

(3) x & y are independent and standard normal random variables

$$E(e^{xy/2}) = E(E(e^{xy/2} | y = y))$$

$$\begin{aligned} E(e^{xy/2} | y = y) &= E(e^{\frac{yx}{2}} | y = y) = E(e^{yx/2}) \\ &= e^{\frac{y^2}{2 \times 2}} \\ &= e^{y^2/8} \end{aligned}$$

$y \quad x \sim N(\mu, \sigma^2)$

$$E(e^{tx}) = e^{t\mu + \frac{\sigma^2 t^2}{2}}$$

$$E(e^{xy/2} | y = y) = e^{y^2/8}$$

$$E(e^{xy/2} | y) = e^{\frac{y^2}{8}} = e^{\frac{y^2}{8}}$$

$$E(e^{\frac{xy}{2}}) = E(e^{\frac{y^2}{8}})$$

$$= \int_{-\infty}^{\infty} e^{y^2/8} \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy$$

$$= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{3}{8}y^2} dy$$

$$= \frac{2}{\sqrt{3}}$$

$$E(f(x) | y) = g(y)$$

$$\boxed{E(x|y) = g(y)}$$

$$\frac{1}{2} - \frac{1}{8} = \frac{3}{8}$$

$$2^2 = \frac{8}{3} \Rightarrow \sigma^2 = \frac{4}{3}$$

$$\frac{y}{\sigma} = 2$$

(2) Compute the distribution of $\int_0^t f(s) dW_s$
 $f(s)$ is a deterministic function.

(2)

$$E\left(e^{\lambda \int_0^t f(s) dW_s}\right)$$

$M_t = \frac{\langle M \rangle_t}{2}$ is a
 of M_t is a Martingale, then $e^{M_t - \frac{\langle M \rangle_t}{2}}$ is a Martingale

$$M_t = \lambda \int_0^t f(s) dW_s \quad \langle M \rangle_t = \lambda^2 \int_0^t f^2(s) ds$$

$$Z_t = e^{\int_0^t x_s dW_s - \frac{1}{2} \int_0^t x_s^2 ds}$$

$$Y_t = \int_0^t x_s dW_s - \frac{1}{2} \int_0^t x_s^2 ds$$

$$Z_t = e^{Y_t}$$

disgession $Y_t = f(x_t)$

$$dY_t = f'(x_t) dx_t + \frac{1}{2} f''(x_t) (dx_t)^2$$

$$Y_t = f(t, x_t)$$

$$dY_t = \frac{\partial f}{\partial t}(t, x_t) dt + \frac{\partial f}{\partial x}(t, x_t) dx_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2}(t, x_t) (dx_t)^2$$

$$Z_t = f(x_t, Y_t)$$

$$dZ_t = \frac{\partial f}{\partial x} dx_t + \frac{\partial f}{\partial y}(x_t, Y_t) dY_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} (dx_t)^2 + \frac{1}{2} \frac{\partial^2 f}{\partial y^2} (dY_t)^2 + \frac{\partial^2 f}{\partial x \partial y} (dx_t) dY_t$$

Prove

$$Z_t = e^{\int_0^t x_s dW_s - \frac{1}{2} \int_0^t x_s^2 ds} \text{ is a Martingale}$$

$$Y_t = \int_0^t x_s dW_s - \frac{1}{2} \int_0^t x_s^2 ds \quad dY_t = x_t dW_t - \frac{x_t^2}{2} dt$$

$$Z_t = e^{Y_t}$$

$$(dW_t)^2 = dt$$

$$(dt)^2 = 0$$

$$dW_t \cdot dt = 0$$

$$dZ_t = e^{Y_t} dY_t + \frac{1}{2} e^{Y_t} (dY_t)^2$$

~~0~~
~~0~~

$$dZ_t = e^{Y_t} \left(x_t dW_t - \frac{x_t^2}{2} dt + \frac{1}{2} x_t^2 dt \right)$$

$$dZ_t = x_t Z_t dW_t \Rightarrow Z_t = \int_0^t x_s Z_s dW_s$$

$\Rightarrow Z_t$ is a Martingale.

