

Re cap :

— Itô's integration

Define $\int_0^T X_t dB_t$ as L^2 limit of

$$\sum_{i=0}^{N-1} X_{i\Delta t} \cdot (B_{(i+1)\Delta t} - B_{i\Delta t})$$

a martingale.

— Itô's formula.

$$\left. \begin{aligned} df(B_t) &= f'(B_t) dB_t + \frac{1}{2} f''(B_t) dt \\ f(B_T) &= f(B_0) + \int_0^T f'(B_t) dB_t + \frac{1}{2} \int_0^T f''(B_t) dt \end{aligned} \right\} \begin{matrix} \text{"differential form"} \\ \text{"integration form"} \end{matrix}$$

(Remark: differential form is just a shorthand notation for the integration form).

Next step: generalizations

- $f(t, B_t)$, $f(t, Z_t)$
- $dZ_t = X_t dt + Y_t dB_t$, study $f(Z_t)$
- Multivariate extension.

Suppose a continuous-time process $(Z_t)_{t \geq 0}$

satisfies $dZ_t = X_t dt + Y_t dB_t$

(for any $T > 0$, $Z_T - Z_0 = \int_0^T X_t dt + \int_0^T Y_t dB_t$)

This covers a lot of important processes.

e.g. $Z_t = f(B_t)$, then by Itô's formula,

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

e.g. Stochastic differential equations.

$$dZ_t = -Z_t dt + dB_t$$

(in general) $dZ_t = -\nabla u(Z_t) dt + dB_t$.

As before, we Taylor-expand f . ($N = \frac{T}{\Delta t}$).

$$\begin{aligned} f(Z_T) - f(Z_0) &= \sum_{i=0}^{N-1} \left(f(Z_{(i+1)\Delta t}) - f(Z_{i\Delta t}) \right) \\ &= \sum_{i=0}^{N-1} \left[\underbrace{f'(Z_{i\Delta t}) \cdot (Z_{(i+1)\Delta t} - Z_{i\Delta t})}_{①} \right] \\ &\quad \left[+ \frac{1}{2} f''(Z_{i\Delta t}) \cdot (Z_{(i+1)\Delta t} - Z_{i\Delta t})^2 \right] \\ &\quad \left[+ o(|Z_{(i+1)\Delta t} - Z_{i\Delta t}|^2) \right] \end{aligned}$$

Since $Z_{(i+1)\Delta t} - Z_{i\Delta t} = \int_0^{\Delta t} X_{i\Delta t+s} ds + \int_0^{\Delta t} Y_{i\Delta t+s} dB_s$

$$\mathbb{E}[(Z_{(i+1)\Delta t} - Z_{i\Delta t})^2] \leq \mathbb{E}\left[\left(\int_0^{i\Delta t} X_s ds\right)^2\right] + 2 \int_0^{i\Delta t} \mathbb{E}[Y_s^2] ds$$

$$\text{So, } o(|Z_{(i+1)\Delta t} - Z_{i\Delta t}|^2) = o(\Delta t).$$

$$\sum_{i=0}^{N-1} o(\Delta t) = o(N \cdot \Delta t) = o(1).$$

So the residual $\rightarrow 0$.

Term ①: Riemann integral + Ito's integral.

$$\lim_{N \rightarrow +\infty} \sum_{i=0}^{N-1} f'(Z_{i\Delta t}) \cdot \underbrace{\left[\int_{i\Delta t}^{(i+1)\Delta t} X_t dt + \int_{i\Delta t}^{(i+1)\Delta t} Y_t dB_t \right]}_{\downarrow}$$

$$= \int_0^T f'(Z_t) \cdot X_t dt + \int_0^T f'(Z_t) Y_t dB_t$$

So Term 1 $\rightarrow \int_0^T f'(Z_t) \cdot (X_t dt + Y_t dB_t) = \int_0^T f'(Z_t) dZ_t$

Term 2:

$$\begin{aligned} & \lim_{N \rightarrow +\infty} \sum_{i=0}^{N-1} f''(Z_{i\Delta t}) \cdot (Z_{(i+1)\Delta t} - Z_{i\Delta t})^2 \\ &= \int_0^T f''(Z_t) Y_t^2 dt. \end{aligned}$$

As before, define quadratic variation process.

$$\langle Z \rangle_t := \lim_{n \rightarrow +\infty} \sum_{i=0}^{n-1} \left(Z_{\frac{(i+1)t}{n}} - Z_{\frac{it}{n}} \right)^2.$$

(Note: notation change from last time,
we'll use $\langle \cdot \rangle_t$ throughout the rest of course)

(increasing, and cts in t).

