

EXERCISES 7

ITO'S FORMULA AND GIRSANOV FORMULA

Exercise 1. Let $(B_t)_{t \geq 0}$ be a Brownian motion w.r.t. a filtration $(\mathcal{F}_t)_{t \geq 0}$ and $(u_t)_{t \geq 0}$ be a continuous and $(\mathcal{F}_t)_{t \geq 0}$ -adapted process such that

$$\forall t \geq 0, \forall \omega \in \Omega, |u_t(\omega)| \leq K,$$

for some constant $K > 0$. We admit the following inequality $\mathbb{E} \left[\exp \left(\int_0^t u_s dB_s \right) \right] \leq \exp(K^2 t/2)$.

- (1) Show that the process $\forall t \geq 0, M_t = \exp \left(\int_0^t u_s dB_s - \frac{1}{2} \int_0^t u_s^2 ds \right)$, is a martingale w.r.t. $(\mathcal{F}_t)_{t \geq 0}$. (Use Itô's formula.)
- (2) We set $\forall t \geq 0, Y_t = - \int_0^t u_s ds + B_t$. Show that the process $(Y_t M_t)_{t \geq 0}$ is a martingale w.r.t. $(\mathcal{F}_t)_{t \geq 0}$.

Exercise 2. Expand as an Itô process the process

$$\forall t \geq 0, X_t = (B_t^1)^2 + (B_t^2)^2 + (B_t^3)^2,$$

where $(B_t^1, B_t^2, B_t^3)_{t \geq 0}$ stands for a Brownian motion of dimension 3.

Exercise 3. Let $(B_t = (B_t^1, B_t^2))_{t \geq 0}$ be two independent Brownian motion w.r.t. a filtration $(\mathcal{F}_t)_{t \geq 0}$ and $(u_t)_{t \geq 0}$ and $(v_t)_{t \geq 0}$ be two continuous and $(\mathcal{F}_t)_{t \geq 0}$ -adapted processes, bounded by some constant K . Show that

$$\forall t \geq 0, M_t = \exp \left(\int_0^t u_s dB_s^1 + \int_0^t v_s dB_s^2 - \frac{1}{2} \int_0^t (u_s^2 + v_s^2) ds \right),$$

is a martingale.

Exercise 4. Let $(B_t)_{t \geq 0}$ be a Brownian motion w.r.t. a filtration $(\mathcal{F}_t)_{t \geq 0}$ and $(u_t)_{t \geq 0}$ and $(v_t)_{t \geq 0}$ be two continuous and adapted processes such that

$$\forall t \geq 0, \mathbb{E} \int_0^t (u_s^4 + v_s^4) ds < +\infty,$$

show that

$$\left(\left(\int_0^t u_s dB_s \right) \left(\int_0^t v_s dB_s \right) - \int_0^t u_s v_s ds \right)_{t \geq 0}$$

is a martingale.

Exercise 5. Let $(B_t^1, B_t^2)_{t \geq 0}$ be a Brownian motion with values in \mathbb{R}^2 . We assume that there exists a function u in $\mathcal{C}^{1,2}([0, +\infty) \times \mathbb{R}^2)$, bounded with bounded derivatives, such that

$$(*) \quad \forall (t, x) \in [0, +\infty[\times \mathbb{R}^2, \frac{\partial u}{\partial t}(t, x) - \frac{1}{2} \left(\frac{\partial^2 u}{\partial x_1^2} + \frac{\partial^2 u}{\partial x_2^2} \right)(t, x) = 0, \quad u(0, x) = h(x).$$

- (1) Show that for any $T > 0$ any $x \in \mathbb{R}$, the process $(u(T-t, x + B_t))_{0 \leq t \leq T}$ is a martingale.
- (2) Deduce that $u(T, x) = \mathbb{E}(h(x + B_T))$.

- (3) Deduce that (\star) admits at most one solution $\mathcal{C}^{1,2}$, bounded with bounded derivatives, with $u(0, \cdot)$ as initial condition.

Exercise 6. Let f be a deterministic locally admissible function.

- (1) Show that

$$\forall t \geq 0, \mathbb{E} \left[\exp \left(\int_0^t f_s dB_s \right) \right] = \exp \left(\frac{1}{2} \int_0^t f_s^2 ds \right).$$

- (2) Show that the process

$$\left(\exp \left(\int_0^t f_s dB_s - \frac{1}{2} \int_0^t f_s^2 ds \right) \right)_{t \geq 0}$$

is a martingale with respect to the natural filtration of B .

Exercise 7. Let $(B_t^1, B_t^2, B_t^3)_{t \geq 0}$ be a three dimensional Brownian motion w.r.t. some filtration $(\mathcal{F}_t)_{t \geq 0}$. For a given vector $(b_1, b_2, b_3) \in \mathbb{R}^3$, we consider the process

$$\forall t \geq 0, X_t = \exp \left(\sum_{i=1}^3 b_i B_t^i - \frac{1}{2} \sum_{i=1}^3 b_i^2 t \right).$$

- (1) Prove that $(X_t)_{t \geq 0}$ is a square integrable martingale.
 (2) Prove that the process $((B_t^1 + B_t^2 - (b_1 + b_2)t)X_t)_{t \geq 0}$ is also a martingale.

Exercise 8. Let $(B_t^1, B_t^2, B_t^3)_{t \geq 0}$ be a three dimensional Brownian motion w.r.t. some filtration $(\mathcal{F}_t)_{t \geq 0}$. For a given matrix σ of size 3×3 , we consider the process

$$\forall t \geq 0, X_t = \sigma \times \begin{pmatrix} B_t^1 \\ B_t^2 \\ B_t^3 \end{pmatrix}.$$

Show that the process $(M_t = \sum_{i=1}^3 (X_t^i)^2 - \text{Trace}(\sigma\sigma^*)t)_{t \geq 0}$ is a martingale.

Exercise 9. Let $(B_t)_{t \geq 0}$ be a Brownian and $(\mathcal{F}_t)_{t \geq 0}$ its natural filtration. For $\mu \in \mathbb{R}$ et $\sigma > 0$, we set

$$\forall t \geq 0, Y_t = \exp(\mu t + \sigma B_t),$$

referred as Geometric Brownian motion.

- (1) We set $r = \mu + \sigma^2/2$ and we define $\forall t \geq 0, \tilde{B}_t = B_t + \sigma^{-1}rt$. What can you say of $(\tilde{B}_t)_{0 \leq t \leq 1}$ under the probability:

$$\forall A \in \mathcal{A}, \mathbb{Q}(A) = \mathbb{E} \left[\exp(-\sigma^{-1}rB_1 - \frac{1}{2}\sigma^{-2}r^2) \mathbf{1}_A \right].$$

- (2) Show that $(Y_t)_{0 \leq t \leq 1}$ is, under the probability \mathbb{Q} , a martingale w.r.t. $(\mathcal{F}_t)_{0 \leq t \leq 1}$.