$Z_t = e^{M_t - \frac{1}{2} \langle M \rangle_t}$ is a Martingale if M_t is a Martingale

② Compute the distribution of $\int_0^t f(s) dW_s$ $f(s)$ is a deterministic function

$$E\left(e^{\lambda \int_0^t f(s) dW_s}\right)$$

$$Z_t = e^{\lambda \int_0^t f(s) dW_s - \frac{1}{2} \int_0^t \lambda^2 f(s)^2 ds}$$

$$E(Z_t) = E(Z_0) = 1$$

$$E\left(e^{\lambda \int_0^t f(s) dW_s - \frac{1}{2} \int_0^t \lambda^2 f(s)^2 ds}\right) = 1$$

$$E(g(s)h(\eta)) = g(s)E(h(\eta))$$

$$E\left(e^{\lambda \int_0^t f(s) dW_s}\right) = e^{\frac{1}{2} \int_0^t \lambda^2 f(s)^2 ds} \quad \sigma^2 = \int_0^t f(s)^2 ds$$

$$\textcircled{1} e^{-\frac{1}{2} \int_0^t \lambda^2 f(s)^2 ds} E\left(e^{\lambda \int_0^t f(s) dW_s}\right) = 1$$

$$E\left(e^{\lambda \int_0^t f(s) dW_s}\right) = e^{\frac{\lambda^2 \sigma^2}{2}}$$

$$\Rightarrow \int_0^t f(s) dW_s \sim N\left(0, \int_0^t f(s)^2 ds\right)$$

$$\textcircled{2} \int_0^t W_s dW_s \rightarrow \text{Normally distributed}$$

(4)

①. $x_t = e^{y_t}$

⑤.

Find the differential of x_t of the process x_t

$$x_t = e^{y_t} \quad y_t = tW_t$$

$$dx_t = e^{y_t} dy_t + \frac{1}{2} e^{y_t} (dy_t)^2$$

$$y_t = tW_t$$

$$* dx_t y_t = x_t dy_t + y_t dx_t + dx_t dy_t$$

$$dy_t = t dW_t + W_t dt + \underbrace{dt dW_t}_0$$

$$dy_t = t dW_t + W_t dt$$

$$dx_t = x_t \left(dy_t + \frac{1}{2} (dy_t)^2 \right)$$

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

$$dx_t = x_t \left(W_t + \frac{t^2}{2} \right) dt + t x_t dW_t$$

Compute $E(e^{\alpha W(t)})$

$$\langle \chi_t \rangle = \int_0^t (\langle d\chi_s \rangle)^2 \quad (6)$$

$$Z_t = e^{\alpha W_t - \frac{\alpha^2 t}{2}} \Rightarrow Z_t \text{ is a mg}$$

$$W(t) = \sqrt{t} Z$$

$$E(Z_t) = E(Z_0) = 1$$

$$E\left(e^{\alpha W_t - \frac{\alpha^2 t}{2}}\right) = 1$$

$$E(e^{\alpha W_t}) = e^{\frac{\alpha^2 t}{2}}$$

$$E(e^{\alpha \sqrt{t} Z}) = e^{\frac{\alpha^2 t}{2}}$$

$$\int_0^t W_s dW_s$$

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

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

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

$$\int_0^t W_s^{n-1} dW_s$$

$$dW_t^n = n W_t^{n-1} dW_t + \frac{1}{2} n(n-1) W_t^{n-2} dt$$

$$W_t^n - \frac{1}{2} \int_0^t n(n-1) W_s^{n-2} ds = n \int_0^t W_s^{n-1} dW_s$$

① Find the distribution of the Random variable

⑦.

$$\int_0^t w_s ds$$

$$d(tw_t) = t dw_t + w_t dt + \underbrace{dt dw_t}_0$$

$$tw_t - \int_0^t s dw_s = \int_0^t w_s ds$$

$$df(t, x_t) = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial x} dx_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} (dx_t)^2$$

$$f(t, w_t) = tw_t$$

$$df(t, w_t) = w_t dt + t dw_t + \frac{1}{2} \times 0 \times (dw_t)^2$$

$$d(tw_t) = w_t dt + t dw_t$$

$$tw_t = t \int_0^t dw_s$$

$$tw_t - \int_0^t s dw_s = t \int_0^t dw_s - \int_0^t s dw_s$$

$$= \int_0^t (t-s) dw_s = \int_0^t w_s ds$$

$$\int_0^t w_s ds = \int_0^t (t-s) dw_s \sim N\left(0, \int_0^t (t-s)^2 ds\right)$$

