

Stochastic Partial Differential Equations and Infinite Dimensional Analysis

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joint work (several papers) with:

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Ref.: BiBoS-Preprint Server, my homepage at Purdue University
(BiBoS = Bielefeld–Bonn–Stochastics Research Centre)

**A – From ODE to PDE
in finitely many variables**

ODE

$$(1) \quad \begin{aligned} dX_t^x &= B(X_t^x) dt \\ X_0^x &= x \in \mathbb{R}^d \end{aligned} \quad \text{on } \mathbb{R}^d$$

$$X_t^x = x + \int_0^t B(X_s^x) ds$$

$$\Rightarrow \begin{aligned} p_t f(x) &:= f(X_t^x) \\ p_0 f &= f \quad \text{for } f : \mathbb{R}^d \rightarrow \mathbb{R} \end{aligned}$$

semigroup
 $(p_{t+s}f)(x) = p_t(p_s f)(x)$
 by flow property

solves **PDE** (2)

$$\Rightarrow \begin{aligned} p_t f(x) &:= f(X_t^x) \\ p_0 f &= f \quad \text{for } f : \mathbb{R}^d \rightarrow \mathbb{R} \end{aligned}$$

semigroup
 $(p_{t+s}f)(x) = p_t(p_s f)(x)$
 by flow property

solves **PDE**

$$(2) \quad \frac{\partial}{\partial t} p_t f(x) = \sum_{i=1}^d B^i(x) \cdot \frac{\partial}{\partial e_i} p_t f(x)$$

$$\begin{aligned} (1) \quad &= B(x) \cdot \nabla_x p_t f(x) \\ &=: L(p_t f)(x) \\ &\quad \uparrow \\ &\quad \text{"generator"} \\ &\quad \text{of (1)} \end{aligned}$$

Here $B = (B^1, \dots, B^d) : \mathbb{R}^d \rightarrow \mathbb{R}^d$ and $e_i = (0, \dots, \underset{\substack{\uparrow \\ i^{\text{th}}}}{1}, \dots, 0)$ canonical basis

ODE

$$(1) \quad \begin{aligned} dX_t^x &= B(X_t^x) dt \\ X_0^x &= x \in \mathbb{R}^d \end{aligned} \quad \text{on } \mathbb{R}^d$$

$$X_t^x = x + \int_0^t B(X_s^x) ds$$

$$\Rightarrow \begin{aligned} p_t f(x) &:= f(X_t^x) \\ p_0 f &= f \quad \text{for } f : \mathbb{R}^d \rightarrow \mathbb{R} \end{aligned}$$

semigroup
 $(p_{t+s}f)(x) = p_t(p_s f)(x)$
 by flow property

solves **PDE** (2)

SODE

$$(1) \quad \begin{aligned} dX_t^x &= B(X_t^x) dt + dW_t \\ X_0^x &= x \in \mathbb{R}^d \end{aligned} \quad \text{on } \mathbb{R}^d$$

$$X_t^x(\omega) = x + \int_0^t B(X_s^x(\omega)) ds + \underbrace{W_t(\omega)}_{\text{Brownian motion on } \mathbb{R}^d}$$

Kolmogorov \iff

$$p_t f(x) := \int f(X_t^x(\omega)) \mathbb{P}(d\omega) =: \mathbb{E}[f(X_t^x)]$$

$$p_0 f = f \quad \text{for } f : \mathbb{R}^d \rightarrow \mathbb{R}$$

semigroup
 $(p_{t+s} f)(x) = p_t(p_s f)(x)$
 by ~~flow property~~
 Markov property

solves PDE (2) (heat equation in finitely many variables)

$$\Rightarrow \begin{aligned} p_t f(x) &:= f(X_t^x) \\ p_0 f &= f \quad \text{for } f : \mathbb{R}^d \rightarrow \mathbb{R} \end{aligned}$$

semigroup
 $(p_{t+s}f)(x) = p_t(p_s f)(x)$
 by flow property

solves **PDE**

$$(2) \quad \frac{\partial}{\partial t} p_t f(x) = \sum_{i=1}^d B^i(x) \cdot \frac{\partial}{\partial e_i} p_t f(x)$$

$$(2) \quad = B(x) \cdot \nabla_x p_t f(x)$$

$$=: L(p_t f)(x)$$

↑
 “generator”
 of (1)