Exercise 10. Let $(B_t)_{t \geq 0}$ be a Brownian motion and $(\mathcal{F}_t)_{t \geq 0}$ its natural filtration, show that we can define a probability \mathbb{Q}_1 on (Ω, \mathcal{F}_1) , equivalent to \mathbb{P} , such that $(B_t + B_t^3)_{0 \leq t \leq 1}$ ($B_t^3 = (B_t)^3$) be a martingale w.r.t. $(\mathcal{F}_t)_{t \geq 0}$ under \mathbb{Q}_1 . Hint: apply first Itô's formula to $(B_t + B_t^3)_{t \geq 0}$.

Exercise 11. Let $(B_t)_{t \geq 0}$ be a Brownian motion and $(\mathcal{F}_t)_{t \geq 0}$ its natural filtration, show that we can define a probability \mathbb{Q}_1 on (Ω, \mathcal{F}_1) , equivalent to \mathbb{P} , such that $((2 + B_t^2) \exp(B_t))_{0 \leq t \leq 1}$ ($B_t^2 = (B_t)^2$) be a martingale w.r.t. $(\mathcal{F}_t)_{t \geq 0}$ under \mathbb{Q}_1 . Hint: apply first Itô's formula to $((2 + B_t^2) \exp(B_t))_{t \geq 0}$.

Worksheet 7

Exercise 1

$$M_t = e^{X_t} = e^{\int_0^t u_s dB_s - \frac{1}{2} \int_0^t u_s^2 ds}$$

$$f(t, X_t) = e^{X_t}$$

$$df = f_t dt + f_x dX_t + \frac{1}{2} f_{xx} dX_t^2 + \dots$$

$$dX_t = u_t dB_t - \frac{1}{2} u_t^2 dt$$

$$dX_t^2 = u_t^2 dB_t^2 + \frac{1}{4} u_t^4 dt^2 - u_t^3 dt dB_t$$

$$df = e^{X_t} dX_t + \frac{1}{2} e^{X_t} dX_t^2$$

$$= e^{X_t} \left[u_t dB_t - \frac{1}{2} u_t^2 dt + \frac{u_t^2}{2} dt \right]$$

$$de^{X_t} = e^{X_t} u_t dB_t$$

$$e^{X_t} - \underbrace{e^{X_0}}_1 = \int_0^t e^{X_s} u_s dB_s$$

$$e^{X_t} = 1 + \underbrace{\int_0^t e^{X_s} u_s dB_s}_{\text{Itô Martingale}}$$

2nd Method

$$E[|M_t|] = E \left[e^{\int_0^t u_s dB_s} \right]^{1/2} E \left[e^{-\int_0^t u_s^2 ds} \right]^{1/2} = e^{-\frac{1}{2} \int_0^t u_s^2 ds}$$

Cauchy Schwartz $E|XY| \leq E X^2 E Y^2$ $\leq e^{-1/2 \int_0^t u_s^2 ds} \left(E e^{2 \int_0^t u_s^2 ds} \right)^{1/2}$

momentum generating

$$E e^{mZ} = e^{\frac{m^2}{2} \sigma^2} \left| N(0, \int_0^t 4u_s^2 ds) \right| = e^{-\frac{1}{2} \int_0^t u_s^2 ds} \left(e^{2 \int_0^t u_s^2 ds} \right)^{1/2} \cdot 2 \cdot \infty$$

$$\mathbb{E}[M_t | \mathcal{F}_s] = e^{\int_0^s u dB_s - \frac{1}{2} \int_0^s u^2 ds} \underbrace{\mathbb{E}\left[e^{\int_s^t u dB_s - \frac{1}{2} \int_s^t u^2 ds} \right]}_{e^{\frac{1}{2} \int_s^t u^2 ds}}$$

because \int^s is adapted, \int^t is \perp of \mathcal{F}_s

now

$$\mathbb{E} e^{\lambda Z} = e^{\lambda M + \frac{1}{2} \lambda^2 \text{var}}^s$$

$$\text{var} = e^{\frac{1}{2} \int_s^t u^2 ds} = e^{\int_0^s u dB_s - \frac{1}{2} \int_0^s u^2 ds}$$

$$2) d(MY) = Y dM + M dY + dY dM$$

$$dY = -u dt + dB$$

$$dM = Mu dB$$

$$= YMu dB + \cancel{(-Mu dt + M dB)}$$

$$+ \cancel{(-Mu^2 dB dt + Mu dt)}$$

drop

$$= MYu dB + M dB$$

$$MY = \int \underbrace{M(Yu + 1)}_{IT\hat{O}} dB$$

Exercise 2.

$$X_t = (B_t^1)^2 + (B_t^2)^2 + (B_t^3)^2$$

$$f(t, B_t^1)$$

$$f(t, x, y, z) = x^2 + y^2 + z^2$$

$$df(t, x, y, z) = f_1 dt + f_2 dx + f_3 dy + f_4 dz$$

$$+ \frac{1}{2} [f_{11} dt^2 + f_{22} dx^2 + f_{33} dy^2 + f_{44} dz^2$$

$$+ 2 f_{12} dt dx + 2 f_{13} dt dy + 2 f_{14} dt dz$$

$$+ 2 f_{23} dx dy + 2 f_{24} dx dz + 2 f_{34} dy dz]$$

$$f_1 = 0 \quad f_2 = 2x \quad f_3 = 2y \quad f_4 = 2z$$

$$f_{12} = 0 \quad f_{22} = 2 \quad f_{33} = 2 \quad f_{44} = 2$$

$$f_{13} = 0 \quad f_{23} = 0 \quad f_{34} = 0$$

$$f_{14} = 0 \quad f_{24} = 0$$

$E(dx dy)$
 $E[(B_{t_i}^1 - B_{t_{i-1}}^1)(B_{t_i}^2 - B_{t_{i-1}}^2)]$
 $= E(\cdot) E(\cdot) = 0$
 drop $dx dy \times dt$

