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Introduction to Stochastic PDEs

Lecture 1

$$du = (\underbrace{Au}_{\text{diff op}} + \underbrace{f(u)}_{\text{nonlinearity}}) dt + \underbrace{\phi dw}_{\text{additive noise}}$$

$dw$  - white noise space-time

First some examples

Ex: Burgers:  $A = \partial_{xx} u$ , periodic + Dirichlet bdy con  
 $\phi = I$

Ex: KPZ: Same as Burgers except  $\phi = \partial_x$   
Very difficult problem  
 $\rightarrow \phi = |\partial_x|^\alpha$ ,  $\alpha$ -small

Ex Reaction Diffusion:  $G \subset \mathbb{R}^d$  smooth bdd domain  
 $A = \Delta + b.c.$   $d=1, \phi = Id \rightarrow$  easy  
 $f(u) = -u^3 + \lambda u$   $d=2, \phi = Id \rightarrow$  no so easy  
 $d=3, \phi = Id \rightarrow$  MH 2014

Ex Navier-Stokes:  $G \subset \mathbb{R}^d$ ,  $u(x,t) \in \mathbb{R}^d$ ,  $x \in G$ ,  $t > 0$

$$\begin{cases} du = (\nu \Delta u + (u \cdot \nabla)u + \nabla p) dt + \phi dw \\ \operatorname{div} u = 0 \\ bc \end{cases}$$

Other eqns:

- Same  $\phi = \phi(u)$
- Nonlinear stochastic Schrödinger

$$u \in \mathbb{C}, x \in \mathbb{R}^d, t > 0$$
$$i du = (\Delta u \pm |u|^\alpha u) dt + u \circ dW$$

Brownian Motion, Wiener processes

$(B(t))_{t \geq 0}$  is a B.M. if

$$B(t) = B(t, \omega), \quad \omega \in \Omega$$

where  $(\Omega, \mathcal{F}, \mathbb{P})$  probability space

- $B(0) = 0$
- $\forall \omega \in \Omega, t \mapsto B(t)$  is cts
- $B$  is a Gaussian process, i.e.  $\forall t_1, \dots, t_n \geq 0$ ,  
 $(B(t_1), \dots, B(t_n))$  is a Gaussian vector.
- $B(t) \sim N(0, t)$
- $B(t) - B(s)$  indep of  $B(s)$ ,  $s \leq t$   
future is indep of ~~the~~ present and past.
- $\mathbb{E}(|B(t) - B(s)|^2) = t - s \rightarrow$  this implies  $B$  DNE  
in the classical sense  
 $B \in C_t^{\frac{1}{2}-\varepsilon}$  a.s.  
 $B \notin C^{\frac{1}{2}}, B \notin BV$  a.s.

Filtration:  $(\mathcal{F}_t)_{t \geq 0}$   $\sigma$ -field

$$\mathcal{F}_s \subset \mathcal{F}_t \subset \mathcal{F} \text{ for } s \leq t.$$

$X$  a random variable ~~is~~:  $X$  is  $\mathcal{F}_t$  measurable if  
"X depends on the past up to t"

(ad)

We ask:  $\forall t, B(t)$  is  $\mathcal{F}_t$ -measurable.

Remarks:

- Formally,  $(f_j)_{j \in \mathbb{N}}$  a Complete orthonormal System (CONS) of  $L^2(\mathbb{R}^+)$

$$\dot{B}(t, \omega) = \sum_j \chi_j(\omega) f_j(t) \quad \chi_j \text{ are indep } N(0,1)$$

→ time white noise

• Formally:  $\mathbb{E}(\dot{B}(t) \dot{B}(s)) = \delta_{t,s}$   
 $\mathbb{E} \int \int B(t) B(s) \phi(t) \psi(s) dt ds = \mathbb{E} \int \phi(t) \psi(t) dt$

1) Wiener process:  $H$  a Hilbert space, a CONS  $(e_k)_{k \in \mathbb{N}}$   
 $(B_k)_k$  a sequence of BM, indep,  
 satisfying (ad) → they are adapted to  $(\mathcal{F}_t)$   
 Let's say  $H = L^2(G)$ .

→ function of  $t, x, \omega$

$$W = \sum_k B_k(t, \omega) e_k(x) \quad \text{a cylindrical Wiener process in } H$$

Problem:

$$\mathbb{E}(|W|_H^2) = \sum_k \mathbb{E}(|B_k(t)|^2) = \sum_k t = \infty$$

$H \hookrightarrow U$  a Hilbert space

$$\mathbb{E}(|W(t)|_U^2) = \sum_{k,l} \mathbb{E}(B_k(t) B_l(t)) (e_k, e_l)$$

$$= \sum_k t |e_k|_H^2 < \infty \text{ iff } H \hookrightarrow U \text{ Hilbert-Schmidt}$$

Ex:  $H = L^2(G), U = H^s(G)$

$$H \xrightarrow{\text{H.S.}} U \text{ iff } s < -d/2$$

Remark:  $\forall k, \dot{B}_k = \sum_j \chi_{j,k} f_j \quad \chi_{j,k} \text{ ind } N(0,1)$

$$\dot{W} = \sum_{j,k} f_j e_k$$

Space time white noise

Formally:  $\mathbb{E}(\dot{W}(t,x) \dot{W}(s,y)) = \delta_{t,s} \delta_{x,y}$

We often smooth the noise.