Here $B = (B^1, \dots, B^d) : \mathbb{R}^d \rightarrow \mathbb{R}^d$ and $e_i = (0, \dots, \underset{\substack{\uparrow \\ i^{\text{th}}}}{1}, \dots, 0)$ canonical basis

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semigroup
 $(p_{t+s} f(x) = p_t(p_s f)(x))$
 by ~~flow property~~
 Markov property

solves **PDE** (heat equation in finitely many variables)

$$(2) \quad \frac{\partial}{\partial t} p_t f(x) \stackrel{\text{It\^o}}{=} \sum_{i=1}^d B^i(x) \cdot \frac{\partial}{\partial e_i} p_t f(x) + \frac{1}{2} \sum_{i=1}^d \frac{\partial^2}{\partial e_i \partial e_i} p_t f(x)$$

$$(3) \quad = B(x) \cdot \nabla_x p_t f(x) + \frac{1}{2} \Delta_x p_t f(x)$$

$$=: L(p_t f)(x)$$

\uparrow
 “generator”
 of (1)

Here $B = (B^1, \dots, B^d) : \mathbb{R}^d \rightarrow \mathbb{R}^d$ and $e_i = (0, \dots, \underset{i^{\text{th}}}{1}, \dots, 0)$ canonical basis

Itô:

$$\begin{aligned}
 p_t f(x) &= E[f(X_t^x)] \\
 &= f(x) + \underbrace{\sum_{i=1}^d E \left[\int_0^t \left(\frac{\partial}{\partial e_i} f \right) (X_s^x) dW_s^i \right]}_{=0} + \sum_{i=1}^d \int_0^t E \left[\underbrace{\left(\frac{\partial}{\partial e_i} f \right) (X_s^x) B^i(X_s^x)}_{p_s \left(\left(\frac{\partial}{\partial e_i} f \right) B^i \right) (x) ds} \right] ds \\
 &\quad + \underbrace{\frac{1}{2} \sum_{i,j=1}^d \int_0^t E \left[\left(\frac{\partial^2}{\partial e_i \partial e_j} f \right) (X_s^x) \underbrace{\langle dW_s^i, dW_s^j \rangle_{\mathbb{R}^d}}_{=\delta_{ij} ds} \right]}_{=p_s \left(\frac{\partial^2}{\partial e_i \partial e_j} f \right) (x) \delta_{ij} ds} \quad \text{(Taylor up to order 2!)} \\
 &= f(x) + \int_0^t \underbrace{p_s(Lf)(x)}_{L(p_s f)(x)} ds
 \end{aligned}$$

SODE

$$(1) \quad \begin{aligned} dX_t^x &= B(X_t^x) dt + dW_t \\ X_0^x &= x \in \mathbb{R}^d \end{aligned} \quad \text{on } \mathbb{R}^d$$

$$X_t^x(\omega) = x + \int_0^t B(X_s^x(\omega)) ds + \underbrace{W_t(\omega)}_{\text{Brownian motion on } \mathbb{R}^d}$$

Kolmogorov \iff

$$p_t f(x) := \int f(X_t^x(\omega)) \mathbb{P}(d\omega) =: \mathbb{E}[f(X_t^x)]$$

$$p_0 f = f \quad \text{for } f : \mathbb{R}^d \rightarrow \mathbb{R}$$

semigroup
 $(p_{t+s} f)(x) = p_t(p_s f)(x)$
 by ~~flow property~~
 Markov property

solves **PDE** (2) (heat equation in finitely many variables)

