

→ Final → Mar 6th (5:30 PM)

NY: Steve Shere → Office Hours 2:30 PM (Sat).

Pitt: Me → Office Hours 2:30 PM (Friday).

Next week: Hope 2:45 PM (~~Wed~~)(Tue) → Problem session (Here).
(Maybe).

FCE: Inactive: 80% Response Rate by TOMORROW

⇒ Final Grades Released ASAP.

(If not Final Grades on March 13th 5:00 PM)

Response rate now: 50%.

Q1: Let $X(t) = \int_0^t f(s) dW(s)$

$f \rightarrow$ Non Random fn.

Find dist of X .

Sol: Know X is normal (sum of ^{Joint} Normals is Normal.
& lim of Normals is \uparrow)

Compute $EX(t)$ & $EX(t)^2$.

(1) $EX(t) = 0$

(2) $EX(t)^2 = \int_0^t f(s)^2 ds$. (Itô Isometry).

Variant: Q: $X(t) = \int_0^t f(s) W(s) ds$.

Compute the dist of X .

↘ f not random.

Sol: $X(t) = \lim_{\|P\| \rightarrow 0} \underbrace{\sum f(t_i) W(t_i) (t_{i+1} - t_i)}_{\text{Riemann}}.$

Expect $X(t) \sim N(\mu, \sigma^2)$. Compute μ & σ .

① $E X(t) = E \int_0^t f(s) W(s) ds = \int_0^t f(s) (E W(s)) ds = 0,$

$$\textcircled{2} \quad \mathbb{E} X(t)^2 \stackrel{\text{trick 1}}{=} \mathbb{E} \left(\int_0^t f(s) W(s) ds \right) \left(\int_0^t f(\tau) W(\tau) d\tau \right)$$

$$= \mathbb{E} \int_0^t \int_0^t f(s) f(\tau) W(s) W(\tau) ds d\tau -$$

$$= \int_{s=0}^t \int_{\tau=0}^t f(s) f(\tau) (s \wedge \tau) ds d\tau. \quad \Delta \text{ compute this.}$$

② Trick 2°. $dX = f(t) w(t) dt.$

Let $F(t) = \int_0^t f(s) ds$ ($F' = f$).

$d(F(t)W(t)) = F dW + \underbrace{W \cdot f dt}_{dX} + d[F, W].$

$\Rightarrow F(t)W(t) = \int_0^t F(s) dW(s) + \underbrace{\int_0^t W(s) f(s) ds}_{X(t)},$

$\Rightarrow X(t) = F(t)W(t) - \int_0^t F(s) dW(s).$

$$\Rightarrow X(t)^2 = F(t)^2 W(t)^2 + \left(\int_0^t F(s) dW(s) \right)^2$$

$$- 2 F(t) W(t) \int_0^t F(s) dW(s).$$

$$\Rightarrow E X(t)^2 = F(t)^2 t + \int_0^t F(s)^2 ds - 2 F(t) E \left(W(t) \int_0^t F(s) dW(s) \right),$$

(Itô Isom).

(*)

Let $M(t) = \int_0^t F(s) dW(s)$. Compute $E W(t) M(t)$.

$$\begin{aligned} d(WM) &= W dM + M dW + d[W, M] \\ &= W dM + M dW + F(t) dt \end{aligned}$$

$$\left(\begin{array}{l} dW \\ dM \end{array} = \begin{array}{l} 1 \\ F(t) \end{array} dW \right) d[W, M] = 1 F(t) d[W, W]$$

$$\Rightarrow W(t)M(t) = \int_0^t W(s) dM(s) + \int_0^t M(s) dW(s) + \int_0^t F(s) ds.$$

$$\Rightarrow E M(t)W(t) = 0 + 0 + \int_0^t F(s) ds.$$

& substitute in (*) for an answer!

$$Q: Y(t) = \int_0^t f(s) W(s) dW(s). \quad \text{is } Y \text{ Normal?}$$

NO: $f(t_i) W(t_i) (W(t_{i+1}) - W(t_i))$, ~~can be an~~ ^{not normal.}

Q 20. $M, N \rightarrow 2 \text{ mg (ds)}$.

$$\left. \begin{aligned} d[M, M] &= \sigma(t) dt \\ d[N, N] &= \tau(t) dt \\ d[M, N] &= \rho(t) dt \end{aligned} \right\} \sigma, \tau, \rho \text{ are non-random.}$$

(a) Compute $E \exp(\lambda M(t) + \mu N(t))$.

