

• $X_t = (X_t^1, \dots, X_t^d)$ is an Ito's process

$f(t, x) = f(t, x^1, \dots, x^d)$ is $C^{1,2}$ (C^1 in t , C^2 in x)

$$df(t, X_t) = \partial_t f(t, X_t) dt + \sum_{i=1}^d \partial_i f(t, X_t) dX_t^i + \frac{1}{2} \sum_{i=1}^d \sum_{j=1}^d \partial_i \partial_j f(t, X_t) d[X^i, X^j]_t$$

If $dX_t^i = b_t^i dt + \sigma_t^i dW_t$, then $d[X^i, X^j]_t = \sigma_t^i \sigma_t^j dt$.

Ex. 9.23

$f \in C^{1,2}$, W d -dimensional BM, $X_t = f(t, W_t)$

$$dX_t = \partial_t f(t, W_t) dt + \sum_{j=1}^d \partial_j f(t, W_t) dW_t^j + \frac{1}{2} \sum_{i=1}^d \sum_{j=1}^d \partial_i \partial_j f(t, W_t) d[W^i, W^j]_t$$

$$d[W^i, W^j]_t = \begin{cases} dt, & i=j \\ 0 & i \neq j \end{cases}$$

$$dX_t = \partial_t f(t, W_t) dt + \sum_{i=1}^d \partial_i f(t, W_t) dW_t^i + \frac{1}{2} \sum_{i=1}^d \partial_i^2 f(t, W_t) dt.$$

$$= (\dots) dt + \sum_{i=1}^d (\dots) dW_t^i.$$

Ex. $M_t = (W_t + \frac{t^2}{2}) \exp(-\int_0^t s dW_s - \frac{t^3}{6})$. Is M a martingale?

Method 1:

$$f(t, x) = (x + \frac{t^2}{2}) \exp(-\int_0^t s dW_s - \frac{t^3}{6}). \quad \text{WRONG.}$$

f is not $C^{1,2}$.

$$d(tW_t) = t dW_t + W_t dt \Rightarrow tW_t = \int_0^t s dW_s + \int_0^t W_s ds$$

$$\Rightarrow \int_0^t s dW_s = tW_t - \int_0^t W_s ds.$$

$$M_t = (W_t + \frac{t^2}{2}) \exp\left(\int_0^t W_s ds - tW_t - \frac{t^3}{6}\right)$$

$$f(t, x) = (x + \frac{t^2}{2}) \exp\left(\int_0^t W_s ds - tx - \frac{t^3}{6}\right).$$

$\partial_t f$, $\partial_x f$, $\partial_x^2 f$...

Method 2:

$$X_t^1 = W_t, \quad X_t^2 = \int_0^t s dW_s, \quad (W_t = \int_0^t 1 dW_s)$$

$$d[X^1, X^1]_t = dt, \quad d[X^2, X^2]_t = t^2 dt, \quad d[X^1, X^2]_t = t dt$$

~~$$f(t, x) = \left(x + \frac{t^2}{2}\right) \exp\left(-x^2 - \frac{t^3}{6}\right)$$~~

$$f(t, x, y) = \left(x + \frac{t^2}{2}\right) \exp\left(-y - \frac{t^3}{6}\right).$$

$$\partial_t f = t \exp\left(-y - \frac{t^3}{6}\right) + \left(x + \frac{t^2}{2}\right) \exp\left(-y - \frac{t^3}{6}\right) \left(-\frac{t^2}{2}\right)$$

$$\partial_x f = 1 \cdot \exp\left(-y - \frac{t^3}{6}\right)$$

$$\partial_y f = -\left(x + \frac{t^2}{2}\right) \exp\left(-y - \frac{t^3}{6}\right)$$

$$\partial_x^2 f = 0, \quad \partial_y^2 f = \left(x + \frac{t^2}{2}\right) \exp\left(-y - \frac{t^3}{6}\right), \quad \partial_x \partial_y f = -\exp\left(-y - \frac{t^3}{6}\right).$$

$$dM_t = df(t, X_t^1, X_t^2) = \partial_t f dt + \partial_1 f dX_t^1 + \partial_2 f dX_t^2 + \frac{1}{2} \partial_1^2 f d[X^1, X^1]_t$$