SODE

$$(1) \quad \begin{aligned} dX_t^x &= B(X_t^x) dt + \sigma(X_t^x) dW_t \\ X_0^x &= x \in \mathbb{R}^d \end{aligned} \quad \text{on } \mathbb{R}^d$$

$$X_t^x(\omega) = x + \int_0^t B(X_s^x(\omega)) ds + \int_0^t \sigma(X_s^x(\omega)) \underbrace{dW_s(\omega)}_{\substack{\text{Brownian motion} \\ \text{on } \mathbb{R}^d}}$$

Kolmogorov
 \iff

$$p_t f(x) := \int f(X_t^x(\omega)) \mathbb{P}(d\omega) =: \mathbb{E}[f(X_t^x)]$$

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solves **PDE** (heat equation in finitely many variables)

$$(2) \quad \frac{\partial}{\partial t} p_t f(x) \stackrel{\text{It\^o}}{=} \sum_{i=1}^d B^i(x) \cdot \frac{\partial}{\partial e_i} p_t f(x) + \frac{1}{2} \sum_{i=1}^d \frac{\partial^2}{\partial e_i \partial e_i} p_t f(x)$$

$$(4) \quad = B(x) \cdot \nabla_x p_t f(x) + \frac{1}{2} \text{Tr}(D^2 p_t f(x))$$

$$=: L(p_t f)(x)$$

\uparrow
 “generator”
 of (1)

Here $B = (B^1, \dots, B^d) : \mathbb{R}^d \rightarrow \mathbb{R}^d$ and $e_i = (0, \dots, \underset{i^{\text{th}}}{1}, \dots, 0)$ canonical basis

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semigroup
 $(p_{t+s}f(x) = p_t(p_s f)(x))$
 by ~~flow property~~
 Markov property

solves **PDE** (heat equation in finitely many variables)

$$(2) \quad \frac{\partial}{\partial t} p_t f(x) \stackrel{\text{It\^o}}{=} \sum_{i=1}^d B^i(x) \cdot \frac{\partial}{\partial e_i} p_t f(x) + \frac{1}{2} \sum_{i,j=1}^d (\sigma^T(x)\sigma(x))^{ij} \cdot \frac{\partial^2}{\partial e_i \partial e_j} p_t f(x)$$

$$(5) \quad = B(x) \cdot \nabla_x p_t f(x) + \frac{1}{2} \text{Tr}(\sigma^T(x)\sigma(x)D^2 p_t f(x))$$

$$=: L(p_t f)(x)$$

\uparrow
 “generator”
 of (1)

Here $B = (B^1, \dots, B^d) : \mathbb{R}^d \rightarrow \mathbb{R}^d$ and $e_i = (0, \dots, \underset{i^{\text{th}}}{1}, \dots, 0)$ canonical basis

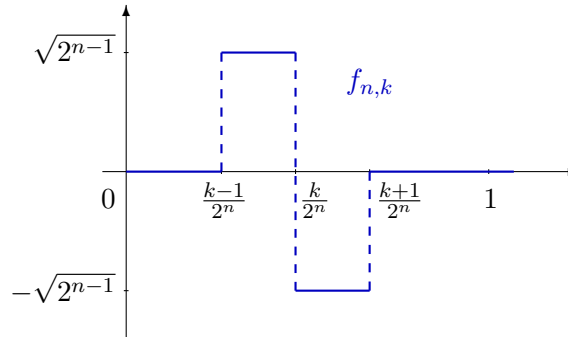
and $\sigma = (\sigma^{ij}) : \mathbb{R}^d \rightarrow \underbrace{M(d \times d)}_{d \times d\text{-matrices}}$

[Levy–Wiener–Ciesielski]

First ingredient: Haarbasis of $L^2([0, 1], dt)$:

↑
Lebesgue
measure

$f_{0,0} \equiv 1$, and for $n \in \mathbb{N}$, $0 < k < 2^n$, k odd,



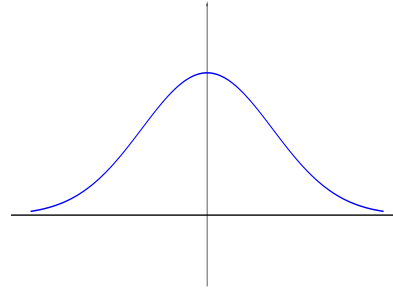
$(f_{n,k})_{\substack{0 < k < 2^n, k \text{ odd} \\ n \in \mathbb{N}}}$ is ONB of $L^2([0, 1], dt)$