$E[(B_{t_i}^1 - B_{t_{i-1}}^1)(B_{t_i}^2 - B_{t_{i-1}}^2)]^2$
 $= dt \cdot dt \Rightarrow dx dx = dt$
 $= dt^2$

$$dX_t = 0 + 2B_t^1 dB_t^1 + 2B_t^2 dB_t^2 + 2B_t^3 dB_t^3$$

$$+ \frac{1}{2} [0 + 2 dt + 2 dt + 2 dt + 0]$$

$$dX_t = 3 dt + \sum_{i=1}^3 2 B_t^i dB_t^i$$

$$X_t = 3t + \sum_{i=1}^3 \int 2 B_t^i dB_t^i$$

Exercise 3

$$X_t = \int_0^t u \, dB_s^1 + \int_0^t v \, dB_s^2 - \frac{1}{2} \int_0^t (u^2 + v^2) \, ds$$

$$f(t, X_t) = e^{X_t} \quad f_{t=0} = f_2 = e^{X_t} \quad f_{xx} = e^{X_t}$$

$$df = f_1 dt + f_2 dx + \frac{1}{2} [f_{11} dt^2 + f_{22} dx^2 + 2f_{12} dt dx]$$

$$dX_t = u \, dB_t^1 + v \, dB_t^2 - \frac{1}{2} u^2 dt - \frac{1}{2} v^2 dt$$

$$dX_t^2 = (u \, dB_t^1 + v \, dB_t^2)^2 + \left(\frac{1}{2} (u^2 + v^2) dt \right)^2 - 2 \frac{1}{2} (u^2 + v^2) dt (u \, dB_t^1 + v \, dB_t^2)$$

cheap unusefull

$$= u^2 (dB_t^1)^2 + v^2 (dB_t^2)^2 + uv \, dB_t^1 \, dB_t^2$$

$$de^{X_t} = 0 + e^{X_t} \left[u \, dB_t^1 + v \, dB_t^2 - \frac{1}{2} (u^2 + v^2) dt \right]$$

$$+ \frac{1}{2} \left[e^{X_t} \left(u^2 (dB_t^1)^2 + v^2 (dB_t^2)^2 + uv \, dB_t^1 \, dB_t^2 \right) \right]$$

$$de^{X_t} = e^{X_t} \left[-\frac{1}{2} (u^2 + v^2) + \frac{1}{2} (u^2 + v^2 + uv) \right] dt$$

$$+ e^{X_t} (u \, dB_t^1 + v \, dB_t^2)$$

$$de^{X_t} = e^{X_t} (u \, dB_t^1 + v \, dB_t^2) + e^{X_t} uv \, dt$$

Exercise 3 2nd way

$$E[M_t | \mathcal{F}_s] = E \left[e^{\int_0^t u dB_1} \mid \mathcal{F}_s \right]$$

$$\rightarrow E \left[e^{\int_0^t v dB_2} \mid \mathcal{F}_s \right]$$

$$\rightarrow E \left[e^{-\frac{1}{2} \int_0^t (u^2 + v^2) ds} \mid \mathcal{F}_s \right]$$

$$= e^{\int_0^s u dB_1} e^{\int_0^s v dB_2} e^{-\frac{1}{2} \int_0^t (u^2 + v^2) ds}$$

because

$$\int_0^s \subset \mathcal{F}_s$$

$$\int_s^t \perp \mathcal{F}_s$$

$$= \underbrace{E \left[e^{\int_s^t u dB_1} \right]}_{\text{momentum generators}} \underbrace{E \left[e^{\int_s^t v dB_2} \right]}_{\text{momentum generators}}$$

momentum generators

$$e^{+\frac{1}{2} \int_s^t u^2 ds} e^{\frac{1}{2} \int_s^t v^2 ds}$$

$$= e^{\int_0^s u dB_1} e^{\int_0^s v dB_2}$$

$$e^{-\frac{1}{2} \int_0^s u^2 ds} + \frac{1}{2} \int_s^t u^2 ds$$

$$e^{-\frac{1}{2} \int_0^s u^2 ds}$$

$$e^{-\frac{1}{2} \int_0^s v^2 ds} + \frac{1}{2} \int_s^t v^2 ds$$

$$e^{-\frac{1}{2} \int_0^s v^2 ds}$$

$$= M_s$$

Exercise 4

$$\underbrace{\int_0^t u dB}_X \quad \underbrace{\int_0^t v dB}_Y - \int_0^t uv ds$$

$$dX = u dB$$

$$dY = v dB$$

$$dXdY = uv dt$$

$$d(XY) = YdX + XdY + dXdY$$

$$d(XY) - dXdY = YdX + XdY$$

$$\int XY - \int dXdY = \underbrace{\int YdX}_{Ito} + \underbrace{\int XdY}_{Ito}$$

each one must be well defined

$$E[I^2] = E\left[\left(\int_0^t Y u dB\right)^2\right] = E\left[\int_0^t Y^2 u^2 dt\right]$$

$$= E\left[\int_0^t \int_0^t v^2 u^2 dt ds\right]$$

Easy way
 $(a^2 - b^2)^2 \geq 0$

$$(a+b)^2 = a^2 + b^2 + 2ab$$

$$(a-b)^2 = a^2 + b^2 - 2ab$$

$$4ab = (a+b)^2 - (a-b)^2$$

$$4a^2b^2 = (a^2 + b^2)^2 - (a^2 - b^2)^2$$

and

$$(a+b)^2 + (a-b)^2 = 2a^2 + 2b^2$$

$$\begin{aligned} & \cancel{4ab} \\ & a^4 - 2a^2b^2 + b^4 \geq 0 \\ & a^4 + b^4 \geq 2a^2b^2 \\ & \sqrt{(a^2 + b^2)^2 + (a^2 - b^2)^2} \\ & = 2a^2 + 2b^2 \end{aligned}$$

$$\Rightarrow a^2b^2 \leq \frac{a^4 + b^4}{2}$$

Hard way

Exercise 4

$$X_t = \underbrace{\left(\int_0^t u dB \right)}_{M_1} \underbrace{\left(\int_0^t v dB \right)}_{M_2} - \int_0^t uv ds$$

in worksheet 4 exercise 4

if $B_1 \perp B_2$

then $X = B_1 B_2$ is Martingale.

$$d(XY) = X dY + Y dX + dX dY$$

$$X = \int u dB \quad dX = u dB$$

$$Y = \int v dB \quad dY = v dB$$

$$d \left(\int u dB \int v dB \right) - (u dB)(v dB) = X v dB + Y u dB$$

$$\int_0^t u dB \int_0^t v dB - \int_0^t uv ds = \underbrace{\int_0^t X v dB}_{ITB} + \underbrace{\int_0^t Y u dB}_{ITB}$$

each well defined

$$E \left[\int_0^t X v dB \right]^2 = E \left[\int_0^t X^2 v^2 ds \right]$$

$$= E \left[\int_0^t \int_0^t u^2 v^2 ds dr \right]$$

Exercise 6.

$$X_t = e^{\int_0^t f dB_s}$$

$$1) \quad I = \int_0^t f_s dB_s \sim N(0, \int_0^t f_s^2 ds)$$

$$E[e^{\lambda N(0, \sigma^2)}] = e^{\lambda \mu + \frac{1}{2} \sigma^2 \lambda^2}$$

