

## LECTURE 6

Let  $T \in D'(\mathbb{R}^d)$  (or  $\mathcal{S}'(\mathbb{R}^d)$ ).

Then  $\forall \varphi \in C_c^\infty(\mathbb{R}^d)$  we can

define canonically  $T \cdot \varphi \in D'(\mathbb{R}^d)$ :

$$T \cdot \varphi (\psi) := T(\varphi \psi), \quad \forall \psi \in C_c^\infty(\mathbb{R}^d).$$

Let now  $T \in \mathcal{D}'(\mathbb{R}^d)$  and

$\varphi \in C_c(\mathbb{R}^d) \setminus C_c^\infty(\mathbb{R}^d)$ . Then

what is  $T \cdot \varphi (\psi) = ?$  since

in general  $\varphi \psi \notin C_c^\infty(\mathbb{R}^d)$ , we can

NOT compute  $T(\varphi \psi)$ .

EVEN WORSE: if  $T_1, T_2 \in \mathcal{D}'(\mathbb{R}^d)$ ,

there is NO CANONICAL WAY of defining  $T_1 T_2 (\varphi)$ , for  $\varphi \in C_c^\infty(\mathbb{R}^d)$ .

In particular:  $T^2, T^3, T^4$  -

are in general ill-defined.

THIS IS WHAT WE MEAN when we

say that  $\int_{\mathbb{R}^d} V(x(\delta_n)) dx$  is

ill-defined for the GFF<sub>an</sub>

$x \mapsto x(\delta_n)$  is NOT a function

if  $d > 2$  and for generic

(even polynomial)  $V: \mathbb{R} \rightarrow \mathbb{R}$ ,

$x \mapsto V(x(\delta_n))$  is ill-defined.

NOTE that "ill-defined"  
 does NOT mean "impossible to  
 define" but rather  
 "impossible to define canonically".  
 We shall see that  $\gamma^2$  and  
 $\gamma^3$  can and must be defined  
 for a specific distribution, but  
 this will be a choice.

## EXAMPLE

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If  $B$  is a two-sided Brownian motion,  
 $B'_t = \frac{dB_t}{dt} \in \mathcal{D}'(\mathbb{R})$

and  $B \in C(R)$ .

What is the product

$$B \cdot B' ?$$

Recall Itô's formula:

$$B'_t = 2 \int_0^t B_s dB_s + t, \quad t \geq 0.$$

In Stratonovich form:

$$B'_t = 2 \int_0^t B_s^\circ dB_s$$

(Recall:  $\forall t \geq 0$

$$\int_0^t B_s dB_s = \lim_{n \rightarrow \infty} \sum_{\frac{i}{n} \leq s < \frac{i+1}{n}} B_{\frac{i}{n}} \left( B_{\frac{i+1}{n}} - B_{\frac{i}{n}} \right)$$

$$\int_0^t B_s \circ dB_s =$$

$$= \lim_{m \rightarrow 0} \sum_{\frac{i+1}{m} \leq t} \frac{B_{\frac{i}{m}} + B_{\frac{i+1}{m}}}{2} \left( B_{\frac{i+1}{m}} - B_{\frac{i}{m}} \right)$$

(limits in  $L^2(\Omega)$ ) .

Then we have two possible

definitions :

$$B_t \circ B_t^I := \frac{d}{dt} \frac{B_t^2 - t}{2} \quad (\text{It\^o})$$

$$B_t \circ B_t^I := \frac{d}{dt} \frac{B_t^2}{2} \quad (\text{Stratonovich})$$

and there many more !

Neither of these are canonical.  
One has to choose, and then  
work with this choice.

EXAMPLE : in Lecture 4

we saw that

$$\dot{z} = (m - \Delta_\varepsilon) z - \lambda (z^3 + 3zy^2 + 3z^2y + y^3)$$

We're going to see that as  $\varepsilon \downarrow 0$ ,

$$Y = Y^{(\varepsilon)} \rightarrow \bar{Y}$$

where  $\bar{Y}$  is a distribution.

So, what about  $y^2$  and  $y^3$ ?

## Reversibility of gradient-type SDEs

Let  $U: \mathbb{R}^d \rightarrow \mathbb{R}$  of class  $C^2$  s.t.  $U, DU$

are Lipschitz:  $|U(x) - U(y)| \leq L|x-y|$

$$|DU(x) - DU(y)| \leq L|x-y|$$

Let  $B = (B^1, \dots, B^d)$  be a

$\mathbb{R}^d$ -valued Brownian Motion (namely

$d$  independent BMs).

We study the SDEs

$$X_t(x) = x - \int_0^t DU(X_s)ds + B_t, \quad t \geq 0.$$

NOTE:  
Has used  
opposite sign

Note: as in Lecture 4, if

$$Z_t := X_t - B_t, \quad \text{then}$$

$$Z_t = x - \int_0^t DU(Z_s + B_s)ds,$$

a Random ODE rather than a SDE.

We want to prove that

$$\int_{\mathbb{R}^d} f(x) \mathbb{E} \left( g(X_t(x)) \right) e^{-2U(x)} dx = \int_{\mathbb{R}^d} g(x) \mathbb{E} \left( f(X_t(x)) \right) e^{-2U(x)} dx$$

for all  $f, g : \mathbb{R}^d \rightarrow \mathbb{R}$  measurable bounded.