Second ingredient: Standard normal distribution on \mathbb{R}^∞

Standard normal distribution on \mathbb{R}^1 :

$$\gamma(dx) := \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{x^2}{2}} dx$$

↑
Gauss
↑
Lebesgue meas.
on \mathbb{R}^1



Set $\gamma_{n,k} := \gamma$.

$$\mathbb{P} := \bigotimes_{\substack{0 < k < 2^n \\ k \text{ odd} \\ n \in \mathbb{N}}} \gamma_{n,k} \quad \text{product measure on } \mathbb{R}^\infty \quad (= \mathbb{R}^{\{(n,k)|\dots\}})$$

Define $\xi_{n,k} : \mathbb{R}^{\{(n,k)|n \in \mathbb{N}, 0 < k < 2^n, k \text{ odd}\} \cup \{(0,0)\}} \rightarrow \mathbb{R}$ (projection) and for $t \in [0, 1]$

$$W_t(\omega) := \sum_{(n,k)} \xi_{n,k}(\omega) \int_0^t f_{n,k}(s) ds \quad \text{converges for } \mathbb{P}\text{-a.e. } \omega \in \mathbb{R}^\infty$$

↑

Brownian motion on \mathbb{R}^1

**B – From SODE to PDE
in infinitely many variables**



$$p_t f(x) := \int_{\Omega} f(X_t^x(\omega)) \mathbb{P}(d\omega), \quad p_0 f = f, \quad \text{for } f : E \rightarrow \mathbb{R}$$

solves **PDE** (heat equation in infinitely many variables)

$$(2) \quad \frac{\partial}{\partial t} p_t f(x) = \sum_{i=1}^{\infty} \langle B(x), e_i \rangle \frac{\partial}{\partial e_i} p_t f(x) + \frac{1}{2} \sum_{i,j=1}^{\infty} \underbrace{\langle \sigma^T(x) \sigma(x) e_i, e_j \rangle}_{=: A(x)} \frac{\partial^2}{\partial e_i \partial e_j} p_t f(x) \text{ (heuristically!)}$$

$$=: L(p_t f)(x).$$

\uparrow
 “generator”
 of (1)

Here $B : E \rightarrow E$ and $\{e_i \mid i \in \mathbb{N}\}$ ONB of E and

$$\sigma : E \rightarrow L(E)$$

\uparrow
 (bounded) linear operators
 on E

For simplicity $A(x) = \sigma^T(x)\sigma(x) = A$ independent of $x \in E$. So, have:

$$(1) \quad \begin{aligned} & \begin{array}{c} \sigma(X_t) \text{ above} \\ \swarrow \quad \searrow \\ \sqrt{A} \end{array} \in L(E), \text{ pos.} \\ dX_t &= B(X_t)dt + \sqrt{A} dW_t \quad \leftarrow \text{B.M. on } E \\ X_0 &= x \in E \quad = \text{sep. Hilbert space with } \langle \cdot, \cdot \rangle \end{aligned}$$

Associated generator (Kolmogorov operator)

$$\begin{aligned} L\varphi(x) &= \sum_{i=1}^N \langle B(x), e_i \rangle \frac{\partial \varphi}{\partial e_i}(x) + \frac{1}{2} \sum_{i,j=1}^N \langle Ae_i, e_j \rangle \frac{\partial^2}{\partial e_i \partial e_j} \varphi(x) \\ &= \langle B(x), D\varphi(x) \rangle + \frac{1}{2} \text{Tr}(AD^2\varphi(x)) \end{aligned}$$

↙ Fréchet derivatives ↘

for $x \in E$ and $\varphi = g(\langle e_1, \cdot \rangle, \dots, \langle e_N, \cdot \rangle) \leftarrow \text{all such } \mathcal{FC}_b^2$