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

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Let $f(x)$ be a function and λ be a number

s.t. $f(0) = 1$, ~~$f(0) = 1$~~ $f''(x) = \lambda f(x)$

Prove ~~$E(f(W_t)) = e^{\frac{\lambda t}{2}}$~~ $E(f(W_t)) = e^{\frac{\lambda t}{2}}$

$$df(W_t) = f'(W_t)dW_t + \frac{1}{2}f''(W_t)dt$$

$$f(W_t) - f(W_0) = \int_0^t f'(W_s)dW_s + \frac{1}{2} \int_0^t f''(W_s)ds$$

$$f(W_t) - f(0) = \int_0^t f'(W_s)dW_s + \frac{1}{2} \int_0^t f''(W_s)ds$$

$$E(f(W_t)) = 1 + \frac{1}{2} E \int_0^t f''(W_s)ds$$

$$= 1 + \frac{1}{2} E \int_0^t \lambda f(W_s)ds$$

$$E(f(W_t)) = 1 + \frac{1}{2} \lambda \int_0^t E(f(W_s))ds$$

$$E(f(W_t)) = \cancel{y(t)} y(t)$$

$$y(t) = 1 + \frac{\lambda}{2} \int_0^t y(s)ds$$

$$\frac{dy(t)}{dt} = y'(t) = \frac{\lambda}{2} y(t)$$

$$\frac{dy(t)}{y(t)} = \frac{\lambda}{2} dt$$

~~$y(t)$~~ $\ln y(t) = \frac{\lambda}{2} t + C$

$$\Rightarrow y(t) = K e^{\frac{\lambda t}{2}} = E(f(W_t))$$

$$E(f(W_t)) = Ke^{\frac{\lambda t}{2}}$$

plug $t=0$

$$E(f(W_0)) = K$$

$$\Rightarrow K=1$$

$$\Rightarrow E(f(W_t)) = e^{\frac{\lambda t}{2}}$$

4 True or False Questions

a) if W & B are independent BMS, then the average of W & B given by $X_t = \frac{1}{2}(W_t + B_t)$ is again a Brownian Motion

A process X_t is said to be a BM if it satisfies the following properties

(i) $X_0 = 0$

(ii) $E(X_t X_s) = \min(t, s) (E[(X_t - X_s + X_s) X_s])$ if $t \geq s$

(iii) X_t is Gaussian distributed with mean 0

$$E[(X_t - X_s) X_s] = E[X_t - X_s] E[X_s] = 0 \cdot s = 0$$

(iv) X_t has continuous sample paths

Show that $X_t = \frac{W_{ct}}{\sqrt{c}}$ is BM $c > 0$ is a constant

$$E(X_t X_s) \text{ for } t \geq s$$

$$E\left(\frac{1}{2}(W_t + B_t) + \frac{1}{2}(W_s + B_s)\right)$$

$$\frac{1}{4} E\left(W_t W_s + \underbrace{W_t B_s}_0 + \underbrace{B_t W_s}_0 + B_t B_s\right)$$

$$\frac{1}{4} \{s + s\} = \frac{s}{2} \neq s$$

Hence X_t is not BM

(b) If X & Y are Martingales then the average of X & Y given by $Z_t = \frac{X_t + Y_t}{2}$ is a Martingale

True : $E(Z_t | F_s) = Z_s$

(c) X has finite non zero quadratic variation then X has infinite first variation

True

$$\sum |X_{t_{i+1}} - X_{t_i}|^2 \leq \max_i |X_{t_{i+1}} - X_{t_i}| \sum |X_{t_{i+1}} - X_{t_i}|$$

If the process has finite first variation

then $\sum |X_{t_{i+1}} - X_{t_i}| < \infty$

$$\Rightarrow \sum |X_{t_{i+1}} - X_{t_i}|^2 = 0$$

$$\textcircled{1} \quad P(X \geq t, N \geq y) = P(X \geq t) P(N \geq y)$$

(11)

$$E(f(X)g(N)) =$$

$$E(f(X)h(N^2)) =$$

(1)

$$P(X \leq tY) = \int_0^{\infty} P(X \leq tY | Y=y) P(Y=y) dy$$

$$X \& Y > 0$$

$$\text{and } X \& Y \perp$$

$$X \sim \exp(1)$$

$$= \int_0^{\infty} (1 - e^{-ty}) f_Y(y) dy$$

$$Y \sim N^2$$

$$= 1 - E(e^{-tY})$$

$$E(e^{-tY}) = E(e^{-tN^2}) = \int_{-\infty}^{\infty} e^{-ty^2} \frac{1}{\sqrt{2\pi}} e^{-y^2/2} dy$$

$$= \frac{1}{\sqrt{1+2t}}$$

$$P(Z \geq t) \quad Z = \frac{X}{Y} \quad Y = N^2(0,1)$$

(2)

$$x_t = e^{-t} W_t^2$$

$$f(t, x) = e^{-t} x^2$$

(12)