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$$+ \frac{1}{2} \partial_2^2 f d[X^2, X^2]_t + \partial_1 \partial_2 f d[X^1, X^2]_t$$

$$\left(\frac{1}{2} \partial_1 \partial_2 f + \frac{1}{2} \partial_2 \partial_1 f\right)$$

$$dM_t = \exp(\dots) \left[t - \frac{t^2}{2} (X_t' + \frac{t^2}{2}) \right] dt + \exp(\dots) dW_t + \exp(\dots) \left(- (X_t' + \frac{t^2}{2}) \right) t dW_t \\ + \frac{1}{2} \exp(\dots) \left(X_t' + \frac{t^2}{2} \right)^2 dt + (-1) \exp(\dots) t dt$$

$$= (\dots) dW_t + \exp(\dots) \left[\cancel{t - \frac{t^2}{2} (W_t' + \frac{t^2}{2})} + \frac{1}{2} (W_t' + \frac{t^2}{2})^2 \cancel{t} \right] dt$$

$$= (\dots) dW_t$$

Ex. 9.24 W 2-dim BM, $X_t = \ln \left((W_t^1)^2 + (W_t^2)^2 \right)$, $t > 0$

Is X a martingale?

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

$$\partial_t f = 0, \quad \partial_x f = \frac{2x}{x^2 + y^2}, \quad \partial_y f = \frac{2y}{x^2 + y^2}$$

$$\partial_x^2 f = \frac{2(x^2 + y^2) - 2x \cdot 2x}{(x^2 + y^2)^2} = \frac{2y^2 - 2x^2}{(x^2 + y^2)^2}, \quad \partial_y^2 f = \frac{2x^2 - 2y^2}{(x^2 + y^2)^2}$$

$$dX_t = \partial_t f dt + \partial_1 f dW_t^1 + \partial_2 f dW_t^2 + \frac{1}{2} \partial_1^2 f dt + \frac{1}{2} \partial_2^2 f dt + \partial_1 \partial_2 f \cdot 0$$

$$= (\dots) dW_t^1 + (\dots) dW_t^2 + \frac{1}{2} \left[\frac{2(W_t^2)^2 - 2(W_t^1)^2}{(\quad)^2} + \frac{2(W_t^1)^2 - 2(W_t^2)^2}{(\quad)^2} \right] dt$$

!0

$$dX_t = (\dots) dW_t^1 + (\dots) dW_t^2.$$

$$\left(M_t = \int_0^t \sigma_s dW_s, \text{ if } \mathbb{E} \int_0^t \sigma_s^2 ds < \infty, \forall t, \right.$$

then M_t is a martingale.)

$$\mathbb{E} X_t = \mathbb{E} \ln \left((W_t^1)^2 + (W_t^2)^2 \right), \quad (Z^1, Z^2) \sim \mathcal{N} \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right)$$

$$= \mathbb{E} \ln \left((\sqrt{t} Z^1)^2 + (\sqrt{t} Z^2)^2 \right)$$

$$= \mathbb{E} \ln \left(t (Z^1)^2 + t (Z^2)^2 \right)$$

$$= \ln t + \mathbb{E} \ln \left((Z^1)^2 + (Z^2)^2 \right) \quad \text{is not constant.}$$

$$dX_t = z_1 f dW_t^1 + z_2 f dW_t^2$$

$$\mathbb{E} \left(\int_0^t |z_1 f|^2 ds + \int_0^t |z_2 f|^2 ds \right) = \infty.$$

• Girsanov theorem.

$Z_t = \exp\left(-\int_0^t b_s \cdot dW_s - \frac{1}{2} \int_0^t |b_s|^2 ds\right)$ is a martingale.

$$d\tilde{\mathbb{P}} = Z_T d\mathbb{P} \quad (\Omega, \mathcal{F}_T)$$

↑
For $A \in \mathcal{F}_T$, $\tilde{\mathbb{P}}(A) = \mathbb{E}(Z_T \mathbb{1}_A)$, $\int_A d\tilde{\mathbb{P}} = \int_A Z_T d\mathbb{P}$.

X is r.v. $\tilde{\mathbb{E}}X = \mathbb{E}(Z_T X)$, $\int_{\Omega} X(\omega) d\tilde{\mathbb{P}}(\omega) = \int_{\Omega} Z_T(\omega) X(\omega) d\mathbb{P}(\omega)$.

Ex. Compute $\mathbb{E}\left(e^{W_T} \int_0^T W_s^2 ds\right)$.

