

$$X(t) = \int_0^{W(t)} e^{-s^2} ds$$

$$f(x) = \int_0^x e^{-s^2} ds. \quad \frac{\partial f}{\partial t} = 0 \quad \frac{\partial f}{\partial x} = e^{-x^2}.$$
$$\frac{\partial^2 f}{\partial x^2} = -2x e^{-x^2}.$$

$$\text{Ito: } dX = \frac{\partial f}{\partial x} dW + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} d[W, W]$$
$$= e^{-W(t)^2} dW(t) + \frac{1}{2} (-2W(t)) (e^{-W(t)^2}) dt$$

$$Y(t) = \sin(X_t).$$

Find Itô Decomp & Q.V.

$$Y(t) = W\left(\frac{1}{t}\right) \cdot t$$

$W \rightarrow$ BM.

$$\text{let } B(t) = t W\left(\frac{1}{t}\right).$$

Claim: B is a B.M.

Note: $\mathcal{F}_t^W =$ filtration of W .

B is NOT ADAPTED to \mathcal{F}^W

Let $\mathcal{F}_t^B =$ filtration generated by B .

B is adapted w.r.t \mathcal{F}_t^B .

To see B is a BM.

① $B(t) \sim t \cdot N(0, \frac{1}{t}) \Rightarrow B$ is Gaussian.

$$E B(t)^2 = t^2 E W(\frac{1}{t})^2 = t^2 \cdot \frac{1}{t} = t.$$

② Also check: $B(t) - B(s) = t W(\frac{1}{t}) - s W(\frac{1}{s})$.

Compute $E(B(t) - B(s))^2 \xrightarrow{\text{Gaussian.}} t - s \quad (s < t)$.

Check:
$$\left(tW\left(\frac{1}{t}\right) - sW\left(\frac{1}{s}\right) \right)^2 = t^2 W\left(\frac{1}{t}\right)^2 + s^2 W\left(\frac{1}{s}\right)^2 - 2stW\left(\frac{1}{t}\right)W\left(\frac{1}{s}\right).$$

$$\Rightarrow E(\quad)^2 = t + s - 2st \left[\left(\frac{1}{t}\right) \wedge \left(\frac{1}{s}\right) \right]$$

$$= t + s - \frac{2st}{t} = t - s.$$

③ Check $B(t) - B(s)$ ind of $\mathcal{F}_s^B \Rightarrow B$ is a B.M.

$B(0)$: Define $B(0) = \lim_{t \rightarrow 0} tW\left(\frac{1}{t}\right) = 0$ a.s.

$$\hat{C}(T, S(T)) = (S_T^2 - K)^+.$$

Find $\hat{C}(T, x)$.

Option 1: RNP formula: \mathbb{E}^Q

$$\hat{C}(t, S(t)) = \mathbb{E}^Q \left(e^{-r(T-t)} (S_T^2 - K)^+ \mid \mathcal{F}_t \right).$$

Option 2: Use hint.

$$\tilde{S} = S^2. \quad d\tilde{S} = 2S dS + d[S, S].$$

$$= 2S(\mu S dt + \sigma S dW) + \sigma^2 S^2 dt.$$

$$= 2\mu S^2 dt + 2\sigma S^2 dW + \sigma^2 S^2 dt$$

$$= \tilde{S}(2\mu + \sigma^2) dt + 2\sigma \tilde{S} dW$$

$\Rightarrow \tilde{S}$ is a GBM with mean return $2\mu + \sigma^2$
& vol 2σ .

\hat{c} = call option on \tilde{S} with strike K & mat T .

$$\Rightarrow \hat{c} = x N(d_+) - K e^{-r(T-t)} N(d_-)$$

$$d_{\pm} = \frac{1}{2\sigma\sqrt{t}} \left(\ln \frac{x}{K} + \left(r \pm \frac{1}{2}(2\sigma)^2 t \right) \right)$$

Q: $W \rightarrow$ BM.

$$B(t) = \int_0^t \text{sign}(W_e(s)) dW(s). \quad \text{sign}(x) = \begin{cases} +1 & x > 0 \\ -1 & x < 0 \\ 0 & x = 0 \end{cases}$$

① Compute $d[W, B] = \text{sign}(W(t)) d[W, W](t)$.

$$= \text{sign}(W(t)) dt.$$

② Is B a BM?