Sol: Ito: $Y(t) = e^{\lambda M(t) + \mu N(t)} = g(M(t), N(t)).$

where $g(x, y) = e^{\lambda x + \mu y}$.

$$\partial_x g = \lambda g, \quad \partial_y g = \mu g.$$

$$\partial_x^2 g = \lambda^2 g, \quad \partial_y^2 g = \mu^2 g$$

$$\partial_x \partial_y g = \lambda \mu g.$$

$$\Rightarrow dY = \partial_x g dM + \partial_y g dN + \frac{1}{2} \left(\partial_x^2 g d[M, M] + \partial_y^2 g d[N, N] + 2 \partial_x \partial_y g d[M, N] \right).$$

$$\Rightarrow dY = \lambda Y dM + \mu Y dN + \frac{1}{2} \left[\lambda^2 Y \sigma + \mu^2 Y \tau + 2 \lambda \mu Y \rho \right] dt$$

$$\Rightarrow Y(t) = 1 + \int_0^t \lambda Y(s) dM(s) + \int_0^t \mu Y(s) dN(s) + \frac{1}{2} \int_0^t (\lambda^2 \sigma + \mu^2 \tau + 2 \lambda \mu \rho) Y ds$$

Let $f(t) = E Y(t)$

$$\Rightarrow f(t) = 1 + 0 + 0 + \frac{1}{2} \int_0^t (\lambda^2 \sigma + \mu^2 \tau + 2 \lambda \mu \rho) f(s) ds.$$

$$\Rightarrow \cancel{f(t)} \circledast f = \frac{1}{2} (\lambda^2 \tau + \mu^2 \tau + 2\lambda\mu\rho) f$$

$$\Rightarrow f(t) = 1 \exp\left(\frac{1}{2} \int_0^t (\lambda^2 \tau(s) + \mu^2 \tau(s) + 2\lambda\mu\rho(s)) ds\right)$$

$$\parallel$$

$$E \exp(\lambda M(t) + \mu N(t))$$

↑
MGF of a joint normal!

⑥ Suppose $\tau(t) = \tau(t) = 1$
& $\rho(t) = 0$

$$\text{Then } E \exp(\lambda M(t) + \mu N(t)) = \exp\left(\frac{1}{2} (\lambda^2 t + \mu^2 t)\right)$$

or MGF of $\mathbb{R}^2 N(0, \begin{pmatrix} t & 0 \\ 0 & t \end{pmatrix})$

to show (M, N) is a 2D BM,

Just compute $E\left(\exp\left(\lambda M(t) + \mu N(t)\right) \mid \mathcal{G}_s\right)$

Q3: Financial Market

- Risky asset. Price GBM (α, σ) . (S) .
- Money market: Interest rate r .

Security: Payoff at maturity T is $\frac{1}{T} \int_0^T S(s) ds$.

Compute the AFP at time $t \leq T$.

Sol: RNP formula: $V(T) = \frac{1}{T} \int_0^T S(s) ds.$

$\tilde{P} \rightarrow$ Risk Neutral measure.

$d\tilde{P} = z(T) dP$ \leftarrow Forget this formula.

& $z(t) = \exp\left(\underbrace{(\alpha - \frac{\sigma^2}{2})}_{\frac{\alpha - \sigma^2}{2}} \frac{W(t)}{\sigma} - \frac{1}{2} \left(\frac{\alpha - \sigma^2}{\sigma}\right)^2 t\right)$

$V(t) = \tilde{E} \left(e^{-r(T-t)} V(T) \mid \mathcal{F}_t \right).$

Only use, under \tilde{P} , $S \sim \text{GBM}(r, \sigma),$

$$\mathbb{E}^Q \left(\frac{e^{-r(T-t)}}{T} \int_0^T S(s) ds \mid \mathcal{F}_t \right) = \mathbb{E}^Q \left(\int_0^t + \int_t^T \mid \mathcal{F}_t \right).$$

$$= \frac{e^{-r(T-t)}}{T} \int_0^t S(s) ds + \frac{e^{-r(T-t)}}{T} \int_t^T \underbrace{\mathbb{E}^Q(S(s) \mid \mathcal{F}_t)}_{\text{complete this. (s > t)}} ds.$$