~~$b_t = -1$~~ $b_t = -1$

$$Z_t = \exp\left(\int_0^t 1 dW_s - \frac{1}{2} \int_0^t 1 ds\right)$$

$= \exp(W_t - \frac{1}{2}t)$ is a martingale.

$$Z_T = \exp(W_T - \frac{1}{2}T).$$

$d\tilde{\mathbb{P}} = Z_T d\mathbb{P}$. $\tilde{W}_t = W_t - t, t \in [0, T]$ is a BM under $\tilde{\mathbb{P}}$.

Then, $\tilde{W}_t = W_t + \int_0^t b_s ds, t \in [0, T]$
is a BM under $\tilde{\mathbb{P}}$.

$$\begin{aligned}
\mathbb{E}\left(e^{W_T} \int_0^T W_s^2 ds\right) &= \mathbb{E}\left(\underbrace{e^{W_T - \frac{1}{2}T}}_{=1} \int_0^T W_s^2 ds \cdot e^{\frac{1}{2}T}\right) \\
&= e^{\frac{1}{2}T} \mathbb{E}\left(Z_T \int_0^T W_s^2 ds\right) = e^{\frac{1}{2}T} \tilde{\mathbb{E}} \int_0^T W_s^2 ds = e^{\frac{1}{2}T} \tilde{\mathbb{E}} \int_0^T (\tilde{W}_s + s)^2 ds \\
&= e^{\frac{1}{2}T} \int_0^T \tilde{\mathbb{E}}(\tilde{W}_s^2 + 2s\tilde{W}_s + s^2) ds = e^{\frac{1}{2}T} \int_0^T (s + 0 + s^2) ds = e^{\frac{1}{2}T} \left(\frac{T^2}{2} + \frac{T^3}{3}\right).
\end{aligned}$$

Ex. Under \mathbb{P} , ~~$dS_t =$~~ α, σ, r are constant.

$$dS_t = r S_t dt + \sigma S_t d\tilde{W}_t.$$

pf. $dS_t = \alpha S_t dt + \sigma S_t dW_t$ ($\tilde{W}_t = W_t + \theta t$)

$$= \alpha S_t dt + \sigma S_t d(\tilde{W}_t - \theta t)$$

$$= \alpha S_t dt + \sigma S_t d\tilde{W}_t - \theta \sigma S_t dt$$

$$= (\alpha - \theta \sigma) S_t dt + \sigma S_t d\tilde{W}_t$$

$$= \left(\alpha - \frac{\alpha - r}{\sigma} \cdot \sigma\right) S_t dt + \sigma S_t d\tilde{W}_t$$

$$= r S_t dt + \sigma S_t d\tilde{W}_t.$$

- The (unique) risk neutral measure is given by

$$d\tilde{\mathbb{P}} = \exp\left(-\int_0^T \theta_s dW_s - \frac{1}{2} \int_0^T \theta_s^2 ds\right) d\mathbb{P}, \quad \theta_t = \frac{\alpha_t - R_t}{\sigma_t}$$

$$dS_t = \alpha_t S_t dt + \sigma_t S_t dW_t, \quad D_t = e^{-\int_t^T R_s ds}.$$

($\tilde{W}_t = W_t + \int_0^t \theta_s ds$, $t \in [0, T]$ is a BM under $\tilde{\mathbb{P}}$).

- Theorem 10.5

A security pays V_T at maturity T .

AFP at time t is $V_t = \frac{1}{D_t} \tilde{\mathbb{E}}_t(D_T V_T)$

($D_t V_t$ is a martingale under $\tilde{\mathbb{P}}$).

- Theorem 11.1

A security pays $V_T = g(S_T)$ at T , AFP at t $f(t, S_t)$,

$$f(t, x) = \int_{-\infty}^{+\infty} e^{-r\tau} g(x \exp((r - \frac{\sigma^2}{2})\tau + \sigma\sqrt{\tau}y)) \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy, \quad \tau = T - t.$$

$$dS_t = rS_t dt + \sigma S_t d\tilde{W}_t \Rightarrow S_t = S_0 \exp\left(\left(r - \frac{\sigma^2}{2}\right)t + \sigma \tilde{W}_t\right).$$

$$S_T = S_0 \exp\left(\left(r - \frac{\sigma^2}{2}\right)T + \sigma \tilde{W}_T\right)$$

$$\Rightarrow S_T = S_t \exp\left(\left(r - \frac{\sigma^2}{2}\right)\tau + \sigma(\tilde{W}_T - \tilde{W}_t)\right).$$

