

Last time:

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

$X_1, X_2, \dots, X_d$

Multi D B.M:  $W = (W_1, W_2, \dots, W_d)$

① Each  $W_i$  is a ~~1D~~ standard 1D BM.

② For  $i \neq j$ ,  $W_i$  is ind of  $W_j$ .

Filtration:

$$\mathcal{F}_t^W = \sigma\left(\bigcup_{i=1}^d \bigcup_{s \leq t} W_i(s)\right)$$

Note:  $W_i$  &  $W_j$  are ind for  $i \neq j$ ,  $[W_i, W_j] = 0$

(alternate notation " $dW_i dW_j = 0$ ").

↳ for  $i = j$   $d[W_i, W_i](t) = dt$ .

$$\Rightarrow d[W_i, W_j](t) = \mathbb{1}_{\{i=j\}} dt$$

Thm (Levy): If  $M$  is a cts mg such that

$$d[M_i, M_j] = \mathbb{1}_{\{i=j\}} dt$$

then  $M$  is a standard  $d$ -dim BM.

Eg from last time:  $W \rightarrow 2D$  BM.

$$f(x) = f(x_1, x_2) = \ln |x|^2 = \ln(x_1^2 + x_2^2).$$

$$\partial_1 f = \frac{1}{|x|^2} \cdot 2x_1, \quad \partial_2 f = \frac{2x_2}{|x|^2}$$

↳ You compute  $\Delta f = \partial_1^2 f + \partial_2^2 f = 0$

Set  $Y(t) = f(W(t)) = f(W_1(t), W_2(t)) = \ln(W_1(t)^2 + W_2(t)^2)$ .

Compute  $dY = \frac{2W_1(t)}{|W(t)|^2} dW_1(t) + \frac{2W_2(t)}{|W(t)|^2} dW_2(t) + \frac{1}{2} \cdot 0 \cdot dt$ .

Guess  $Y$  is a mg.

But  $Y$  can not be a mg.

Knows, if  $Y$  is a mg, then  $EY(t) = EY(0)$   $\leftarrow$  constant in time.

$$\text{Compute } EY(t) = E \ln(|W(t)|^2)$$


$$= \int_{\mathbb{R}^2} \ln(|x|^2) \frac{1}{2\pi t} e^{-|x|^2/2t} dx_1 dx_2.$$

$$y = \frac{x}{\sqrt{t}}$$
$$dx_1 dx_2 = dy \cdot t$$

$$= \int_{\mathbb{R}^2} \ln(t|y|^2) e^{-|y|^2/2} \frac{dy_1 dy_2}{2\pi}.$$

$$= \underbrace{\ln t \int_{\mathbb{R}^2} ( ) dy_1 dy_2}_{=1} + \int_{\mathbb{R}^2} \ln(|y|^2) e^{-|y|^2/2} dy_1 dy_2.$$

$$= \ln t + \int_{\mathbb{R}^2} ( ) dy_1 dy_2.$$


  
 does not dep on t.

= not constant in time.

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Reason:  $\int_0^t \sigma(s) dW$  is only guaranteed to be a mg  
 if  $E \int_0^t \sigma(s)^2 ds < \infty$ .

## Risk Neutral Measures:

Security  $\rightarrow$  payoff  $V(T)$  ( $\mathcal{F}_T$  measurable).

RNPF says ~~price~~ AFP of this security is

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

$\tilde{\mathbb{E}}$  = cond exp wrt the RN M.  
(IOU, today).

Def: Two  $\sigma$  measures  $P$  &  $\tilde{P}$  are said to be equivalent if ~~whenever~~  $P(A) = 0 \Leftrightarrow \tilde{P}(A) = 0$ .

Eg: Let  $Z$  be a rv, such that ①  $E Z = 1$  & ②  $Z > 0$

$$\text{Define } \tilde{P}(A) = \int_A Z dP$$

Then ~~the~~ ~~&~~ ~~the~~  $\tilde{P}$  is a  $\sigma$  measure  
&  $P$  &  $\tilde{P}$  are equiv.

Thm: If  $P$  &  $\tilde{P}$  are equiv, then ~~if~~ there exists a RV  $Z$  such that  $\tilde{P}(A) = \int_A Z dP$  for all measurable sets  $A$ .  
(Radon Nicodým Thm).

Notation: If  $\tilde{P}$  is defined by  $\tilde{P}(A) = \int_A z \, dP$

for all  $A$ , then

① We say  $d\tilde{P} = z \, dP$

& ②  $z$  is called the Radon-Nikodym derivative of  $\tilde{P}$  w.r.t  $P$ .  
( $z = \frac{d\tilde{P}}{dP}$ ).

Remark: If  $X$  is a R.V., then denote  $\tilde{E}X = \text{Exp of } X \text{ w.r.t } \tilde{P}$

$$\begin{aligned} \text{If } d\tilde{P} = z \, dP, \text{ then } \tilde{E}X &= \cancel{z} \, EX && E(Xz) \\ &= \int X \, d\tilde{P} = \int X z \, dP && \text{''} \end{aligned}$$



Thm (Cameron Martin Girsanov). Let  $b = (b_1, b_2, \dots, b_d)$  some  
d-dimensional adapted process.