Correlated noise:  $K$ -Hilbert space,  $\phi \in \mathcal{L}(H, K)$

$$\phi dW = \sum \phi e_k dB_k. \quad \phi W \text{ lives in } K \text{ iff}$$

$$\phi \in \mathcal{L}_2^K(H, K)$$

$\leftarrow$  Hilbert-Schmidt

## 2) Stochastic Integral

$\int_0^t \phi(s) dW(s)$  where  $\phi$  has values in  $\mathcal{L}(H, W)$ .

$(e_k)$  - CONS of  $H$ ,  $(f_j)$  - CONS of  $K$

$$\int_0^t \phi(s) dW(s) = \sum_{k,l} \underbrace{\int_0^t (\phi(s) e_k, f_l) dB_k(s) f_l}_{\text{Standard Ito-integral in dim 1}}$$

convergence of a series in  $K$

$$\phi_{k,l}(s) = (\phi(s) e_k, f_l), \quad \int_0^t \phi_{k,l}(s) dB_k(s), \quad B_k \neq B_l$$

$$\text{Ito int} \approx \sum_i \phi_{k,l}(B_k(t_{i+1}) - B_k(t_i))$$

$(t_i)$  subdivision of  $[0, t]$

assume  $\phi_{k,l}$  deterministic

$$\mathbb{E} \left( \int_0^t \phi_{k,l}(s) dB_k(s) \right) = 0, \quad \mathbb{E} \left( \left( \int_0^t \phi_{k,l}(s) dB_k(s) \right)^2 \right) =$$

$$2 \sum_i \phi_{k,l}(t_i) \phi_{k,l}(t_j) (B(t_{i+1}) - B(t_i)) (B(t_{j+1}) - B(t_j))$$

$$+ \sum_i \mathbb{E} \phi_{k,l}^2(t_i) (B(t_{i+1}) - B(t_i))^2$$

This is okay if  $\phi$  is adaptive.

$$\phi_{k,l} \in \mathcal{L}^2(\Omega \times [0, t]) \xrightarrow{\text{isometry}} \int_0^t \phi_{k,l}(s) dB_k \in \mathcal{L}^2(\Omega)$$

if  $\phi_{k,l}$  adaptive  $\rightarrow$  in fact "predictable"

$$\mathbb{E} \left( \int_0^t \phi(s) dW(s) \right) = 0,$$

$$\begin{aligned} \mathbb{E} \left( \left| \int_0^t \phi(s) dW(s) \right|_K^2 \right) &= \sum_k \mathbb{E} \left| \sum_k \int_0^t (\phi(s) e_k, f_k) dB_k \right|^2 \\ &= \sum_{k,k} \mathbb{E} \left( \int_0^t (\phi(s) e_k, f_k) dB_k \right)^2 \\ &\quad \text{(cross terms vanish)} \\ &= \sum_{k,k} \int_0^t \mathbb{E} (\phi(s) e_k, f_k)^2 ds \\ &= \mathbb{E} \int_0^t |\phi(s)|_{\mathcal{L}_2(H,K)}^2 ds \end{aligned}$$

Remark: If  $\int_0^t \phi(s) \circ dB(s) = \sum_i \phi \left( \frac{t_i + t_{i+1}}{2} \right) (B(t_{i+1}) - B(t_i))$   
 Stratonovich

Not defined for as many processes, but more natural in physics.

### 3) Linear Heat Eqn

$\mathcal{O} \subset \mathbb{R}^d$  open bdd set on  $[0,1]^d$

$$A = \Delta + b.c.$$

$W$ : cylindrical Wiener process on  $H = L^2(\mathcal{O})$

$\Phi$ : Linear operator on  $H$

$$\begin{cases} du = A u dt + \Phi dW \\ u(0) = u_0 \end{cases}$$

$\rightarrow$  soln is  $u(t) = e^{At} u_0 + \int_0^t e^{A(t-s)} \Phi dW$   
 is the only weak soln (PDE sense)  $\forall K$  if the heat eqn is well-posed in  $K^*$ .

Question: What is the smoothness of  $u$ ?

Ex:  $\Phi = I$

$u$  will live in  $H^{2B}(G)$  iff  $\int_0^t |e^{A(t-s)}|^2 ds < \infty$   
 $L_2(H, H^{2B}(G))$

$$|v|_{H^{2B}} \sim |(-A)^B v|$$

$$\int_0^t |(-A)^B e^{A(t-s)}|^2 ds = \sum_k \int_0^t \lambda_k^{2B} e^{-2\lambda_k(t-s)} ds$$

where  $(\lambda_k)$  are eigenvalues of  $-A$

$$= \sum_k \frac{1}{2\lambda_k} (1 - e^{-2\lambda_k t})$$

it is known in many cases  
that  $\lambda_k \sim ck^{2/d}$

$< \infty$

$$\Leftrightarrow \text{iff } \frac{2}{d}(2B-1) < -1$$

$d=1 \Rightarrow B < 1/4 \rightarrow u$  has  $1/2$ -reg  
 $d=2 \Rightarrow B < 0 \rightarrow u$  has 0-reg  
 $d=3 \Rightarrow B < -1/4$

Case  $d=1$ :  $G = (0,1)$ ,  $u_0 = 0$

$$u(t) = \int_0^t e^{A(t-s)} dW(s)$$

$$u(t, x) = \sum_k \int_0^t (e^{A(t-s)} e_k)(x) dB_k(s)$$

$$\mathbb{E}(|u(t, x) - u(s, y)|^2) \leq C(|t-s|^{1/2} + |x-y|)$$

$$e_k = \sin(kx)$$

Thus  $u \in C_t^{1/4-} C_x^{1/2-}$  a.s.

$$\mathbb{E}(|u|_{C_t^{1/4-} C_x^{1/2-}}^p) < \infty$$