$S \rightarrow$ GBM under \tilde{P}
(r, σ).

$$S(s) = S(0) \exp \left(\left(r - \frac{\sigma^2}{2} \right) s + \sigma \tilde{W}(s) \right),$$

$$S(t, s) = S(t) \exp \left(\left(r - \frac{\sigma^2}{2} \right) (s-t) + \sigma (\tilde{W}(s) - \tilde{W}(t)) \right).$$

$$\Rightarrow \mathbb{E}^Q(S(s) | \mathcal{F}_t) = S(t) \exp\left((r - \frac{\sigma^2}{2})(s-t)\right) \cdot$$

$$\cdot \tilde{\mathbb{E}}\left(\exp\left(\sigma(\tilde{W}(s) - \tilde{W}(t))\right) \mid \mathcal{F}_t\right)$$

$$= S(t) \exp\left((r - \frac{\sigma^2}{2})(s-t)\right) \cdot \tilde{\mathbb{E}} \exp\left(\sigma \sqrt{s-t} \overset{\text{std normal.}}{\downarrow} N\right)$$

$$= S(t) \exp\left((r - \frac{\sigma^2}{2})(s-t)\right) \left\{ \exp\left(\frac{\sigma^2(s-t)}{2}\right) \right\}$$

$$= S(t) \exp\left(r(s-t)\right)$$

$$\therefore V(t) = \frac{e^{-r(T-t)}}{T} \int_0^t S(s) ds + \left(\frac{e^{-r(T-t)}}{T} \int_t^T \exp(r(s-t)) ds \right) S(t)$$

$$V(t) = \frac{e^{-r(T-t)}}{T} \int_0^t S(s) ds + \frac{S(t)}{rT} \left(1 - e^{-r(T-t)} \right)$$

**

Q: What is the trading strategy to build the R. Portfolio?

$$\hookrightarrow dV = (\Delta(t)) dS + r(V(t) - \Delta(t)S(t)) dt$$

from ** : $\Delta(t) = \frac{1 - e^{-r(T-t)}}{rT}$

~~Q: $X \sim N(0, 1)$. $a \in \mathbb{R}$.~~
 ~~$X+a \sim N(\dots)$~~

RNP formula:

$$V(t) = \tilde{\mathbb{E}} \left(\exp \left(- \int_t^T R(s) ds \right) V(T) \mid \mathcal{F}_t \right)$$

Q: $W \rightarrow \text{BM}$.

Let $dB = \text{sign}(W(t)) dW$

Q: Is B a BM?

$$\text{sign}(x) = \begin{cases} 1 & x > 0 \\ -1 & x \leq 0 \end{cases}$$

Yes! ① B is a dt's mg. ✓

$$(B(t) = \int_0^t \text{sign}(W(s)) dW(s).)$$

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

By Levy $\Rightarrow B$ is a BM.

② Compute $[B, W]$ & $E(BW)$.

$$d[B, W] : \rightarrow \left. \begin{array}{l} dB = \text{sign}(W(t)) dW \\ dW = 1 \quad dW \end{array} \right\} d[B, W] = \text{sign}(W(t)) dt$$

$$E(\cancel{BW}) E(B(t)W(t)) = 0$$

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

$$= B dW + W dB + \text{Sign}(W(t)) dt.$$

$$\Rightarrow B(t)W(t) = 0 + \int_0^t B(s) dW(s) + \int_0^t W(s) dB(s) + \int_0^t \text{Sign}(W(s)) ds.$$

$$\Rightarrow E B(t)W(t) = 0 + 0 + 0 = 0.$$

Q: $B(t) \sim N(0, t).$

$W(t) \sim N(0, t).$

$E B(t)W(t) = 0$

Q: Are B & W independent??

NO: If B & W are ind

$\Rightarrow [B, W] = 0.$ But $[B, W] = \int_0^t \text{Sign}(W(s)) ds,$