$W \rightarrow$  d-dim B.M.

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

Let  $Z(t) = \exp\left(-\int_0^t b(s) \cdot dW(s) - \frac{1}{2} \int_0^t |b(s)|^2 ds\right)$ .

Fix  $T > 0$ . & define a new measure  $\tilde{\mathbb{P}} = \tilde{\mathbb{P}}_T$  by

$$d\tilde{\mathbb{P}} = Z(T) d\mathbb{P}.$$

If  $Z$  is a mg, then  $\tilde{W}$  is a BM wrt  $\tilde{\mathbb{P}}$  (up to time  $T$ ).  
(under  $\tilde{\mathbb{P}}$ ).

Rank:  $b \cdot dW = \sum_{i=1}^d b_i dW_i$

Rank: Compute  $dZ$ . Let  $M(t) = \int_0^t b(s) \cdot dW(s) = \sum_1^d \int_0^t b_i(s) dW_i(s)$

Note  $[M, M]_t = \sum_1^d \int_0^t b_i^2(s) ds = \int_0^t |b(s)|^2 ds$ .

Let  $f(t, x) = e^{-x - \frac{1}{2} \int_0^t |b(s)|^2 ds} \Rightarrow Z(t) = f(t, M(t))$ .

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

$= -\frac{1}{2} \exp(\dots) |b(t)|^2 dt - Z dM + \frac{1}{2} Z d[M, M]$ .

$= -\frac{1}{2} Z |b|^2 dt - Z dM + \frac{1}{2} Z |b(t)|^2 dt = \cancel{Z dM}$

$$= -Z \sigma \cdot dW$$

$\Rightarrow Z$  is a mg (~~is~~ PROVIDED  $E \int_0^t Z^2 |\sigma|^2 ds < \infty$ )

Dont know.

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Need  $Z$  to be a mg because:

we need  $\tilde{P}$  to be a  $\sigma$  measure.

Note  $\tilde{P}(\Omega) = E Z(T)$

If  $Z$  is a mg, then  $E Z(T) = E Z(0) = 1$ .

Strategy of Proof: (1) Check  $\tilde{W}$  is a mg under  $\tilde{P}$ . } use Lem 1'  
 (2) Check  $d[\tilde{W}_i, \tilde{W}_j] = \mathbb{1}_{i=j} dt$ .

Lemma: Let  $X$  be a  $\mathcal{F}_t$  meas RV.  $0 \leq s \leq t$ .

$$\text{then } \tilde{E}(X | \mathcal{F}_s) = \frac{1}{z(s)} E(z(t)X | \mathcal{F}_s).$$

cond exp  
 wrt  $\tilde{P}$

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

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

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

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

$$= \int_A \tilde{E}(X | \mathcal{F}_s) z(s) dP. \quad \dots \textcircled{1}$$

Also,  $\int_A \tilde{E}(X | \mathcal{F}_s) d\tilde{P} = \int_A X d\tilde{P} = \int_A X z(T) dP$

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

$$= \int_A E(X z(t) | \mathcal{F}_s) dP \quad \dots \textcircled{2}$$

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

$\iff MZ$  is a mg under  $P$ .

Pf: ① Say  $MZ$  is a mg under  $P$ .

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

$\stackrel{MZ \text{ a mg}}{=} \frac{1}{Z(s)} M(s) Z(s) = M(s)$

QED.

② Conversely, say  $M$  is a mg under  $\tilde{P}$ .

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

QED.

$\Rightarrow$  for every  $A \in \mathcal{F}_s$ ,

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

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

$$\Rightarrow \tilde{E}(X | \mathcal{F}_s) = \frac{1}{z(s)} E(X z(t) | \mathcal{F}_s).$$

Proof of Girsanov:

$$\text{Claim ①: } d[\tilde{W}_i, \tilde{W}_j] = \mathbb{1}_{\{i=j\}} dt$$

$$\text{Pf: } d[\tilde{W}_i, \tilde{W}_j] = d[W_i, W_j] = \mathbb{1}_{\{i=j\}} dt. \quad \checkmark$$

Claim ②:  $\tilde{W}_i$  is a mg under  $\tilde{P}$ .

Pf: By Lemma 2, only NTS  $\tilde{W}_i Z$  is a mg under  $\tilde{P}$ .

Check: Compute  $d(\tilde{W}_i Z) = \tilde{W}_i dZ + Z d\tilde{W}_i + d[\tilde{W}_i, Z]$ .

$$= -Z \tilde{W}_i b_i dW + Z (b_i dt + dW) - Z b_i dt$$

dt terms cancel!  $\Rightarrow \tilde{W}_i Z$  is a mg under  $\tilde{P}$ .



By Levy  $\Rightarrow \tilde{W}$  is a BM under  $\tilde{P}$ .

QED.

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