$\uparrow \in C_b^2(\mathbb{R}^N)$ $\uparrow \in \mathbb{N}$ arbitrary

ONB of E

Altogether:

$$(1) \quad \begin{aligned} dX_t &= B(X_t)dt + \sqrt{A} dW_t \\ X_0 &= x \in E \end{aligned}$$

Associated generator (**Kolmogorov operator**)

$$\begin{aligned} L\varphi(x) &= \sum_{i=1}^N \langle B(x), e_i \rangle \frac{\partial \varphi}{\partial e_i}(x) + \frac{1}{2} \sum_{i,j=1}^N \langle Ae_i, e_j \rangle \frac{\partial^2}{\partial e_i \partial e_j} \varphi(x) \\ &= \langle B(x), D\varphi(x) \rangle + \frac{1}{2} \text{Tr}(AD^2\varphi(x)) \end{aligned}$$

for $x \in E$ and $\varphi = g(\langle e_1, \cdot \rangle, \dots, \langle e_N, \cdot \rangle) : E \rightarrow \mathbb{R}$.

The associated heat equation is also called **Kolmogorov** (backward) **equation**

$$(2) \quad \frac{\partial}{\partial t} u(t, x) = Lu(t, x), \quad u(0, \cdot) = f,$$

where $f : E \rightarrow \mathbb{R}$.

Want to solve (2), then (1)!

Once (2) is solved one has to apply highly developed machinery to get solution of (1).

In this talk we concentrate on solving (2):

Two approaches to solve (2) will be presented:

- L^p -approach
- Weighted function space (=WFS -) approach

C – Two stochastic PDE as examples

- (a) Porous media equation (L^p -approach)
- (b) Stochastic Navier–Stokes equation, $d = 2$
(WFS-approach)

(a) Stochastic porous media equation (L^p -approach)

with Dirichlet boundary conditions

$$dX_t = \underbrace{\Delta \Psi(X_t)}_{B(X_t)} dt + \sqrt{A} dW_t \quad \text{with } \Psi : \mathbb{R} \rightarrow \mathbb{R}$$

\downarrow
 \uparrow
 trace class

on $E := H^{-1}(\Lambda)$, $\Lambda \subset \mathbb{R}^d$, open;

so,

$$L\varphi(x) = \langle \Delta \Psi(x), D\varphi(x) \rangle + \frac{1}{2} \text{Tr} A D^2 \varphi(x), \quad x \in H^{-1}(\Lambda), \quad \text{for } \varphi : H^{-1}(\Lambda) \rightarrow \mathbb{R}.$$

Remark:

- (i) $A \equiv 0$: enormous literature.
- (ii) $A \not\equiv 0$: first papers [Da Prato / R.: JEE '04], [Barbu / Bogachev/Da Prato/R.: JFA '06] Subsequently, many others. Mainly, on SPDE, not on Kolmogorov equations: Kim, Wu, Zhang, ...
Among most recent: Da Prato/R./Rosowski/Wang: Comm. P.D.E. '06], [Ren/R./Wang: BiBos-preprint '06].

(c) Stochastic Navier–Stokes equation, $d = 2$ (WFS-approach)

$$dX_t = \left[\underbrace{\nu \Delta_s X_t - \langle X_t, \nabla \rangle_{\mathbb{R}^2} X_t}_{B(X_t)} \right] dt + \sqrt{A} dW_t,$$

on

viscosity
Stokes-Laplacian with Dirichlet boundary conditions
gradient on \mathbb{R}^2
trace class or even finite dim. range

$$E := \{x \in L^2(\Lambda \rightarrow \mathbb{R}^2, dx) \mid \underbrace{\operatorname{div} x}_{\text{in the sense of distributions}} = 0\},$$

$\Lambda \subset \mathbb{R}^2$, open, bounded, $\partial\Lambda$ smooth;
 so,

$$L\varphi(x) = \langle \nu \Delta_s x - \langle x, \nabla \rangle_{\mathbb{R}^2} x, D\varphi(x) \rangle_E + \frac{1}{2} \operatorname{Tr} AD^2\varphi(x), \quad x \in E, \quad \text{for } \varphi : E \rightarrow \mathbb{R}.$$