Momentum
Generator

$$\Rightarrow E\left[e^{\int_0^t f dB_s}\right] = e^{\frac{1}{2} \int_0^t f^2 ds}$$

$$\begin{aligned} E|X_t| &\leq (E X_t^2)^{1/2} = \left(E \left[e^{2 \int_0^t f dB_s} \right]\right)^{1/2} \\ &= \left(E \left[e^{2 N(0, \int_0^t f^2 ds)} \right]\right)^{1/2} \\ &= e^{\frac{1}{2} \int_0^t f^2 ds} < \infty \end{aligned}$$

$$\begin{aligned} s \leq t \\ E[X_t | \mathcal{F}_s] &= E\left[e^{\int_0^t f dB - \frac{1}{2} \int_0^t f^2 ds} \mid \mathcal{F}_s \right] \\ &= e^{\int_0^s f dB - \frac{1}{2} \int_0^s f^2 ds} E\left[e^{\int_s^t f dB} \mid \mathcal{F}_s \right] \\ &= \underbrace{\hspace{10em}}_{\text{Ind of } \mathcal{F}_s} E\left[e^{\int_s^t f dB} \right] \\ &= e^{\int_0^s f dB - \frac{1}{2} \int_0^s f^2 ds} e^{\frac{1}{2} \int_s^t f^2 ds} \\ &= e^{\int_0^s f dB - \frac{1}{2} \int_0^t f^2 ds} = X_s \end{aligned}$$

Exercise 5

$$u(T-t, x+B_t)$$

$$\begin{aligned}
du &= -u_1 dt + u_2 dx + \frac{1}{2} u_{22} dx^2 \\
&\quad + u_3 dy + \frac{1}{2} u_{33} dy^2 \\
&\quad + u_{23} dx dy
\end{aligned}$$

$$= u_2 dx + u_3 dy + \left[-u_1 dt + \frac{1}{2} u_{22} dx^2 + \frac{1}{2} u_{33} dy^2 + u_{23} dx dy \right]$$

as $B^1 \perp B^2 \quad dx dy = 0$

$$d(u(T-t, x+B_t)) = \frac{\partial u}{\partial x_1} dB_t^1 + \frac{\partial u}{\partial x_2} dB_t^2 + \underbrace{0}_{\text{PDE} = 0}$$

now $\int_0^T \Rightarrow$ Martingale.

~~$u(T-t, x+B_t) = u(T, x)$~~

$$\underbrace{u(0, x+B_T)}_{h(0, x+B_T)} = \underbrace{u(T, x)}_{\text{known number}} + \int_0^T \frac{\partial u}{\partial x_i} dB_t^i$$

$$\mathbb{E} h(x+B_T) = u(T, x) + \hat{\mathbb{E}} \text{ Ito } 0$$

Exercise 9) $dX_t = a dt + \theta dB_t$

$\exists u$ s.t. $d\tilde{B}_t = dB_t + u dt$.

$$= a dt + \theta [d\tilde{B}_t - u dt]$$

$$= [a - u\theta] dt + \theta d\tilde{B}_t.$$

Now given that $u = \frac{r}{\sigma}$

$$\Rightarrow dY_t = Y_t \mu dt + \sigma Y_t dB_t.$$

$$= Y_t \mu dt + \sigma Y_t [d\tilde{B}_t - \frac{r}{\sigma} dt]$$

$$= Y_t \left[\mu - \frac{r}{\sigma} \right] dt + \sigma Y_t d\tilde{B}_t$$

$$\Rightarrow \underline{\mu = r} \text{ so } Y_t \text{ is Mart}$$

we get

$$\frac{dQ}{dP} = M_t = e^{-\int_0^t \frac{r}{\sigma} dB_t - \frac{1}{2} \int_0^t \frac{r^2}{\sigma^2} dt}$$

$$= e^{-\int_0^t \frac{\mu}{\sigma} dB_t - \frac{1}{2} \int_0^t \frac{\mu^2}{\sigma^2} dt}.$$

and Y_t is Mart. under Q .

$$10) B_t + B_t^3 = f(B_t)$$

$$f(x) = x + x^3$$

$$f_1 = 0$$

$$f_2 = 1 + 3x^2$$

$$f_{22} = 6x$$

$$d(B_t + B_t^3) = (1 + 3B_t^2) dB_t + \frac{1}{2} 6B_t dt$$

$$\Rightarrow B_t + B_t^3 = \underbrace{\int_0^t 3B_s ds}_{\text{drift}} + \underbrace{\int_0^t (1 + 3B_s^2) dB_s}_{\text{Ito Martingale}}$$

$$dX_t = a dt + \theta dB_t$$

$$d\tilde{B}_t = dB_t + u dt \quad \text{defines } M_t \text{ and } \mathbb{Q}$$

$$\Rightarrow dX_t = a dt + \theta [d\tilde{B}_t - u dt]$$

$$= [a - u\theta] dt + \theta d\tilde{B}_t$$

choose u st $a - u\theta = 0 \Rightarrow dX_t$ is Ito (Martingale) in \mathbb{Q}

$$u = \frac{a}{\theta} = \frac{2(1+B_t^2)}{3B_t + 1 + 3B_t^2}$$

is ok because denominator is ≥ 0 .

$$\Rightarrow M_t = e^{-\int_0^t u dB_s - \frac{1}{2} \int_0^t u^2 ds}$$

$$\begin{aligned} \text{check well defined } \int_0^t u^2 ds &\Rightarrow \int_0^t \mathbb{E} \left[\frac{3B_s}{1+3B_s^2} \right]^2 ds \\ &= \int_0^t \mathbb{E} \frac{9B_s^2}{(1+3B_s^2)^2} ds \end{aligned}$$

Girsanov 2

$$\text{Let } dY_t = \beta_t dt + \theta_t dB_t$$

$$d\tilde{B}_t = dB_t + u_t dt \quad \Rightarrow \quad M_t = e^{-\int u dB - \frac{1}{2} \int u^2 ds}$$

$$dY_t = \beta_t dt + \theta_t [d\tilde{B}_t - u_t dt]$$

$$= \beta_t - u_t \theta_t dt + \theta_t d\tilde{B}_t$$

$$= \alpha_t dt + \theta_t d\tilde{B}_t$$

$$\alpha = \beta - u\theta$$

Example

$$dX_t = \alpha(X_t) dt + dB_t$$

$$= \alpha(X_t) dt + [d\tilde{B} - u_t dt]$$

$$= (\alpha - u) dt + d\tilde{B}_t$$

$$u = -\alpha(X_t)$$

$$\text{define } M_t = \exp \left\{ - \int u dB - \frac{1}{2} \int u^2 ds \right\}$$

$$= \exp \left\{ \int \alpha dB - \frac{1}{2} \int \alpha^2 ds \right\}$$

$$\tilde{B} + \int \alpha dt \text{ is BM in } \mathbb{R}$$

Exercise 7

$$X_T = \exp\left(\sum_i^3 b_i B_t^i - \frac{1}{2} \sum_i^3 b_i^2 t\right)$$

$$X_T^2 = \exp\left(2 \sum_i^3 b_i B_t^i - \sum_i^3 b_i^2 t\right)$$

$$E[X_T^2] = \exp(-\sum_i^3 b_i^2 t) E\left[\exp\left(2 \sum_i^3 b_i B_t^i\right)\right]$$

$$= \exp(-\sum_i^3 b_i^2 t) \prod_i^3 E \exp(2b_i B_t^i)$$

1st $B_t = t$

$$= \exp(-\sum_i^3 b_i^2 t) \prod_i^3 \underbrace{E \exp(2b_i B_t^i)}_{\text{Momentum Generator}} = \exp\left(-\sum_i^3 b_i^2 t\right) \prod_i^3 \exp\left(\frac{1}{2} t (2b_i)^2\right)$$

$$= \prod_i^3 \exp(-b_i^2 t) \prod_i^3 \exp(2t b_i^2)$$

$$= \prod_i^3 \exp(b_i^2 t)$$

Girsanov theorem (Itô)

