

1<sup>st</sup> lecture:  $-\lambda \sum_{k=1}^N \delta(x_k)^2$   
 $\lambda > 0, \mathbb{E} \left[ e^{-\lambda \sum_{k=1}^N \delta(x_k)^2} \right] = \left( \frac{1}{2\lambda + 1} \right)^N$

rest of the argument works the same

Linear SPDE

$\partial_t u = \partial_x^2 u + \xi$       $x \in (0,1), t > 0$   
 $\xi$  space-time W.N.

Weak form      $W_t(\varphi) = \langle \xi, \mathbb{1}_{[0,t]} \varphi \rangle$

$\varphi \in C_c^\infty(0,1)$   
 $\langle u_t, \varphi \rangle = \langle u_0, \varphi \rangle + \int_0^t \langle u_s, \varphi'' \rangle ds + W_t(\varphi)$

Orthonormal basis of  $L^2(0,1)$

$e_n(x) = \sqrt{2} \sin(n\pi x)$       $x \in (0,1), n \geq 1$

Taking  $\varphi = e_n$       $\hat{u}_t(n) = \langle u_t, e_n \rangle$   
 $\hat{u}_t(n) = \hat{u}_0(n) - \int_0^t \hat{u}_s(n) ds + W_t(e_n)$

Since  $(W_t(e_n), t \geq 0)_{n \geq 1}$  iid sequence of B.M.  
 This way, we get a collection of SPDE's driven by independent B.M.s.

$\forall n \geq 1, (\hat{u}_t(n), t \geq 0)$  is an Ornstein-Uhlenbeck process.

$dX_t = -\lambda X_t dt + dB_t$       $X_0 = x \in \mathbb{R}$

Solution to such an SDE:

$X_t = e^{-\lambda t} x + \int_0^t e^{-\lambda(t-s)} dB_s$

(Apply Itô to  $e^{\lambda t} X_t$ )

Fact:  $X_t \sim \mathcal{N}(e^{-\lambda t} x, \frac{1-e^{-2\lambda t}}{2\lambda})$

$\forall x \in \mathbb{R}, X_t \xrightarrow[t \rightarrow 0]{} \mathcal{N}(0, \frac{1}{2\lambda})$

Inv. measure for  $(X_t, t \geq 0)$ :

$\mu$  on  $\mathbb{R}$       $\int \mathbb{P}_t f(x) \mu(dx) = \int f(x) \mu(dx)$

where  $\mathbb{P}_t$  semigroup,  $f \in C_b(\mathbb{R})$ .

Prop:  $\mu_\lambda = \mathcal{N}(0, \frac{1}{2\lambda})$  is the unique inv. measure for  $X$ .

Proof:  $\forall x \in \mathbb{R}, \int f d\mu_\lambda = \lim_{t \rightarrow 0} \mathbb{P}_t \int f(x) \mu_\lambda(dx)$

$= \lim_{t \rightarrow 0} \mathbb{P}_t \circ \mathbb{P}_t \int f(x) \mu_\lambda(dx) = \int \mathbb{P}_t f(x) \mu_\lambda(dx)$

So  $\mu_\lambda$  invariant.  
 Uniq, If  $\nu$  is inv.

$\int f d\nu = \int \mathbb{P}_t f(x) \nu(dx) \xrightarrow[t \rightarrow 0]{} \int f(x) \mu_\lambda(dx)$

$\forall n \geq 1, (\hat{u}_t(n), t \geq 0)$  admits  $\mu_\lambda$  its inv. meas.  $\square$

$\mu = \otimes_{n \geq 1} \mu_\lambda$  meas. on  $L^2(0,1)$ .  
 inv for SPDE  $\partial_t u = \partial_x^2 u + \xi$

Exercise: Show that  $\mu \sim$  law of the Brownian bridge on  $(0,1)$

Goal  $\int_0^t u = \frac{1}{2} \sigma^2 u + u \cdot \xi$

→ Stochastic integrals.

Even for SDE's,  $\int_0^t X = X_t dB_t$

Issue with  $\int B_s dB_s$   $B \sim \mathcal{O}(\sqrt{t})$

Th (Young)  $\mathcal{C}^\alpha \times \mathcal{C}^\beta \rightarrow \mathcal{C}^{\alpha+\beta}$   
 $\mathcal{C}^\alpha$  Hilbert space  $(dB_s, B_s) \mapsto B_s dB_s$   
 this map is continuous  
 $\| \int_0^t \alpha + \beta > 0$

In the case of BM.  $\alpha = \frac{1}{2}$   $\beta = \frac{1}{2}$   
 $\alpha + \beta < 0$ .