Remark:

- (i) $A \equiv 0$: **OVERWHELMING** literature
- (ii) $A \neq 0$: on SPDE: **OVERWHELMING** literature
 $A \neq 0$: on Kolmogorov equations: Da Prato/Debussche (also $d = 3!$), Barbu, Flandoli, Gozzi,...
WFS-approach: [R./Sobol: Ann. Prob. '06] for $d = 1.$, [R./Sobol: Preprint '06] for $d = 2$ and also for **geostrophic equation**
- (iii) Existence of **infinitesimally invariant measures** also proved for $d \geq 2$: [Bogachev / R.: PTRF '00]

**D – Strategies to solve
the Kolmogorov equation**

to solve

$$(2) \quad \frac{\partial}{\partial t} u(t, x) = Lu(t, x), \quad u(0, \cdot) = f.$$

semigroup approach!

Construct

$$e^{tL} f(x) =: u(t, x), \quad t \geq 0.$$

If e^{tL} exists, then by operator calculus

$$(\lambda - L)^{-1} = \int_0^{\infty} e^{-\lambda t} e^{tL} dt, \quad \lambda > \lambda_0.$$

So, try to construct $(\lambda - L)^{-1}$, $\lambda > \lambda_0$, and invert Laplace transform, (well-known method: “Hille-Yosida Theorem”).

For implementation two major steps necessary:

Step 1

Show “dissipativity”, i.e.

$$\|(\lambda - L)\varphi\|_{W(E)} \geq (\lambda - \lambda_0)\|\varphi\|_{W(E)} \quad \forall \varphi \in \mathcal{FC}_b^2, \quad \lambda > \lambda_0,$$

for suitable norm $\|\cdot\|_{W(E)}$ in Banach space $W(E)$ of functions $f : E \rightarrow \mathbb{R}^d$ such that $\mathcal{FC}_b^2 \subset W(E)$. So, $\lambda - L$ is invertible for all $\lambda > \lambda_0$.

Step 2

Show “density of range”, i.e. $(\lambda - L)(\mathcal{FC}_b^2)$ is dense in $(W(E), \|\cdot\|_{W(E)})$ for one (hence all) $\lambda > \lambda_0$. (Easier to achieve for weaker norms!)

Cannot take: $W = C_b(E)$, since coefficients of L **not** continuous in general.

In this talk:

Only Step 1 in

- **L^p - approach** for stochastic porous media equation.
Here $W(E) := L^p(E, \mu)$ for suitable measures on E !
- **WFS-approach** for stochastic Navier-Stokes equation
Here $W(E) :=$ weighted space of sequentially weakly continuous functions.

E - L^p -Approach

General idea of L^p -approach:

Step 1: Reference measures on E .

Solve $L^* \mu = 0$ “ μ is L -infinitesimally invariant”. (i.e. solve an elliptic problem first!)

i.e. find probability measure μ on $\overbrace{\mathcal{B}(E)}^{\text{Borel } \sigma\text{-algebra}}$ such that $L\varphi \in L^1(E, \mu)$ and

$$\int L\varphi \, d\mu = 0 \quad \forall \varphi \in \mathcal{FC}_b^2.$$

Then not hard to show: (L, \mathcal{FC}_b^2) is dissipative on $L^p(E, \mu)$
(so has closure $(\bar{L}, D(\bar{L}))$ on $L^p(E, \mu)$ for all $p \in [1, \infty)$)

Step 2:

Show: $(\lambda - L)(\mathcal{FC}_b^2)$ dense in $L^p(E, \mu)$

Then $\exists e^{t\bar{L}}, t > 0$, on $L^p(E, \mu)$ hence

$$L^p(E, \mu) - \frac{d}{dt} \underbrace{e^{t\bar{L}} f}_{u(t, \cdot)} = \bar{L} \underbrace{(e^{t\bar{L}} f)}_{u(t, \cdot)}, \quad t > 0, f \in D(\bar{L}), \quad \text{“solution in } L^p\text{”}$$