Let $u_t dW_t + dB_t = dY_t$ an Itô process in \mathbb{P}

then $M_t = e^{-\int_0^t u dB_s - \frac{1}{2} \int_0^t u^2 ds}$ is a Martingale in \mathbb{P}

s.t. $\frac{d\mathbb{Q}}{d\mathbb{P}} = M_t$.

and in \mathbb{Q} Y_t is an RBM

Exercise 10

$$B_t^3 - B_t = Y_t \quad \int_0^t dY_t = \int_0^t \frac{1}{2} (3B_t) dt + \underbrace{\int_0^t (3B_t^2 - 1) dB_t}_{\text{Itô}}$$
$$Y_t = 3 \int_0^t B_t dt + B_t + \int_0^t 3B_t^2 dB_t$$

Lipschitz Condition

$$|b(t, x) - b(t, y)| + |\sigma(t, x) - \sigma(t, y)| \leq L |x - y|$$

- Every where differentiable functions with bounded 1st derivative are Lipschitz Cont.

Example e^x, x^2 not in \mathbb{R}
yes in compact set $[0, T]$

Growth

$$|b(t, x)| \leq K \cdot (1 + |x|)$$

All linear SDE

$$b(t, x) = c_1(t)x + c_2(t)$$

$$\begin{aligned} |b(t, x)| &= |c_1(t)x + c_2(t)| \leq \left| \sup_t (c_1, c_2) x + \sup_t (c_1, c_2) \right| \\ &= \left| \sup_t (c_1, c_2) \right| |1 + x| \end{aligned}$$

$$\left| \sup_t c_1, c_2 \right| < \infty$$

if c_1, c_2 continuous and the set is compact $[0, T]$

Worksheet 9

Growth condition $|b(t, x)| + |\sigma(t, x)| \leq C(1 + |x|)$

Exercise 1

$$dX_t = b X_t ds + \sigma X_t d\beta_s$$

a, b must be continuous in the compact set $[0, T]$ and Lipschitz

i) $dX_t = a(s, X_s) ds + b(s, X_s) d\beta_s$

$a = b x$ $b = \sigma x$
 continuous linear functions
 reparametrized as $f(x)$

Lipschitz $|a(t, x) - a(t, y)| + |b(t, x) - b(t, y)| \leq K |x - y|$

$\Rightarrow |bx - by| + |\sigma x - \sigma y| = (|b| + |\sigma|) |x - y|$

ii) $f = x_0 e^{(b - \frac{\sigma^2}{2})t + \sigma \beta_t}$ $f_1 = (b - \frac{\sigma^2}{2}) f$
 $f_2 = \sigma f$ $f_{22} = \sigma^2 f$

$$dX_t = (b - \frac{\sigma^2}{2}) f dt + \sigma f d\beta_t$$

$$= b f + \sigma f d\beta_t$$

$$= b X_t + \sigma X_t d\beta_t$$

Another Method is to consider $d \ln X_t$

Exercise 2

$$dX_t = bX_t ds + dB_t$$

$$dX_t = a(s, X_s) ds + b(s, X_s) dB_t$$

$$a(s, X_s) = bX_s \quad b(s, X_s) = 1$$

continuous coefficients a, b

see bounded increments

a, b

continuous
bounded
derivatives
 \Rightarrow Lipschitz

and

cont bounded

\Rightarrow growth

Lipschitz:

$$|a(t, x) - a(t, y)| + |b(t, x) - b(t, y)|$$

$$= |bx - by| + |1 - 1| = |b| |x - y|$$

so unique strong solution on $[0, T]$

$$|bx| + 1 = |b||x| + 1 \leq |b||x| + |b| \leq |b|(|x| + 1)$$

$$f = X_t = x_0 + \int_0^t bX_s ds + B_t$$

~~$$f = x_0 + \int_0^t bX_s ds + B_t$$~~

~~$$dX_t = bX_t dt + dB_t$$~~

$$dX_t = bX_t dt + dB_t$$

$$dI = -bI dt$$

$$dX_t - bX_t dt = dB_t$$

$$I = e^{-bt}$$

$$I dX_t - bIX_t dt = I dB_t$$

$$\Rightarrow X_t = e^{bt} X_0$$

$$d(IX_t) = I dB_t$$

$$+ e^{bt} \int_0^t e^{-bs} dB_s$$

$$e^{-bt} X_t - e^{-b0} X_0 = \int_0^t e^{-bs} dB_s$$

Exercise 3

$$X_t = x + \int_0^t \sin(2X_s) ds + \int_0^t (1 + \cos^2(X_s))^{1/2} dB_s$$

$$dX_t = \sin(2X_t) dt + (1 + \cos^2(X_t))^{1/2} dB_t$$

$$dX_t = a(s, X_s) ds + b(s, X_s) dB_s$$

bounded in t

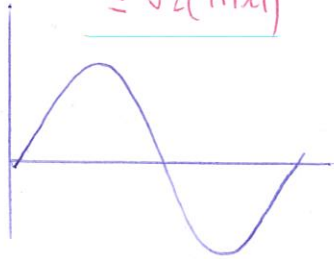
$\sup_t < \infty \Rightarrow$ growth,

$$\sin 2x, (1 + \cos^2(x))^{1/2}$$

are cont diff bounded derivative \Rightarrow Lipschitz

$$|\sin 2x| \leq 1 \leq 1 + |x| \quad | \cos^2 x | \leq 1 \leq 1 + |x|$$

$$|\sin 2x - \sin 2y| + | (1 + \cos^2 x)^{1/2} - (1 + \cos^2 y)^{1/2} | \leq \sqrt{2} (|x-y|)$$