We define

$$L_t := \int_0^t \langle \nabla U(X_s), dB_s \rangle$$

$$= \int_0^t \sum_{i=1}^d \frac{\partial U}{\partial x_i}(X_s) dB_s^i$$

and  $M_t = \mathbb{E}(L)_t$ , the associated

exponential (local) martingale, i.e.

$$M_t = \exp \left( \int_0^t \langle \nabla V(x_s), dB_s \rangle - \frac{1}{2} \int_0^t \|\nabla V(x_s)\|^2 ds \right)$$

Since  $\|\nabla V\|$  is bounded ( $V$  is Lipschitz), by Novikov's condition

$(M_t)_{t \geq 0}$  is a martingale,

$$E(M_t) = 1 \quad \text{and} \quad M_t dP =: Q_t$$

is a new probability measure on  $\Omega$ . By Girsanov's theorem

under  $Q_t$

$$X_s(u) - u = B_s - \langle L, B \rangle_s, \quad s \in [0, t]$$

is a BM in  $\mathbb{R}^d$ .

By Itô's formula

$$\begin{aligned} U(X_t^{(x)}) &= U(x) + \int_0^t \langle \nabla U(X_s), dX_s \rangle \\ &\quad + \frac{1}{2} \int_0^t \text{Tr} [\nabla^2 U(X_s)] ds \\ &= U(x) + \int_0^t \left( -\|\nabla U(X_s)\|^2 + \frac{1}{2} \text{Tr} [\nabla^2 U(X_s)] \right) ds \\ &\quad + \int_0^t \langle \nabla U(X_s), dB_s \rangle. \end{aligned}$$

Then :

$$\begin{aligned} \int_0^t \langle \nabla U(X_s), dB_s \rangle &= \\ &= U(X_t) - U(x) + \int_0^t \left( \|\nabla U(X_s)\|^2 - \frac{1}{2} \text{Tr} [\nabla^2 U(X_s)] \right) ds \end{aligned}$$

and

$$M_t = \exp \left( U(X_t) - U(x) + \int_0^t \nu(X_s) ds \right)$$

$$\text{with } \nu(y) := \frac{1}{2} \left( \|\nabla U(y)\|^2 - \text{Tr} [\nabla^2 U(y)] \right)$$

NOW :

$$\begin{aligned}
 & \int f(x) \mathbb{E} [g(X_{t(x)})] e^{-2V(x)} dx = \\
 &= \int f(x) \mathbb{E}_{Q_t} \left[ g(X_{t(x)}) e^{-2V(x)} \frac{1}{M_t} \right] dx \\
 &= \int \mathbb{E}_{Q_t} \left[ f(x) g(X_{t(x)}) e^{-V(x_t) - V(x) - \int_0^t v(x_s) ds} \right] dx
 \end{aligned}$$

Girsanov

$$\downarrow \int \mathbb{E} \left[ f(x) g(x + \beta_t) e^{-V(x + \beta_t) - V(x) - \int_0^t v(x + \beta_s) ds} \right] dx$$

$$[y = x + \beta_t]$$

$$\downarrow \int \mathbb{E} \left[ f(y - \beta_t) g(y) e^{-V(y) - V(y - \beta_t) - \int_0^t v(y - \beta_s + \beta_t) ds} \right] dy$$

$$(B_{t-s} - \beta_t)_{s \in [0,t]} \stackrel{(d)}{=} (B_s)_{s \in [0,t]}$$

$$\int \mathbb{E} \left[ f(y + \beta_t) g(y) e^{-V(y) - V(y + \beta_t) - \int_0^t v(y + \beta_s) ds} \right] dy$$

$$= \int g(x) E(f(X_t(x))) e^{-2U(x)} dx$$

Exercise: Prove in the same way that

$\forall \Phi: C([0, t]; \mathbb{R}^d) \rightarrow \mathbb{R}$  measurable bounded:

$$\int [E[\Phi(X_{t-s}), s \in [0, t)] e^{-2U(x)} dx$$

$$= \int [E[\Phi(X_s), s \in [0, t)] e^{-2U(x)} dx$$

We have obtained that the

transition semigroup  $P_t f(x) = E(f(X_t(x)))$

is symmetric in  $L^2(\mathbb{R}^d, e^{-2U(x)} dx)$

Moreover: if  $g = 1$

$$\int (P_t f) e^{-2U} dx = \int f e^{-2U} dx$$

namely  $e^{-2V} dx$  is an invariant

measure for  $(X_t(x))_{t \geq 0, x \in \mathbb{R}^d}$ .

If  $\int e^{-2V} dx < +\infty$  then

$$\nu := \frac{1}{\int e^{-2V} dx} \cdot e^{-2V(x)} dx \text{ is}$$

a probability measure and if

$X_0 \sim \nu$ ,  $(X_t)_{t \geq 0}$  is stationary.

$$X_t \sim \nu \quad \forall t \geq 0.$$

NOTE: if  $\nabla U$  is not bounded,  
then for the Novikov criterion we  
need additional information.

For example: if  $d=1$  and

$$U(x) = \lambda \frac{x^4}{4}, \quad U' = \lambda x^3,$$

then everything works if  $\lambda \geq 0$ ,  
but not if  $\lambda < 0$ .

In the latter case, the solution  
to the SDE even explodes  
in finite time.