Yes: Levy. ① Cts, mg (true bc Ito integrals).
② $d[B, B] = dt$

$$d[B, B] = [\text{sign}(W(t))]^2 dt = dt.$$

By Levy $\Rightarrow B$ is a B.M.

$$[B, B](t) = \int_0^t \text{sign}(W(s))^2 ds$$

$$= \int_0^t \mathbb{1}_{\{W(s) \neq 0\}} ds \stackrel{\text{a.s.}}{=} \int_0^t 1 ds.$$

$B \rightarrow \text{BM.}$
 $W \rightarrow \text{B.M.}$

$Q: \text{Is } BW \text{ a mg?}$
 $\text{Cov}(BW)?$

Compute $E(BW)$

$$\hookrightarrow d(BW) = B dW + W dB + d[B, W].$$

$$\Rightarrow d(BW) = B dW + W dB + \text{sign}(W) ds.$$

$$E(B(t)W(t)) = E\left(\int_0^t B(s) dW(s) + W(s) dB(s)\right) + \underbrace{\int_0^t E \text{sign}(W(s)) ds}_0.$$

$$\Rightarrow E(BW) = 0$$

Q: Are B & W independent?

① $B(t) \sim N(0, t)$

② $W(t) \sim N(0, t)$

③ $E B(t)W(t) = 0$

Guess B & W are ind! FALSE!

B & W are NOT Jointly Gaussian.

Check B & W are not Independent!

Option 1: Use Hint & suffer ($E B(t)W(t)^2$)

Option 2: M & N are cts ind mg $\Rightarrow [M, N] = 0$

$d[B, W] = \text{sign}(W(t)) dt \neq 0 \Rightarrow B, W$ not ind!!

~~$[X+Y]$~~ =

$$[X+Y, X+Y] = [X, X] + [Y, Y] + 2[X, Y].$$

Claim: $[X, Y+Z] = [X, Y] + [X, Z].$

Intuition: $[X, Y] = \lim_{\|P\| \rightarrow 0} \sum \Delta_i X \Delta_i Y.$

$$\Delta_i X = X(t_{i+1}) - X(t_i).$$

$$\sum (\Delta_i X) (\Delta_i (Y+Z)) = \sum (\Delta_i X) (\Delta_i Y + \Delta_i Z).$$

Paraphrase check:

$$\begin{aligned} X &= M + B_1 \\ Y &= N + B_2 \\ Z &= L + B_3 \end{aligned}$$

M, N, L mg's.
 B_1, B_2, B_3 fake first var.

$$[X, Y+Z] = [M, L+N].$$

$$M(L+N) - [M, L] - [M, N]$$

$$= (ML - [M, L]) + (MN - [M, N]).$$

is a mg.

$$\Rightarrow [M, L] + [M, N] = [M, L+N]$$

QED!

$$X(t) = \int_0^t s \, dW(s) \quad Y(t) = \int_0^t W(s) \, ds.$$

$$E X(t)^n \quad \& \quad E Y(t)^n$$

$$X(t) = \int_0^t \sigma(s) \, dW(s) \quad \& \quad \sigma \text{ is } \boxed{\text{not random.}}$$

\Rightarrow $X(t)$ is gaussian &

$$X(t) \sim N\left(0, \int_0^t \sigma(s)^2 \, ds\right).$$

$$\mathbb{Q} \quad Y(t) = \int_0^t (t-s) dW(s).$$

Is Y a mg ?

Recall: $I(t) = \int_0^t \sigma(s) dW(s)$, σ adapted.

$\Rightarrow I(t)$ is a mg .

But $\int_0^t \sigma(s,t) dW(s)$ NEED not be a mg !

Check Y is not a mg.

$$\textcircled{1} Y(t) = \int_0^t t dW(s) - \int_0^t s dW(s).$$

$$= tW(t) - \underbrace{\int_0^t s dW(s)}.$$

Is a mg.

Pick $\tau < t$: Compute $E(Y(t) | \mathcal{F}_\tau) = tW(\tau) - \int_0^\tau s dW(s).$

$$\neq Y(\tau)$$

$\Rightarrow Y$ is not a mg.

$$M(t) = \int_0^t \text{Sign}(W(s)) dW(s)$$

$$E M(t)^2 = E \int_0^t \text{Sign}(W(s))^2 ds = \int_0^t ds = t.$$

$$\lim_{t \rightarrow \infty} M(t)^2 = \infty. \text{ OK??}$$

Need for any $t < \infty$ $E M(t)^2 < \infty$.

OK for $\lim_{t \rightarrow \infty} E M(t)^2 = +\infty$.