$\sin x$ is bounded

$\cos^2 x$ is bounded

\sqrt{x} is bounded in a compact $[0, T]$

$$\Phi = \int_0^x (1 + \cos^2(u))^2 du \quad \Phi'_2 = 2(1 + \cos^2 u) \quad \Phi'_1 = [-2 \cos u \sin u]$$

$$df(X_t) = f_1 dt + f_2 dX_t + \frac{1}{2} f_{22} dX_t^2 + \dots$$

$$= 0 + [1 + \cos(X_t)]^2 dX_t + \frac{1}{2} [2 \cos u \sin u] dX_t^2$$

$$= [1 + \cos X_t]^2 [\sin 2X_t dt + (1 + \cos^2 X_t)^{1/2} dB_t]$$

$$+ \frac{1}{2} 2 [1 + \cos^2 X_t] [-2 \cos X_t \sin X_t] [1 + \cos^2 X_t dt]$$

$$d\Phi(X_t) = (1 + \cos X_t)^{5/2} dB_t$$

Exercise 4

$$dX_t = \frac{-\sin t}{2 + \cos t} X_t dt + (2 + \cos t) dB_t.$$

$$a(t, x) = \frac{-\sin t}{2 + \cos t} x$$

$$b(t, x) = 2 + \cos t.$$

are continuous functions

$$\sup_t (a, b) < \infty$$

$$2 + \cos t \neq 0$$

positive



$$|a(t, x) - a(t, y)| + |b(t, x) - b(t, y)|$$

$$= \left(\underbrace{\left| \frac{-\sin t}{2 + \cos t} \right|}_{\leq \frac{1}{2}} + \underbrace{|2 + \cos t|}_{\leq 3} \right) |x - y|$$

$$\leq \frac{1}{2}$$

$$\leq 3$$

$$X_t = (2 + \cos t) B_t. \quad f(t, B_t) = (2 + \cos t) x.$$

$$(f_1 + \frac{1}{2} f_{22}) dt + f_2 dB_t = df.$$

$$\left(-\sin(t) B_t + \frac{1}{2} \cdot 0 \right) dt + (2 + \cos t) dB_t.$$

$$dX_t = -\sin t B_t dt + (2 + \cos t) dB_t$$

~~$$X_t = \frac{X_t}{2 + \cos t} (2 + \cos t)$$~~

~~$$\frac{X_t}{2 + \cos t} = X_t$$~~

$$dX_t = \frac{-\sin t}{2 + \cos t} \underbrace{(2 + \cos t) B_t}_{X_t} dt + (2 + \cos t) dB_t$$

Exercise 5.

$$i) \quad X_t = x + \int_0^t s X_s ds + \int_0^t e^s dB_s$$

$$dX_t = tX_t dt + e^t dB_t \\ = a(t, X_t) dt + b(t, X_t) dB_t$$

$$a(t, x) = tx$$

$$b(t, x) = e^t$$

continuous and bounded in compact $[0, T]$

$$|a(t, x) - a(t, y)| + |b(t, x) - b(t, y)|$$

$$= |tx - ty| + |e^t - e^t| \leq T|x - y| \quad \text{Lipschitz.}$$

so $\exists!$ strong solution.

$$ii) \quad y_t = 1 + \int_0^t sy_s ds \quad dy_t = ty_t dt \\ y_t = e^{t^2/2} \quad y_0 = 1$$

$$d(y^{-1}X) = X d(y^{-1}) + y^{-1}dX$$

$$= -X \frac{dy}{y^2} + \frac{1}{y} dX$$

$$= -e^{-t^2/2} X dy + e^{-t^2/2} dX$$

$$= -e^{-t^2} t e^{t^2/2} X dt + e^{-t^2/2} [tX dt + e^t dB_t]$$

$$= -te^{-t^2/2} X dt + te^{-t^2/2} X dt - e^{t^2/2} dB_t$$

$$= -e^{t^2/2} dB_t \quad \Rightarrow \quad \frac{X}{y} \text{ is Martingale.}$$

$$\text{iii) } d\left(\frac{X}{y}\right) = -e^{t^2/2} dB_t$$

$$X_t = e^{t^2/2} X_0 - e^{t^2/2} \int_0^t e^{-s^2/2} dB_s$$

$$\text{iv) } N\left(e^{t^2/2} X_0, e^{2t^2/2} \int_0^t e^{2(t-s)^2/2} ds\right)$$

Exercise 6

$$M_t = 1 - \int_0^t s^{1/2} M_s dB_s$$

$$dM_t = -t^{1/2} M_t dB_t$$

$$M_t = f(t, B_t)$$

$$dM_t = f_1 dt + f_2 dB_t + \frac{1}{2} f_{22} dt + \dots$$

$$\begin{cases} f_1 + \frac{1}{2} f_{22} = 0 \\ f_2 = -t^{1/2} f. \end{cases} \Rightarrow f_{22} = t f.$$

$$\Rightarrow f_1 + \frac{t}{2} f = 0.$$

$$X' T + \frac{t}{2} X T = 0$$

$$\frac{T'}{T} + \frac{t}{2} = 0$$

$$\frac{T'}{T} = -\frac{t}{2}$$

$$T = T_0 e^{-t^2/4}$$

$$\Rightarrow X' T = -t^{1/2} X T$$

$$\frac{X'}{X} = -t^{1/2}$$

$$\frac{dX}{X} = -t^{1/2} dt$$

$$X = X_0 e^{-\int t^{1/2} dt}$$

$$e^{-\int t^{1/2} dt}$$

$$X = X_0$$

$$F = M_t = X T = M_0 e^{-\int_0^t s^{1/2} ds} = M_0 e^{-\frac{t^2}{4}}$$

$$\text{ii) } dM_t = -t^{1/2} M_t dB_t \\ = a(t, M_t) dt + b(t, M_t) dB_t$$

$$a=0 \quad \text{~~bounded in } [0, T]~~$$

$$b(t, x) = t^{1/2} x.$$

cont. bounded in $[0, T]$

$$|b(t, x) - b(t, y)| = |t^{1/2}(x-y)| \leq T^{1/2}|x-y| \\ \Rightarrow \exists! \text{ solution.}$$

$$d \ln(M_t) = 0 dt + \frac{1}{M_t} dM_t + \frac{1}{2} \left(\frac{-1}{M_t^2} \right) dM_t^2 + \dots$$

$$= \frac{1}{M_t} [-t^{1/2} M_t dB_t] - \frac{1}{2M_t^2} [t M_t^2 dt]$$

$$d \ln(M_t) = -t^{1/2} dB_t - \frac{t}{2} dt.$$

$$\ln M_t - \ln M_0 = \int_0^t -t^{1/2} dB_t - \frac{t^2}{4}$$

$$M_0 = 1$$

$$\ln M_t = 1 - \frac{t^2}{4} - \int_0^t t^{1/2} dB_t$$

$$M_t = M_0 e^{-t^2/4 - \int_0^t t^{1/2} dB_t}$$

Exercise 6 Continuation

$$dX_t = \cos t dt + s^{1/2} X \downarrow B'_s$$

$$\downarrow M_t = -s^{1/2} M_t \downarrow B_t^2$$

$$\downarrow (e^{s^{1/2}} X_t M_t) = e^{s^{1/2}} M_t dX + e^{s^{1/2}} X_t dM + e^{s^{1/2}} dM dX$$