Using the piecewise const approximation, one can show

$$\lim_{n \rightarrow +\infty} \sum_{i=0}^{n-1} f''(Z_{i\delta t}) (Z_{(i+1)\delta t} - Z_{i\delta t})^2 = \int_0^T f''(Z_t) d\langle Z \rangle_t$$

Need to show $d\langle Z \rangle_t = Y_t^2 dt$.

Indeed:

$$Z_{(i+1)\delta t} - Z_{i\delta t} = \underbrace{\int_0^{\delta t} X_{i\delta t+s} ds}_{O(\delta t)} + \underbrace{\int_0^{\delta t} Y_{i\delta t+s} dB_s}_{O(\sqrt{\delta t})}.$$

$$(Z_{(i+1)\delta t} - Z_{i\delta t})^2 = \left(\int_0^{\delta t} Y_{i\delta t+s} dB_s \right)^2 + O(\delta t^2) + O(\delta t^{3/2})$$

$$\sum_{i=0}^{N-1} (Z_{(i+1)\delta t} - Z_{i\delta t})^2 = \underbrace{\sum_{i=0}^{N-1} \left(\int_0^{\delta t} Y_{i\delta t+s} dB_s \right)^2}_{\int_0^t Y_s^2 ds} + O(N \cdot \delta t^{3/2}) \xrightarrow{\delta t \rightarrow 0}$$

since $\delta t = \frac{T}{N}$

Conclusion:

$$\begin{aligned} df(Z_t) &= f'(Z_t) dZ_t + \frac{1}{2} f''(Z_t) d\langle Z \rangle_t \\ &= f'(Z_t) X_t dt + f'(Z_t) Y_t dB_t + \frac{1}{2} f''(Z_t) Y_t^2 dt. \end{aligned}$$

e.g. $f(x) = x^2$

$$\begin{aligned} d(Z^2) &= 2Z_t dZ_t + d\langle Z \rangle_t \\ &= 2Z_t(X_t dt + Y_t dB_t) + Y_t^2 dt. \end{aligned}$$

Suppose that $dZ_t^{(i)} = X_t^{(i)} dt + Y_t^{(i)} dB_t$
for $i \in \{1, 2\}$.

We can apply formula for $d(Z^2)$

to $\frac{Z_t^{(1)} + Z_t^{(2)}}{2}$ and $\frac{Z_t^{(1)} - Z_t^{(2)}}{2}$

$$\begin{aligned} d(Z_t^{(1)} \cdot Z_t^{(2)}) &= d\left(\left(\frac{Z_t^{(1)} + Z_t^{(2)}}{2}\right)^2\right) - d\left(\left(\frac{Z_t^{(1)} - Z_t^{(2)}}{2}\right)^2\right) \\ &= Z_t^{(1)} dZ_t^{(2)} + Z_t^{(2)} dZ_t^{(1)} + Y_t^{(1)} Y_t^{(2)} dt. \end{aligned}$$

"polarization trick".

$$\int_0^T Y_s^{(1)} Y_s^{(2)} ds = \lim_{n \rightarrow \infty} \sum_{i=0}^{n-1} \left(Z_{\frac{(i+1)t}{n}}^{(1)} - Z_{\frac{it}{n}}^{(1)} \right) \cdot \left(Z_{\frac{(i+1)t}{n}}^{(2)} - Z_{\frac{it}{n}}^{(2)} \right)$$

$$\langle \bar{z}^{(1)}, \bar{z}^{(2)} \rangle_t := \lim \left[\text{---} \right]$$

Following this notation, we have product rule.

$$d(\bar{z}_t^{(1)} \cdot \bar{z}_t^{(2)}) = \bar{z}_t^{(2)} d\bar{z}_t^{(1)} + \bar{z}_t^{(1)} d\bar{z}_t^{(2)} + d\langle \bar{z}^{(1)}, \bar{z}^{(2)} \rangle_t.$$

How about $f(t, \bar{z}_t)$?