$$\begin{aligned} V_t &= e^{-rt} \mathbb{E}_t \left(e^{-rT} K_T \right) = e^{-rT} \mathbb{E}_t \left(g(S_T) \right) \\ &= e^{-rT} \mathbb{E}_t \left(g\left(S_t \exp\left(\left(r - \frac{\sigma^2}{2}\right)\tau + \sigma(\tilde{W}_T - \tilde{W}_t)\right) \right) \right) \\ &= f(t, S_t) \end{aligned}$$

$$\begin{aligned} f(t, x) &= e^{-rT} \mathbb{E} \left(g\left(x \exp\left(\left(r - \frac{\sigma^2}{2}\right)\tau + \sigma(\tilde{W}_T - \tilde{W}_t)\right) \right) \right), \quad \tilde{W}_T - \tilde{W}_t \sim N(0, \tau). \\ &= e^{-rT} \int_{-\infty}^{+\infty} g\left(x \exp\left(\left(r - \frac{\sigma^2}{2}\right)\tau + \sigma\sqrt{\tau} y\right) \right) \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy. \end{aligned}$$

Ex. European call with strike K . $g(x) = (x-K)^+ = \max(x-K, 0)$.

$$c(t, x) = e^{-r\tau} \int_{-\infty}^{+\infty} \left(x \exp\left(\left(r - \frac{\sigma^2}{2}\right)\tau + \sigma\sqrt{\tau}y\right) \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} - K \right)^+ \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy.$$

$$x \exp\left(\left(r - \frac{\sigma^2}{2}\right)\tau + \sigma\sqrt{\tau}y\right) - K > 0$$

$$x^+ = \begin{cases} x, & x > 0 \\ 0, & x \leq 0 \end{cases}$$

$$\exp\left(\left(r - \frac{\sigma^2}{2}\right)\tau + \sigma\sqrt{\tau}y\right) > \frac{K}{x}$$

$$\left(r - \frac{\sigma^2}{2}\right)\tau + \sigma\sqrt{\tau}y > \ln\left(\frac{K}{x}\right) = -\ln\left(\frac{x}{K}\right)$$

$$y > \frac{-\ln\left(\frac{x}{K}\right) - \left(r - \frac{\sigma^2}{2}\right)\tau}{\sigma\sqrt{\tau}} = -\frac{\ln\left(\frac{x}{K}\right) + \left(r - \frac{\sigma^2}{2}\right)\tau}{\sigma\sqrt{\tau}} = -d_-(\tau, x).$$

$$\begin{aligned} \Rightarrow c(t, x) &= e^{-r\tau} \int_{-d_-}^{+\infty} \left(x \exp\left(\left(r - \frac{\sigma^2}{2}\right)\tau + \sigma\sqrt{\tau}y\right) - K \right) \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy \\ &= e^{-r\tau} x \int_{-d_-}^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left(\left(r - \frac{\sigma^2}{2}\right)\tau + \sigma\sqrt{\tau}y - \frac{y^2}{2}\right) dy - e^{-r\tau} \int_{-d_-}^{+\infty} K \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy \\ &= x \int_{-d_-}^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}(\sigma^2\tau - 2\sigma\sqrt{\tau}y + y^2)\right) dy - e^{-r\tau} K (1 - N(-d_-)) \\ &= x \int_{-d_-}^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}(y - \sigma\sqrt{\tau})^2\right) dy - e^{-r\tau} K (1 - N(-d_-)). \end{aligned}$$

$$z = y - \sigma\sqrt{\tau}$$

$$c(t, x) = x \int_{-d_- - \sigma\sqrt{\tau}}^{+\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{z^2}{2}\right) dz - e^{-r\tau} K (1 - N(-d_-)).$$

$$= x (1 - N(-d_- - \sigma\sqrt{\tau})) - e^{-r\tau} K (1 - N(-d_-)).$$

$$N(-x) = \mathbb{P}(N(0,1) \leq -x) = \mathbb{P}(-N(0,1) \geq x) = \mathbb{P}(N(0,1) \geq x) = 1 - \mathbb{P}(N(0,1) \leq x) \\ = 1 - N(x).$$

$$c(t, x) = x N(d_- + \sigma\sqrt{\tau}) - e^{-r\tau} K N(d_-)$$

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

$$\Rightarrow c(t, x) = x N(d_+) - e^{-r\tau} K N(d_-).$$