$$= e^{s^{1/2}} M [\cos t dt + s^{1/2} X \downarrow B'_s]$$

$$+ e^{s^{1/2}} X [-s^{1/2} M \downarrow B_t^2]$$

$$+ e^{s^{1/2}} [\cos t dt + s^{1/2} X \downarrow B'_s] [-s^{1/2} M \downarrow B_t^2]$$

$$= e^{s^{1/2}} M [\cos t dt]$$

$$+ e^{s^{1/2}} M X s^{1/2} [\downarrow B'_t - \downarrow B_t^2]$$

$$+ e^{s^{1/2}} [-s^{1/2} M \cos t dt \downarrow B_t^2]$$

$$= s M X \downarrow B'_t \downarrow B_t^2]$$

$$\int e^{s^{1/2}} X_t M_t = \int e^{s^{1/2}} M \cos t dt$$

$$B' \perp B^2 \\ \downarrow B' \downarrow B^2 = 0$$

$$+ \int e^{s^{1/2}} M X s^{1/2} \downarrow B'_t$$

$$- \int e^{s^{1/2}} M X s^{1/2} \downarrow B_t^2 = \int e^{s^{1/2}} M \cos t dt$$

Exercise 7

$$X_t = 1 - \frac{1}{2} \int_0^t X_s ds - \int_0^t Y_s dB_s$$

$$Y_t = -\frac{1}{2} \int_0^t Y_s ds + \int_0^t X_s dB_s$$

$$dX_t = -\frac{X_t}{2} dt - Y_t dB_t$$

$$dY_t = -\frac{Y_t}{2} dt + X_t dB_t$$

$$f(t, x) = x^2$$

$$df = f_1 dt + f_2 dx + \frac{1}{2} f_{22} dx^2 + \dots$$

$$= 0 + 2x dx + \frac{1}{2} 2 dx^2$$

$$(dX_t)^2 = Y_t^2 dt + \dots$$

$$(dY_t)^2 = X_t^2 dt + \dots$$

$$d(X_t^2) = 2X_t dX_t + d(X_t^2)$$

$$d(Y_t^2) = 2Y_t dY_t + d(Y_t^2)$$

$$\bullet d(X_t^2) = 2X_t \left[-\frac{X_t}{2} dt - Y_t dB_t \right] + Y_t^2 dt$$

$$= (-X_t^2 + Y_t^2) dt - 2X_t Y_t dB_t$$

$$\bullet d(Y_t^2) = 2Y_t \left[-\frac{Y_t}{2} dt + X_t dB_t \right] + X_t^2 dt$$

$$= -Y_t^2 + X_t^2 dt + 2X_t Y_t dB_t$$

$$d(X_t^2) + d(Y_t^2) = 0$$

$$X_t^2 - X_0^2 + Y_t^2 - Y_0^2 = 0$$

$$X_t^2 + Y_t^2 = X_0^2 + Y_0^2$$

$$= 1 \quad \checkmark$$

$$X_0 = 1$$

$$Y_0 = 0$$

$$\text{iii) } X_t = \cos \theta_t$$

$$dX_t = d \cos \theta_t = -\sin \theta_t d\theta_t + \frac{1}{2} \cos \theta_t d\theta_t^2$$

$$= -Y_t d\theta_t - \frac{1}{2} X_t d\theta_t^2$$

$$= -Y_t dB_t - \frac{X_t}{2} dt$$

$$\Rightarrow d\theta_t = dB_t \quad \#$$

$$\Rightarrow X_t = \cos \beta_t$$

$$Y_t = \sin \beta_t$$

Exercises Linear SDE

For a Unique solution (strong)

$$\bar{E} X_0^2 < \infty$$

$a(t, x), b(t, x)$ continuous

Linear

$$a = c_1 x + c_2$$

$$|a(t, x) - a(t, y)| + |b(t, x) - b(t, y)|$$

$$b = \sigma_1 x + \sigma_2$$

$$\leq k |x - y|$$

continuous

Lipschitz in 2nd variable

$$|c_1 x + c_2 - (c_1 y + c_2)| + |\sigma_1 x + \sigma_2 - (\sigma_1 y + \sigma_2)|$$

$$= |c_1 (x - y)| + |\sigma_1 (x - y)| \leq \underbrace{(|c_1| + |\sigma_1|)}_{\text{finite}} |x - y|$$

Every linear has unique strong solution

finite

Geometric Brownian Motion

(Linear, Multiplicative Noise, Homogeneous)
constant coefficients.

Method 1

$$dX_t = cX_t dt + \sigma X_t dB_t$$

propose $X_t = f(t, B_t)$ unknown

$$df = (f_1 + \frac{1}{2} f_{22}) dt + f_2 dB_t.$$

$$\begin{cases} f_1 + \frac{1}{2} f_{22} = cf \\ f_2 = \sigma f \end{cases}$$

\Rightarrow

$$f_{22} = \sigma^2 f.$$

$$\begin{cases} f_1 = (c - \frac{1}{2} \sigma^2) f \\ f_2 = \sigma f \end{cases}$$

separation of variables $f = T(t) h(B_t)$

$$\dot{T}h = (c - \frac{1}{2}\sigma^2)Th$$

$$T = T_0 e^{(c - 0.5\sigma^2)t}$$

$$Th' = \sigma Th$$

$$h = h_0 e^{\sigma B_t}$$

$$\Rightarrow f = Th = X_0 e^{(c - 0.5\sigma^2)t + \sigma B_t}$$

$$X_t = X_0 e^{(c - 0.5\sigma^2)t + \sigma B_t}$$

Method 2.

$$dX_t = cX_t dt + \sigma X_t dB_t$$

$$\frac{dX_t}{X_t} = c dt + \sigma dB_t$$

change variable

$$f(t, X_t) = \ln(X_t)$$

$$df = f_1 dt + f_2 dX_t$$

$$+ \frac{1}{2} [f_{11} dt^2 + f_{22} dX_t^2 + 2f_{12} dt dX_t]$$

~~dX_t/X_t~~

$$f_1 = 0 \quad f_2 = \frac{1}{X_t} \quad f_{22} = -\frac{1}{X_t^2}$$

$$d \ln X_t = \frac{1}{X_t} [cX_t dt + \sigma X_t dB_t] + \frac{1}{2} \left(-\frac{1}{X_t^2}\right) [\sigma^2 X_t^2 dt]$$

$$d \ln X_t = \left(c - \frac{\sigma^2}{2}\right) dt + \sigma dB_t$$

with this change of variables is now Additive Noise
any term depend on the unknown $\ln X_t$.