No canonical way of defining  $B_s dB_s$

→ Need another way to define  $\int B_s dB_s$

→ Different ways of doing that: Itô & Stratonovich

$\xi$  white noise on  $\mathbb{R}_t \times \mathbb{R}^d$   
 $W_t = \sum_{n=1}^d W_t^n e_n$   $(e_n)_{n=1}^d$  basis  $L^2(\mathbb{R}^d)$   
 $W_t(\cdot) = \int_0^t \xi(\cdot, \tau) d\tau$   
 $(\mathbb{F}_t, t \geq 0)$  natural filtration associated to  $(W_t, t \geq 0)$ .

Elementary param  $f: (t, x, \omega) \rightarrow \sum_{k=1}^d \xi_k(t, x) \frac{\partial}{\partial x_k} \mathbb{1}_A$   
 for  $0 \leq a < b$ ,  $Z$  bounded,  $\mathbb{F}$ -meas.  
 $A \in \mathcal{B}(\mathbb{R}^d)$ ,  $\text{Leb}(A) < \infty$   
 Linear combinations of elem. param.  
 Simple processes.

def Let  $\mathcal{P}_T$  be the Banach space of  $f: \mathbb{R}_+ \times \mathbb{R}^d \times \Omega \rightarrow \mathbb{R}$  st.

- (1)  $f$  meas. in  $\mathbb{F}$ -field generated by all simple processes.
  - (2)  $\exists (f_n)_{n \geq 1}$  simple processes st.:
- $$\mathbb{E} \left[ \int_0^T \int_{\mathbb{R}^d} (f - f_n)^2(t, x, \omega) dt dx \right] \xrightarrow{n \rightarrow \infty} 0$$

For elementary param  $f = \sum_{k=1}^d \xi_k \frac{\partial}{\partial x_k} \mathbb{1}_A$   
 $\int_0^T \int_{\mathbb{R}^d} f(t, x) \xi(t, x) dx = Z \cdot (W_{T \wedge A} - W_{0 \wedge A})$   
 $\int_0^T \int_{\mathbb{R}^d} f(t, x) dW_t(dx)$

Lemma  $(\int_0^T \int_{\mathbb{R}^d} f(t, x) dW_t(dx), T > 0)$   
 is an  $\mathbb{F}_T$ -martingale, in  $L^2(\Omega)$ .  
 $\mathbb{E} \left[ \left( \int_0^T \int_{\mathbb{R}^d} f(t, x) dW_t(dx) \right)^2 \right] = \int_0^T \int_{\mathbb{R}^d} \mathbb{E} [f(t, x)^2] dt dx$   
 Proof: exercise.

def/Prop Let  $f \in \mathcal{P}_T$ .

We define  $\int_0^T \int_{\mathbb{R}^d} f(t,x) dW_t(dx)$

$$= \lim_{n \rightarrow \infty} \int_0^T \int_{\mathbb{R}^d} f_n(t,x) dW_t(dx)$$

where  $(f_n)_{n \geq 1}$  seq. of simple process  $\mathcal{P}_T \rightarrow f$ .

This is an  $\mathcal{F}_T$ -martingale, and

$$\mathbb{E} \left[ \int_0^T \int_{\mathbb{R}^d} f(t,x) dW_t(dx) \right]^2 = \int_0^T \int_{\mathbb{R}^d} \mathbb{E} [f(t,x)^2] dt dx$$

Proof: Limit  $L^2$  ✓

Remark: limit of martingales.

$$B \in \mathcal{F}_t \quad \mathbb{E} [M_{t+s}^n | \mathcal{F}_t] = \mathbb{E} [M_t^n | \mathcal{F}_t]$$

$$\mathbb{E} [M_{t+s}^n | \mathcal{F}_t] = \mathbb{E} [M_t^n | \mathcal{F}_t] \checkmark$$

Define

$$(*) \begin{cases} \partial_t u = \frac{1}{2} \partial_x^2 u + u \cdot \xi \\ u(0, \cdot) = u_0(\cdot) \\ x \in \mathbb{R} \end{cases}$$

Th: Let  $u_0: \mathbb{R} \rightarrow \mathbb{R}$  in  $L^2(\mathbb{R}, dx)$ .

There exists a unique solution to (\*)

which lives in  $\mathcal{P}_T$ , for any  $T > 0$ .

Proof \* Existence (mild form)

Look for a fixed point of the map

$$\mathcal{M}_{T, u_0}: \mathcal{P}_T \rightarrow \mathcal{P}_T$$

$$v \mapsto \begin{cases} (t,x) \mapsto \int_{\mathbb{R}} P_t(x-y) u_0(y) dy \\ + \int_0^t \int_{\mathbb{R}} P_{t-s}(x-y) v(s,y) \xi ds dy \end{cases}$$