Obtain the Ito process decomposition of x_t

$$y_t = f(t, x_t) \quad y_t = f(t, W_t)$$

$$dy_t \equiv \frac{\partial f}{\partial t}(t, x_t) dt + \frac{\partial f}{\partial x}(t, x_t) dx_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2}(t, x_t) (dx_t)^2$$

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

$$dx_t = e^{-t} (1 - W_t^2) dt + 2e^{-t} W_t dW_t$$

(b) Compute the Ito process decomposition of

$$y_t = \sin x_t$$

$$dy_t = \cos x_t dx_t + \frac{1}{2} \sin(x_t) (dx_t)^2$$

$$(dx_t)^2 = 4e^{-2t} W_t^2 dt$$

What is the quadratic variation of x_t ?

Compute $\langle x \rangle_t$

$$\langle x \rangle_t = \int_0^t (dx_s)^2 = \int_0^t 4e^{-2s} W_s^2 ds$$

③ $\{M_t\}$ is a continuous Martingale. Fix a time $t_0 > 0$ ⑬
 and let $\Delta = \{\Delta_t\}_{t \geq 0}$ be the simple process

$$\text{defined by } \Delta_t = \begin{cases} \Delta_0 & \text{if } t \leq t_0 \\ \Delta_{t_0} & \text{if } t > t_0 \end{cases} \quad \Delta_0 \text{ is a constant}$$

Δ_{t_0} is \mathcal{F}_{t_0} measurable

Define the stochastic integral $\int \Delta$ with respect to M by the formula

$$I_t = \Delta_0 M_{t \wedge t_0} + \Delta_{t_0} (M_t - M_{t \wedge t_0}) \quad t \geq 0$$

$$a \wedge b = \min\{a, b\}$$

(a) I is a Martingale

$$E(I(t) | \mathcal{F}_s) \stackrel{?}{=} I(s) \quad t \geq s$$

(i) $s \leq t \leq t_0$

$$I_t = \Delta_0 M_t + \Delta_{t_0} (M_t - M_t) = \Delta_0 M_t$$

$$I_t = \Delta_0 M_t$$

$$E(I_t | \mathcal{F}_s) = E(\Delta_0 M_t | \mathcal{F}_s) = \Delta_0 M_s = I_s$$

(ii) $s \leq t_0 \leq t$

$$\mathcal{F}_s \subseteq \mathcal{F}_{t_0}$$

$$I(t) = D_0 M_{t_0} + \Delta_{t_0} (M_t - M_{t_0})$$

$$\begin{aligned} E(I(t) | \mathcal{F}_s) &= E(D_0 M_{t_0} + \Delta_{t_0} (M_t - M_{t_0}) | \mathcal{F}_s) \\ &= D_0 M_s + E(\Delta_{t_0} (M_t - M_{t_0}) | \mathcal{F}_s) \\ &= D_0 M_s + E(E(\Delta_{t_0} (M_t - M_{t_0}) | \mathcal{F}_{t_0}) | \mathcal{F}_s) \end{aligned}$$

$$\begin{aligned} E(\Delta_{t_0} (M_t - M_{t_0}) | \mathcal{F}_{t_0}) &= \Delta_{t_0} E(M_t - M_{t_0} | \mathcal{F}_{t_0}) \\ &= 0 \end{aligned}$$

$$E(I(t) | \mathcal{F}_s) = D_0 M_s = I(s)$$

$$I(s) = D_0 M_s + \Delta_{t_0} (M_s - M_s) = D_0 M_s$$

(iii) $t_0 \leq s \leq t$

$$I(t) = D_0 M_{t_0} + \Delta_{t_0} (M_t - M_{t_0})$$

$$\begin{aligned} E(I(t) | \mathcal{F}_s) &= E(D_0 M_{t_0} + \Delta_{t_0} (M_t - M_{t_0}) | \mathcal{F}_s) \\ &= D_0 M_{t_0} + \Delta_{t_0} (M_s - M_{t_0}) \\ &= I_s \end{aligned}$$

Hence $I(t)$ is a Martingale.