$$\ln \frac{X_t}{X_0} = \left(c - \frac{\sigma^2}{2}\right)t + \sigma B_t$$

$$X_t = X_0 e^{\left(c - \frac{\sigma^2}{2}\right)t + \sigma B_t}$$

Careful when integration of f(g)

Chain Rule Ito Lemma
Watch out!! X_t depends on B_t

$$d \ln X_t \neq \frac{dX_t}{X_t}$$

needs 2nd order corrections

~~$$d \ln X_t = \frac{1}{X_t} dX_t + \frac{1}{2} \frac{d^2 X_t}{X_t^2}$$~~

for example

$$d \ln B_t = \frac{dB_t}{B_t} - \frac{(dB_t)^2}{2(B_t)^2}$$

Ornstein Uhlenbeck (Langevin)

(homogeneous, Additive noise,
constant coefficients, linear)

$$dX_t = cX_t dt + \sigma dB_t$$

$$\Delta t = 1 \quad X_{t+1} - X_t = cX_t + \sigma (B_{t+1} - B_t)$$

$$X_{t+1} = (c+1)X_t + \underbrace{\sigma (B_{t+1} - B_t)}_{N(0,1)} \quad \text{Autoregressive order 1.}$$

for 2 Ito Processes X, Y

$$d(XY) = YdX + XdY + dXdY$$

and ~~keep~~ drop 2nd order. ↗

Integrating factor

$$YdX_t - cYX_t dt = Y\sigma dB_t$$

if Y doesn't have random part we recover $d(XY) = YdX + XdY$
and

$$d(YX_t) = Y\sigma dB_t$$

$$Y = e^{-ct} + o(B_t)$$

$$X_t = e^{ct} \int_0^t e^{-ct} \sigma dB_t + X_0 e^{ct}$$

X_t is Gaussian because $\Delta_i B$ is gaussian, $e^{-ct_{i-1}}$ is fixed
Suppose $X_0 = 0$ and S_n is sum of independent gaussians

$$S_n = \sum_{i=1}^n e^{-ct_{i-1}} \Delta_i B$$

$$\mu_n = E[S_n] = 0$$

$$\text{var}(S_n) = \sum_{i=1}^n e^{-2ct_{i-1}} \Delta_i t$$

$$\lim_{n \rightarrow \infty} \mu_n = 0$$

$$\lim_{n \rightarrow \infty} \text{var}(S_n) = \int_0^t e^{-2ct} dt$$

$$\Rightarrow X_t \xrightarrow[n \rightarrow \infty]{L} N\left(0, \sigma^2 \int_0^t e^{-2ct} dt\right)$$

covariance $\text{cov}(X_s, X_t)$

$$= \text{cov} \left(\sigma e^{ct} \int_0^t e^{-cu} du, \sigma e^{cs} \int_0^s e^{-cu} du \right)$$

$$= \sigma^2 e^{c(t+s)} \text{cov} \left(\int_s^t e^{-cu} du + \int_0^s e^{-cu} du, \int_0^s e^{-cu} du \right)$$

$\int_s^t \parallel \int_0^s$

$$= \sigma^2 e^{c(t+s)} \text{var}(X_t)$$

$$= \frac{\sigma^2}{2c} \left[e^{c(t+s)} - e^{c(t-s)} \right]$$

Exercise 8.12

2 independent driving Brownian Motions

$$\cancel{X_t} \neq \cancel{X_t} *$$

$$dX_t = cX_t dt + \sigma_1 X_t dB_t^{(1)} + \sigma_2 X_t dB_t^{(2)}$$

Recall $\tilde{B}_t = \frac{1}{(\sigma_1^2 + \sigma_2^2)^{1/2}} (\sigma_1 B_t^{(1)} + \sigma_2 B_t^{(2)})$ $B_t^{(1)} \perp B_t^{(2)}$

$$dX_t = cX_t dt + X_t d[\sigma_1 B_t^{(1)} + \sigma_2 B_t^{(2)}]$$

$$= cX_t dt + X_t d[(\sigma_1^2 + \sigma_2^2)^{1/2} \tilde{B}_t]$$

$$dX_t = cX_t dt + (\sigma_1^2 + \sigma_2^2)^{1/2} X_t d\tilde{B}_t$$

GBM

$$X_t = X_0 e^{[c - 0.5(\sigma_1^2 + \sigma_2^2)^{1/2}]t + (\sigma_1^2 + \sigma_2^2)^{1/2} \tilde{B}_t}$$

Vasicek interest rate

$$dr_t = c[M - r_t]dt + \sigma dB_t$$

$$dr_t - c(M - r_t)dt = \sigma dB_t$$

$$I dr_t - I c(M - r_t)dt = \sigma dB_t I$$

$$I dr_t + I c(r_t - M)dt = \sigma I dB_t$$

$$I d(r_t - M) + I c(r_t - M)dt = \sigma I dB_t.$$

I deterministic

$$d(I(r_t - M)) = \sigma I dB_t$$

$$dI = I c dt \quad I = e^{ct}.$$

$$r_t - M = e^{-ct} \int_0^t I_0' (r_0 - M) + e^{-ct} \sigma \int_0^t e^{ct} dB_t$$

$$r_t = r_0 e^{-ct} + M(1 - e^{-ct}) + \sigma e^{-ct} \int_0^t e^{ct} dB_t.$$

for $r_0 = \text{constant}$

$\mu_0 = c \Rightarrow$ Langevin

$$E(r_t) = r_0 e^{-ct} + M(1 - e^{-ct})$$

$$\text{var}(r_t) = \frac{\sigma^2}{2c} (1 - e^{-2ct}) ?$$

Example 2.3.5 Itô Exponential $f - f_0 = \int_0^t f dB_t$

$$dX_t = X_t dB_t. \quad X_t = f(t, B_t)$$

$$df(t, B_t) = f_1 + \frac{1}{2} f_{22} dt + f_2 dB_t$$

$$\begin{cases} f_1 + \frac{1}{2} f_{22} = 0 \\ f_2 = f \end{cases} \Rightarrow f_{22} = f_2 = f.$$

$$f_1 + \frac{1}{2} f_{22} = 0$$

$$\dot{t}h + \frac{1}{2} h'' h^2 \Gamma = 0$$

$$+K = \frac{\dot{f}}{f} + \frac{1}{2} \frac{f''}{f} + \frac{1}{2} \frac{f'^2}{f^2} + \frac{K}{f}$$

$$\frac{f''}{f} + \frac{f'^2}{f^2} - K = 0$$

$$-2 \dot{t}h = h'' \Gamma$$

$$-2 \frac{\dot{t}}{t} = \frac{h''}{h} = K.$$

$$t = t_0 e^{-\frac{K}{2} t}$$

$$h = h_0 e^{Kx}$$

$$f = f_0 e^{Kx - \frac{K}{2} t}.$$