$$f(t, \bar{z}_t) - f(0, \bar{z}_0) = \sum_{i=0}^{N-1} \underbrace{f((i+1)\delta t, \bar{z}_{(i+1)\delta t}) - f(i\delta t, \bar{z}_{i\delta t})}_{\text{---}}$$

$$f(i\delta t, \bar{z}_{(i+1)\delta t}) - f(i\delta t, \bar{z}_{i\delta t})$$

*Similar to time-homogeneous
case*

$$\boxed{f((i+1)\delta t, \bar{z}_{(i+1)\delta t}) - f(i\delta t, \bar{z}_{i\delta t})}$$

just the derivative

Conclusion

$f(t, x)$

$$df(t, \bar{z}_t) = \frac{\partial f}{\partial t}(t, \bar{z}_t) dt + \frac{\partial f}{\partial x}(t, \bar{z}_t) d\bar{z}_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2}(t, \bar{z}_t) \cdot d\langle \bar{z} \rangle_t.$$

e.g. $f(t, x) = \exp(at + bx)$.

$$\bar{z}_t = e^{at + bBt}$$

$$dZ_t = a \cdot e^{at+bB_t} dt + b \cdot e^{at+bB_t} dB_t \\ + \frac{b^2}{2} e^{at+dB_t} dt.$$

$$= \left(a + \frac{b^2}{2}\right) \cdot Z_t dt + b \cdot Z_t dB_t.$$

Special case: when $a = -\frac{b^2}{2}$.

$Z_t = \exp\left(-\frac{b^2 t}{2} + b B_t\right)$ is a martingale.

SDE $dZ_t = r Z_t dt + b Z_t dB_t$

Let $a = r - \frac{b^2}{2}$, and we get

the solution $Z_t = \exp(b B_t + (r - \frac{b^2}{2})t)$.

e.g. (OU process). $f(t, x) = e^t \cdot x$

$$dZ_t = -Z_t dt + dB_t, \quad Z_0 = 0.$$

$$\left(\frac{\partial^2 f}{\partial x^2} = 0\right)$$

$$d(e^t Z_t) = e^t dZ_t + e^t Z_t dt + 0$$

$$= e^t (-Z_t dt + dB_t) + e^t Z_t dt$$

$$= e^t dB_t.$$

$$S_0 \quad Z_t = e^{-t} \int_0^t e^s dB_s.$$

$$\sim \mathcal{N}\left(0, e^{-2t} \int_0^t e^{2s} ds\right).$$

Multivariate extension.

$$B_t = (B_t^{(1)}, B_t^{(2)}, \dots, B_t^{(d)})$$

where $B_t^{(1)}, B_t^{(2)}, \dots, B_t^{(d)}$ are iid BM's.

Ito's formula.

$$df(t, B_t) = \frac{\partial f}{\partial t}(t, B_t) dt + \nabla_x f(t, B_t)^T dB_t + \frac{1}{2} \Delta_x f(t, B_t) \frac{dt}{dt}$$

$$\text{where } \Delta_x f(t, x) = \sum_{i=1}^d \frac{\partial^2 f}{\partial x_i^2}(t, x).$$

$$\text{e.g. } X_0 \sim \mu \text{ (on } \mathbb{R}^d)$$

$$X_t = X_0 + B_t.$$

Let $p_t(x)$ be the density of X_t at point x .

For $s < t$,

$$p_t(x) = \int p_s(y) \cdot \phi\left(\frac{y-x}{t-s}\right) dy$$

(ϕ is d -dim

$$= \mathbb{E}[p_s(x + B_t - B_s)] \text{ Gaussian density}).$$

$$\frac{\partial P_t}{\partial t}(x) = \frac{\partial}{\partial t} \mathbb{E}[P_s(x + B_t - B_s)]$$

By Itô's formula,

$$dP_s(x + B_t - B_s) = \nabla_x P_s(x + B_t - B_s)^T dB_t + \frac{1}{2} \Delta_x P_s(x + B_t - B_s) dt$$

$$P_t(x) - P_s(x)$$

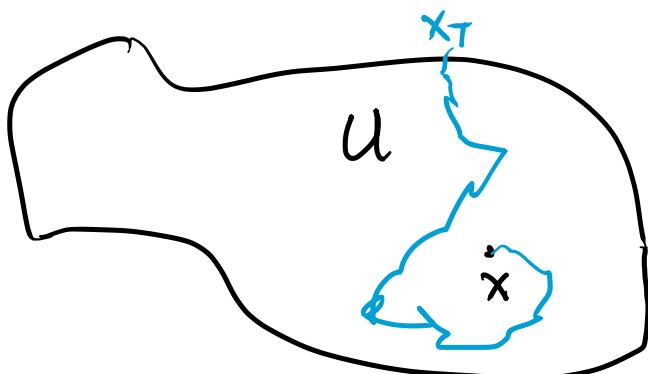
$$= \mathbb{E}[P_s(x + B_t - B_s)] - P_s(x)$$

