

Last time: $X = (X_1, \dots, X_d)$

$$f = f(t, x_1, \dots, x_d) \quad \Phi = \Phi(t, x) \quad (\text{with } x = (x_1, x_2, \dots, x_d))$$

$$\text{Itô: } d f(t, X) = \frac{\partial f}{\partial t} dt + \sum_{i=1}^d \frac{\partial f}{\partial x_i} dX_i + \frac{1}{2} \sum_{i,j=1}^d \frac{\partial^2 f}{\partial x_i \partial x_j} d[X_i, X_j](t)$$

Multi dim BM: $W = (W_1, \dots, W_d)$ is a d -dim std B.M. if

① Each W_i is a std 1D B.M.

& ② for $i \neq j$ W_i is ind of W_j

$$\text{Compute: } d[W_i, W_j](t) = \begin{cases} dt & \text{if } i=j \\ 0 dt & \text{if } i \neq j \end{cases}$$

Recitation tomorrow
(Q7.5).

Levy: M is a d -dim cts mg. If $d[M_i, M_j] = \begin{cases} dt & i=j \\ 0 dt & i \neq j \end{cases} \Rightarrow M$ is a B.M.

$$M = (M_1, M_2, \dots, M_d)$$

Eg 1°. Say $f = f(t, x_1, \dots, x_d)$. & $W = d$ dim B.M.

$$\text{Compute } d(f(t, W(t))) \stackrel{\text{Ito}}{=} \partial_t f(t, W(t)) dt + \sum_{i=1}^d \partial_i f(t, W(t)) dW_i(t) + \frac{1}{2} \sum_{i,j=1}^d \partial_i \partial_j f(t, W(t)) d[W_i, W_j](t).$$

$$= \partial_t f dt + \sum \partial_i f dW_i(t) + \frac{1}{2} \sum_{i=1}^d \partial_i^2 f(t, W(t)) dt$$

Note $\sum_{i=1}^d \partial_i^2 f = \partial_1^2 f + \partial_2^2 f + \dots + \partial_d^2 f = \text{laplacian of } f = \Delta f$

$$\Rightarrow d f(t, W(t)) = \left(\partial_t f + \frac{1}{2} \Delta f \right) dt + \sum \partial_i f dW_i(t).$$

Eg 2: Choose $d=2$. $W = (W_1, W_2)$ (2D B.M.).

$$f(t, x) = f(t, x_1, x_2) = \ln(|x|) = \ln \sqrt{x_1^2 + x_2^2} \\ = \frac{1}{2} \ln(x_1^2 + x_2^2).$$

$$\frac{\partial f}{\partial t} = 0 \quad \frac{\partial f}{\partial x_1} = \frac{1}{2} \frac{1}{|x|^2} 2x_1 = \frac{x_1}{|x|^2}$$

$$\frac{\partial f}{\partial x_2} = \frac{x_2}{|x|^2}$$

~~$\frac{\partial^2 f}{\partial x_i \partial x_j}$~~ $\frac{\partial^2 f}{\partial x_1^2} + \frac{\partial^2 f}{\partial x_2^2} = \Delta f = \text{You check} = 0$

$$\Rightarrow d \ln |W(t)| = \sum_{i=1}^2 \frac{W_i(t)}{|W(t)|^2} dW_i + \frac{1}{2} \frac{\Delta f}{0} dt$$

$$\Rightarrow d(\ln |W(t)|) = \underbrace{\frac{W_1(t) dW_1(t)}{|W(t)|^2}}_{mg} + \underbrace{\frac{W_2(t) dW_2(t)}{|W(t)|^2}}_{mg}.$$

$$E \ln |W(t)| = \int_{\mathbb{R}^2} \ln |x| e^{-|x|^2/2t} \frac{dx_1 dx_2}{2\pi t} \xrightarrow{t \rightarrow \infty} +\infty$$

Claim: RHS is a "local martingale" (& not a mg).

For $\int_0^t \sigma(s) dW(s)$ to be a mg need σ to be adapted. & $E \int_0^t \sigma(s)^2 ds < \infty$.