Remark. Then $\int e^{t\bar{L}} f \, d\mu = \int f \, d\mu \quad \forall t > 0$ “ μ invariant”

F – L^p -Approach for
Stochastic Porous Medium Equation

Now **Step 1** for stochastic porous medium equation (=SPME):

For simplicity $\Psi(x) = x^3$. So,

$$dX_t = \Delta(X_t^3) dt + \sqrt{A} dW_t \quad (\text{SPME})$$

$$\begin{aligned} & \uparrow \\ & = (W_t^i e_i)_{i \in \mathbb{N}} \\ & \text{where } W_t^i \text{ indep. B. motions on } \mathbb{R}^1 \end{aligned}$$

on $E := H^{-1}(\Lambda)$ ($:=$ dual of $H_0^1(\Lambda)$), $\Lambda \subset \mathbb{R}^d$, open, bdd., $\partial\Lambda$ smooth.

\uparrow
Dirichlet bd. cond.

Have

$$H_0^1(\Lambda) \subset L^2(\Lambda) \subset H^{-1}(\Lambda) \xrightarrow[\text{bijection}]{\Delta^{-1}} H_0^1(\Lambda).$$

- $\{e_i \mid i \in \mathbb{N}\}$ = eigenbasis of Dirichlet Laplacian on $H^{-1}(\Lambda)$.
- $A \in L(H^{-1}, H^{-1})$, $Ae_i = \lambda_i e_i$ (“diagonal”)
- $\lambda_i \geq 0 \forall i \in \mathbb{N}$, and $\sum_{i=1}^{\infty} \lambda_i < \infty$ (“trace class”).

In this case for $\varphi \in \mathcal{FC}_b^2(H^{-1})$

$$L\varphi(x) = \frac{1}{2} \sum_{i=1}^{\infty} \lambda_i \frac{\partial^2}{\partial e_i^2} \varphi(x) + {}_{H^{-1}} \langle \Delta x^3, D\varphi(x) \rangle_{H_0^1},$$

$x \in L^2(\Lambda)$ ($\subset H^{-1}(\Lambda)$) s.th. $x^3 \in H_0^1$.

(So, can only be written in this form for *special* $x \in H^{-1}(\Lambda)$)

Step 1: Solve $L^* \mu = 0$.

Let $V_2 : H^{-1}(\Lambda) \rightarrow [0, \infty]$ “Lyapunov function”

$$V_2(x) := \begin{cases} \frac{1}{2} \int_{\Lambda} x^2(\xi) \, d\xi, & x \in L^2(\Lambda), \\ +\infty, & \text{else.} \end{cases}$$

Then for $x \in L^2(\Lambda)$ ($\subset H^{-1}(\Lambda)$) s.th. $x^2, x^3 \in H_0^1(\Lambda)$

$$LV_2(x) = \underbrace{\frac{1}{2} \sum_{i=1}^{\infty} \lambda_i \int_{\Lambda} e_i^2(\xi) \, d\xi}_{=: C = \text{const.}} + \underbrace{H^{-1} \langle \Delta x^3, x \rangle_{H_0^1}}_{= -\frac{3}{4} \int_{\Lambda} |\nabla x^2(\xi)|^2 \, d\xi} = \underbrace{C}_{\geq 0} - \underbrace{\Theta_2(x)}_{\geq 0}$$

$=: -\Theta_2(x)$

Restrict to $x \in \text{span}\{e_1, \dots, e_N\}$

$$L_N V_2(x) = \underbrace{\frac{1}{2} \sum_{i=1}^N \lambda_i \int_{\Lambda} e_i^2(\xi) \, d\xi}_{\leq C} + \underbrace{H^{-1} \langle P_N(\Delta x^3), x \rangle_{H_0^1}}_{= -\Theta_2(x)} \leq \underbrace{C - \Theta_2(x)}_{\text{independent of } N!}$$

↑
operator on $\mathbb{R}^N!$

Relatively easy to show:

([Bogachev / R.: Th. Prob. Appl. '00])

