma P \sigma^T$ .

i.e. the following table:

	$dt$	$dB_i$	$dB_j$
$dt$	0	0	0
$dB_i$	0	$dt$	$\rho_{ij} dt$
$dB_j$	0	$\rho_{ij} dt$	$dt$

## Application

Thm 3.10 (Martingale representation thm).

Let  $M_t$  be an  $\mathcal{F}_t^B$ -martingale such that  $M_T \in L^2$ . Then there exists a unique  $\mathcal{F}_t^B$ -adapted process  $\{H_t\}_{t \leq T} \in L^2$

$$M_t = M_0 + \int_0^t H_s dB_s \quad \text{a.s. for all } t.$$

$\hookrightarrow$  deterministic.

Remark: We know "an Ito integral is a m.g.". This tells us the converse: "Given a m.g., it can be written as an Ito integral."

Why is this important?  $\rightarrow$  application to finance.

$$\underbrace{\Phi_t}_{\text{Contract value}} = \Phi_0 + \int_0^t \underbrace{H_s}_{\text{hedging strategy}} dB_s \quad \hookrightarrow \text{stock price.}$$

$H_t$  is the amount of stock at the beginning of each infinitesimal period.

How to find  $H$ ? e.g. Ito!

In the simple case where  $M_t = f(t, B_t)$ , by Ito,

$$M_t = M_0 + \int_0^t (f_s + \frac{1}{2} f_{xx}) (s, B_s) ds + \int_0^t f_x (s, B_s) dB_s.$$

It has to be that

$$f_t + \frac{1}{2} f_{xx} = 0.$$

H.E.

and  $H_t(w) = f_t(t, B_t)(w)$ .

process

↳ derivative.

e.g.  $M_t = B_t^2 - t$ .

m.g. rep.  $\Rightarrow \int_0^t 2B_s dB_s$ .

$M_t = \exp(\pi B_t - \frac{1}{2} \pi^2 t)$  → what's this?

m.g. rep.  $\Rightarrow \pi \int_0^t e^{\pi B_s - \frac{1}{2} \pi^2 s} dB_s$ .

(Wald m.g. / GBM)

$= \int_0^t \pi M_s dB_s$ .

Exercise. Find  $f(x)$  s.t.

$M_t = B_t^3 + \int_0^t f(B_s) dB_s$ . B a m.g.