

Ex 1 (1) Apply Itô's formula to $\Phi(t) \cdot \left[x + \int_0^t \Phi^{-1}(s) a(s) ds + \int_0^t \Phi^{-1}(s) \sigma(s) dW_s \right]$

$$\begin{aligned} dZ_t &= d\Phi(t) \left[x + \int_0^t \Phi^{-1}(s) a(s) ds + \int_0^t \Phi^{-1}(s) \sigma(s) dW_s \right] \\ &\quad + \Phi(t) \left[\Phi^{-1}(t) a(t) dt + \Phi^{-1}(t) \sigma(t) dW_t \right] \\ &= b(t) \Phi(t) \left[x + \int_0^t \Phi^{-1}(s) a(s) ds + \int_0^t \Phi^{-1}(s) \sigma(s) dW_s \right] + a(t) dt + \sigma(t) dW_t \\ &= (a(t) + b(t) Z_t) dt + \sigma(t) dW_t \end{aligned}$$

It is clear that $Z_0 = x$ #

(2) Take $a(t) = a$, $b(t) = -b$ and $\sigma(t) = \sigma$. Then, $\Phi(t) = e^{-bt}$ and

$$\begin{aligned} Z_t &= e^{-bt} \left[x_0 + \int_0^t e^{bs} a ds + \int_0^t e^{bs} \sigma dW_s \right] \\ &= x_0 e^{-bt} + \frac{a}{b} (1 - e^{-bt}) + \sigma \int_0^t e^{b(s-t)} dW_s \quad \# \end{aligned}$$

(3) Consider stochastic integral $\int_0^t e^{b(s-t)} dW_s$ which is a mart.

We have $E \left[\int_0^t e^{b(s-t)} dW_s \right] = 0$

$$\begin{aligned} \text{Var} \left[\int_0^t e^{b(s-t)} dW_s \right] &= E \left[\left(\int_0^t e^{b(s-t)} dW_s \right)^2 \right] \stackrel{\text{Itô's isometry}}{=} \int_0^t e^{2b(s-t)} ds \\ &= \frac{1}{2b} (1 - e^{-2bt}) \end{aligned}$$

Therefore, $E[Z_t] = x_0 e^{-bt} + \frac{a}{b} (1 - e^{-bt}) \rightarrow \frac{a}{b}$ as $t \rightarrow \infty$

$$\text{Var}[Z_t] = \text{Var} \left[\sigma \int_0^t e^{b(s-t)} dW_s \right] = \frac{\sigma^2}{2b} (1 - e^{-2bt}) \rightarrow \frac{\sigma^2}{2b} \text{ as } t \rightarrow \infty \quad \#$$

(4) To show $B_t = \sqrt{\frac{2b}{\sigma^2}} \int_0^t e^{b(s-t)} dW_s$ is a BM, we first observe that B is Gaussian

Since $Y_t = \int_0^t e^{bs} dW_s$ is Gaussian. Hence, we only need to show that B has zero mean and covariance $E[B_t B_s] = \min(t, s)$. Indeed,

$$E[B_t] = E \left[\sqrt{\frac{2b}{\sigma^2}} \int_0^{\min(t, s)} e^{bs} dW_s \right] = 0$$

For $t, s > 0$,

$$E[B_t B_s] = E[(B_t - B_s) B_s + B_s^2]$$

$$= E\left[\sqrt{2b} \int_{\frac{\ln(s+t)}{2b}}^{\frac{\ln(t+t)}{2b}} e^{bs} dw_s \cdot \sqrt{2b} \int_0^{\frac{\ln(s+t)}{2b}} e^{bs} dw_s + 2b \left(\int_0^{\frac{\ln(s+t)}{2b}} e^{bs} dw_s\right)^2\right]$$

$$= 2b \int_0^{\frac{\ln(s+t)}{2b}} e^{2bs} ds = e^{2b(\frac{\ln(s+t)}{2b})} - 1 = s$$

Similarly, for $s \geq t \geq 0$,

$$E[B_t B_s] = t$$

Finally, it is clear that $Y_t = \frac{1}{\sqrt{2b}} B_{e^{2bt}-1}$. Plugging this expression into the formula for Z_t , we get

$$Z_t = X_0 e^{-bt} + \frac{a}{b} (1 - e^{-bt}) + \frac{\sigma e^{-bt}}{\sqrt{2b}} B_{e^{2bt}-1} \quad \#$$

Ex 2 (1) Applying Ito's formula to $e^{h_t - \frac{1}{2} \langle h \rangle_t}$

$$dE(h)_t = E(h)_t [dh_t - \frac{1}{2} d\langle h \rangle_t] + \frac{1}{2} E(h)_t d\langle h \rangle_t = E(h)_t dh_t$$

It is clear that $E(h)_0 = 1$ #

(2) Since $E(h)$ is a local mart. there exists a sequence of stopping times $T_n \uparrow \infty$ s.t. for $t \geq s \geq 0$,

$$E[E(h)_{t+}^{T_n} | \mathcal{F}_s] = E(h)_s^{T_n}$$

Since $E(h)$ is non-negative, by Fatou's lemma,

$$\lim_{n \rightarrow \infty} E[E(h)_{t+}^{T_n} | \mathcal{F}_s] \geq E\left[\lim_{n \rightarrow \infty} E(h)_{t+}^{T_n} | \mathcal{F}_s\right] = E[E(h)_+ | \mathcal{F}_s]$$

On the other hand, $\lim_{n \rightarrow \infty} E(h)_s^{T_n} = E(h)_s$, from which we conclude

$$E[E(h)_+ | \mathcal{F}_s] \leq E(h)_s \quad \#$$

$$(3) E(h)_+ E(-h)_+ = e^{h_t - \frac{1}{2} \langle h \rangle_t} e^{-h_t - \frac{1}{2} \langle -h \rangle_t}$$

Note that $\langle h \rangle_t = \langle -h \rangle_t$. Indeed, $dh_t^2 = 2h_t dh_t + d\langle h \rangle_t$

$$d(-h_t)^2 = 2(-h_t) d(-h)_t + d\langle -h \rangle_t$$

Hence, $E(h)_+ E(-h)_+ = e^{-\langle h \rangle_t}$

Hence, $E(M)_t E(N)_t = e^{-\langle M \rangle_t}$

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(4) We first show that $\langle M+N \rangle_t = \frac{1}{2}(\langle M+N \rangle_t - \langle M \rangle_t - \langle N \rangle_t)$ (*)

Indeed, applying Itô's formula to $(M+N)^2$:

$$(M_t + N_t)^2 = 2 \int_0^t (M_s + N_s) d(M_s + N_s) + \langle M+N \rangle_t$$

$$\text{Hence, } \langle M+N \rangle_t = \underbrace{(M_t^2 - 2 \int_0^t M_s dM_s)}_{\langle M \rangle_t} + \underbrace{(N_t^2 - 2 \int_0^t N_s dN_s)}_{\langle N \rangle_t} + \underbrace{2(M_t N_t - \int_0^t M_s dN_s - \int_0^t N_s dM_s)}_{\langle M \cdot N \rangle_t}$$

In turn, $E(M+N)_t e^{\langle M \cdot N \rangle_t}$

$$= E(M_t + N_t - \frac{1}{2} \langle M+N \rangle_t + \langle M \cdot N \rangle_t)$$

$$= E(M_t - \frac{1}{2} \langle M \rangle_t) E(N_t - \frac{1}{2} \langle N \rangle_t) = E(M)_t E(N)_t$$

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