Risk Neutral measures:

Motivation: Say financial market. Interest rate r .

$X(t)$ = wealth of the R. Pf of a security
with payoff $V(T)$ at time T .

Suppose

$e^{-rt} X(t)$ is a mg.

Discounted wealth.

Then can compute $X(t)$ in terms of $V(T)$!!

Note $E(V(T) | \mathcal{F}_t) = E(X(T) | \mathcal{F}_t) = e^{+rT} E(e^{-rT} X(T) | \mathcal{F}_t)$

$$= e^{+rT} e^{-rt} X(t)$$

$$\Rightarrow X(t) = e^{-r(T-t)} E(V(T) | \mathcal{F}_t)$$

If Disc wealth is a mg then price securities by

Usually disc wealth is not a mg.

Risk Neutral Measure: Construct a new measure under which.
the discounted wealth process is a mg

① Equivalent measures: $P \rightarrow$ prob measure.

\tilde{P} a new prob measure (Financial app: Risk Neutral measure).

P & \tilde{P} are equivalent if ~~whenever~~ $P(A) = 0 \iff \tilde{P}(A) = 0$.

Eg: Z be a R.V. Assume $Z > 0$ almost surely.
& $E Z = 1$.

Define a new measure \tilde{P} by

$$\tilde{P}(A) = \int_A Z dP = E(\mathbb{1}_A Z) \in [0, 1].$$

Note $P(A) = 0 \iff \tilde{P}(A) = 0$ (i.e. \tilde{P} & P are equiv.).

Notation: If $\tilde{P}(A) = \int_A z dP$ write $d\tilde{P} = z dP$.

Reason: Can check $\int_{\Omega} X d\tilde{P} = \int_{\Omega} X z dP$

$\tilde{E} \rightarrow$ Expected value wrt the new measure \tilde{P} .

$\tilde{E}(X | \mathcal{F}) \rightarrow$ cond exp of X given the σ alg \mathcal{F}
under the new measure \tilde{P} .

Then (Cameron-Martin-Girsanov theorem).

$$b(t) = (b_1(t), b_2(t), \dots, b_d(t)) \leftarrow d \text{ dim adapted process.}$$

$$W(t) = (W_1(t), \dots, W_d(t)) \leftarrow d \text{ dim B.M.}$$

$$\text{Let } \tilde{W}(t) = W(t) + \int_0^t b(s) ds.$$

$$\begin{aligned} \text{Let } Z(T) &= \exp\left(-\int_0^T b(s) \cdot dW(s) - \frac{1}{2} \int_0^T |b(s)|^2 ds\right) \\ &= \exp\left(-\int_0^T \sum_{i=1}^d b_i(s) dW_i(s) - \frac{1}{2} \int_0^T \sum_{i=1}^d b_i(s)^2 ds\right). \end{aligned}$$

If Z is a mg, then define $d\tilde{P} = Z(T) dP$.

Then \tilde{W} is a B.M. under \tilde{P} (up to time T).

Note: Compute dZ .

$$\text{let } M(t) = - \int_0^t b(s) \cdot dW(s) = - \int_0^t \sum b_i(s) dW_i(s)$$

$$\Rightarrow [M, M](t) = \sum_{i=1}^d \int_0^t b_i(s)^2 ds = \int_0^t |b(s)|^2 ds.$$

$$\Rightarrow Z(t) = \exp\left(M(t) - \frac{1}{2} [M, M](t)\right).$$

Itô: let $f(t, x) = \exp\left(x - \frac{1}{2} [M, M](t)\right)$.

$$\partial_t f = -\frac{1}{2} f \partial_t [M, M]. \quad \partial_x f = f \quad \& \quad \partial_x^2 f = -f.$$

$$dZ = \partial_t f dt + \partial_x f dM + \frac{1}{2} \partial_x^2 f d[M, M].$$

$$= -\frac{1}{2} z \cancel{d[M, M]} + z dM + \frac{1}{2} z \cancel{d[M, M]}$$

$$= z dM = z(t) (-b(t) \cdot dW(t)).$$

$$dz(t) = - \sum_{i=1}^d z b_i(t) dW_i(t) \quad \leftarrow \text{looks like a mg!}$$

(could be a LOCAL mg).