$$= \mathbb{E}\left[\int_s^t \nabla_x P_s(x + B_r - B_s)^T dB_r + \frac{1}{2} \Delta_x P_s(x + B_r - B_s) dr\right]$$

$$= \frac{1}{2} \mathbb{E}\left[\int_s^t \Delta_x P_s(x + B_r - B_s) dr\right]$$

$$t \rightarrow s^+. \quad \frac{\partial P_t}{\partial t}(x) = \frac{1}{2} \Delta_x P_t(x).$$

e.g.



$$U \subseteq \mathbb{R}^d$$

a bounded region
w/ smooth boundary.

In PDE literature, Laplace's equation.

$$\begin{cases} \Delta f(x) = 0 & \text{for } x \in U \\ f = g & \text{at boundary } \partial U \end{cases}$$

Suppose we have a solution.
Starting from some $x \in \text{interior}(U)$.

$$X_t = x + B_t$$

$$f(X_t) = f(x) + \int_0^t \nabla f(X_s)^T dB_s + \cancel{\frac{1}{2} \int_0^t \Delta f(X_s) ds}.$$

Let $T :=$ hitting time of boundary

(stopping time, $\mathbb{P}(T < +\infty) = 1$).

Before T , $\Delta f(X_s) = 0$ (by PDE).

$$f(X_t) = f(x) + \int_0^t \nabla f(X_s)^T dB_s, \quad \text{is an M.G.}$$

Under bdd/smooth assumptions, we can apply OST.
(e.g. when f is bdd in U).

$$\mathbb{E}[f(X_T)] = f(x).$$

This gives us explicit solution to the PDE

$$f(x) = \mathbb{E}[g(x + B_T)].$$

(c.f. gambler's ruin problem in 1D).

Application of the Lap Eq. ($d \geq 2$).



$$U := \{x \in \mathbb{R}^d : R_1 \leq \|x\|_2 \leq R_2\}$$

$$g = 1 \quad \text{on } \{x : \|x\|_2 = R_2\}$$

$$g = 0 \quad \text{on } \{x : \|x\|_2 = R_1\}$$

$f(x) = P_x$ (BM hits outer sphere first).

The solution (known to be unique)

$$f(x) = \begin{cases} \frac{\log \|x\|_2 - \log R_1}{\log R_2 - \log R_1} & (d=2) \\ \frac{R_1^{2-d} - \|x\|_2^{2-d}}{R_1^{2-d} - R_2^{2-d}} & (d \geq 3) \end{cases}$$

By taking $R_2 \rightarrow +\infty$ while R_1, x fixed.

$$\lim_{R_2 \rightarrow +\infty} f(x) = \begin{cases} 0 & (d=2) \\ 1 - \left(\frac{\|x\|_2}{R_1}\right)^{2-d} & (d \geq 3) \end{cases}$$

So we got, for $R_1 < \|x\|_2$.

$$P_x \left(\text{never hit } \mathbb{B}(0, R_1) \right) = \begin{cases} 0 & (d=2) \\ 1 - \left(\frac{R_1}{\|x\|_2} \right)^{d-2} & (d \geq 3). \end{cases}$$

(c.f. recurrence of SRW in $d=2/ d \geq 3$).

e.g. Feynman-Kac formula.

Suppose that Z satisfies SDE

$$dZ_t = a(Z_t)dt + b(Z_t)dB_t.$$

"diffusion process"

Consider a linear PDE.

$$\frac{\partial V}{\partial t}(t, x) = \frac{1}{2} b^2(x) \cdot \frac{\partial^2 V}{\partial x^2}(t, x) + a(x) \cdot \frac{\partial V}{\partial x}(t, x) + u(x) \cdot V(t, x).$$

$$V(0, x) = f(x)$$

$$V(t, x) = \mathbb{E}_x \left[f(Z_t) \cdot \exp \left(\int_0^t u(Z_s) ds \right) \right].$$

is the solution.

Idea: apply Itô's formula to

$$M_s = V(t-s, Z_s) \cdot \exp \left(\int_0^s u(Z_r) dr \right).$$