\exists prob. measure μ_N on $\text{span}\{e_1, \dots, e_N\} \cong \mathbb{R}^N$ s.th. $L_N^* \mu_N = 0$, so

$$0 = \int L_N V_2 \, d\mu_N \leq C - \int \Theta_2 \, d\mu_N \quad \text{❄}$$

$$0 = \int L_N V_2 \, d\mu_N \leq C - \int \Theta_2 \, d\mu_N \quad \text{❄}$$

Consider μ_N on $H^{-1}(\Lambda)$ ($\supset \text{span}\{e_1, \dots, e_N\}$), then

$$\sup_N \mu_N(\underbrace{\{\Theta_2 > R\}}_{\substack{\text{have compact} \\ \text{complements} \\ \text{in } H^{-1}(\Lambda)}}) \stackrel{\text{Chebychev}}{\leq} \frac{1}{R} \cdot \sup_N \int \Theta_2 \, d\mu_N \stackrel{\text{❄}}{\leq} \frac{1}{R} \cdot C \xrightarrow{R \rightarrow \infty} 0$$

$\stackrel{\text{Prokhorov}}{\Rightarrow} \exists \mu := \lim_{k \rightarrow \infty} \mu_{N_k}$ in weak topology of measures on $H^{-1}(\Lambda)$

$$\text{and } \int \Theta_2 \, d\mu \leq C.$$

(Can show similarly: $\int |\nabla x^3|_{L^2}^2 \mu(dx) < \infty$,
 so $\mu(\{x \in L^2(\Lambda) \mid x^2, x^3 \in H_0^1(\Lambda)\}) = 1.$)

Then show (again work!)

$$L^* \mu \quad \left(\stackrel{!}{=} \lim_{k \rightarrow \infty} L_{N_k}^* \mu_{N_k} \right) = 0.$$

G – WFS -Approach

General idea of WFS-approach for

Step 1

Prove a **weighted maximum principle** in infinite dimension, i.e.

show (in applications by finite dimensional approximation):

There exist two functions $\mathbb{V}, \mathbb{W} : E \rightarrow \mathbb{R}_+, \mathbb{V} \leq \mathbb{W}$ both with weakly **compact levels** sets $\{\mathbb{V} \leq R\}, \{\mathbb{W} \leq R\}, R > 0$, such that for some $\lambda_0 > 0$

$$\sup_{x \in \{\mathbb{W} < \infty\}} \frac{(\lambda_0 - L)u}{\mathbb{W}}(x) \geq \sup_{x \in \{\mathbb{V} < \infty\}} \frac{u}{\mathbb{V}}(x),$$

Then

(a variant of) $(L, \mathcal{F}C_0^2)$ is dissipative on $W(E)$,

where the Banach space $W(E)$ is defined by

$$W(E) := \left\{ u : \{\mathbb{V} < \infty\} \rightarrow \mathbb{R} \mid f|_{\{\mathbb{V} \leq R\}} \text{ is weakly continuous } \forall R > 0 \text{ and } \lim_{R \rightarrow \infty} \sup_{\{\mathbb{V} \geq R\}} \frac{|f|}{\mathbb{V}} = 0 \right\} =: C_{\mathbb{V}}$$

equipped with the norm

$$\|u\|_{W(E)} := \sup_{\{\mathbb{V} < \infty\}} \frac{|u|}{\mathbb{V}}.$$

H – WFS -Approach for Stochastic Navier-Stokes Equation

Step 1

Weighted maximum principle holds with $(\kappa, \alpha > 0, \kappa > \alpha(1 + \nu^{-2}))$

$$\mathbb{V}(x) := e^{\kappa\|x\|_E^2} (1 + \|\nabla x\|_E^2)^\alpha$$

and

$$\mathbb{W}(x) := \nu \mathbb{V}(x) (\kappa \|\nabla x\|_2^2 + \alpha \|\Delta x\|_2),$$

$$x \in E = \{x \in L^2(\Lambda \rightarrow \mathbb{R}^d, d\xi) \mid \operatorname{div} x = 0\}.$$