Need to assume Z is a mg in the thm.

Note: for $\tilde{\mathbb{P}}$ to be a prob measure we need

$$E Z(T) = \underline{1}.$$

Since Z is a mg, $E Z(T) = E Z(0) = \underline{1}.$

Pf of Girsanov:

lemma let $0 \leq s \leq t$. $X \rightarrow \mathcal{F}_t$ meas R.V.

$$\text{then } \underset{\uparrow}{\mathbb{E}}(X | \mathcal{F}_s) = \frac{1}{z(s)} \underset{\uparrow}{\mathbb{E}}(z(t)X | \mathcal{F}_s). \dots (*)$$

Cond exp
wrt \rightarrow new meas \tilde{P}

Cond exp wrt P .

Pf: Recall: If Y is any R.V.

$$\text{then } \int_A \mathbb{E}(Y | \mathcal{F}_s) d\tilde{P} = \int_A Y d\tilde{P} \quad \text{for every } A \in \mathcal{F}_s.$$

Check \otimes : Let $A \in \mathcal{F}_s$.

$$\int_A \tilde{E}(X | \mathcal{F}_s) d\tilde{P} \stackrel{\textcircled{1}}{=} \int_A z(T) \tilde{E}(X | \mathcal{F}_s) d\tilde{P}$$

$$= \int_A E(z(T) \tilde{E}(X | \mathcal{F}_s) | \mathcal{F}_s) dP \quad (\because A \in \mathcal{F}_s).$$

$$= \int_A \tilde{E}(X | \mathcal{F}_s) \underbrace{E(z(T) | \mathcal{F}_s)}_{z(s)} dP$$

$$= \int_A \tilde{E}(X | \mathcal{F}_s) z(s) dP \quad \dots \quad \otimes$$

②

$$\int_A \tilde{E}(X | \mathcal{F}_s) d\tilde{P} \stackrel{\textcircled{2}}{=} \int_A X d\tilde{P} = \int_A z(t) X dP$$

$$= \int_A E(z(t)X | \mathcal{F}_s) dP$$

$$= \int_A E\left(\underbrace{E(z(t)X | \mathcal{F}_t)}_{\downarrow} | \mathcal{F}_s\right) dP$$

$$= \int_A E(z(t)X | \mathcal{F}_s) dP. \dots \textcircled{***}$$

$$\Rightarrow z(s) \tilde{E}(X | \mathcal{F}_s) = E(z(t)X | \mathcal{F}_s).$$

//

Lemma 2: An adapted process M is a mg under \tilde{P}

$\Leftrightarrow ZM$ is a mg under P .

Pf: Say MZ is a mg under P .

NTS M is a mg under \tilde{P} .

$$\text{Compute } \tilde{E}(M(t) | \mathcal{F}_s) = \frac{1}{Z(s)} E(Z(t)M(t) | \mathcal{F}_s)$$

$$= \frac{1}{Z(s)} Z(s) M(s) = M(s) //$$

Pf of Girsanov. $d\tilde{W} = b(t)dt + dW(t)$.

NTS \tilde{W} is a ~~new~~ B.M. under \tilde{P} .

Pf: Levy. ① Check \tilde{W} is a ~~cts~~ mg under \tilde{P}

② Check $d[\tilde{W}_i, \tilde{W}_j] = \begin{cases} 0 dt & i \neq j \\ dt & i = j \end{cases}$.

Note ② is immediately true! ($d[\tilde{W}_i, \tilde{W}_j] = d[W_i, W_j]$).

For ①. Only need to check $Z\tilde{W}$ is a mg under P .

$$d(Z\tilde{W}_j) = Z d\tilde{W}_j + \tilde{W}_j dZ + d[Z, \tilde{W}_j].$$

$$= \underbrace{\tilde{W}_j}_{mg} dZ + Z \cancel{b_j dt} + Z \underbrace{dW_j}_{mg} - Z \cancel{b_j dt}$$

OED.