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Lecture 1

Passive Scalar in Turbulent Velocity Field

## The Transport Equation

$$\begin{aligned}\frac{\partial\theta(t,x)}{\partial t} &= -\mathbf{v}(t,x) \cdot \nabla\theta(t,x), \quad t > 0, \quad x \in \mathbb{R}^d; \\ \theta(0,x) &= \theta_0(x);\end{aligned}$$

$$\mathbf{v} = (v^1, \dots, v^d) \in \mathbb{R}^d, \quad d \geq 2.$$

If each  $v^i$  is Lipschitz continuous in  $x$ , then

$$\theta(t,x) = \theta_0(X_{t,0}^x);$$

$X$  is the flow of  $\mathbf{v}$ :

$$\frac{dX_{s,t}^x}{dt} = \mathbf{v}(t, X_{s,t}^x), \quad t > s, \quad X_{s,s}^x = x.$$

## Turbulent Transport

What if  $\mathbf{v}$  is not Lipschitz continuous in  $x$ ?

- Example — Kolmogorov's theory:  $\mathbf{v}$  is Hölder  $\approx 1/3$ .
- Difficulty — Existence but no uniqueness for the flow equation.

$$\frac{dX_{s,t}^x}{dt} = \mathbf{v}(t, X_{s,t}^x), \quad t > s, \quad X_{s,s}^x = x.$$

- How to find  $\theta$ ?

$$\theta(t, x) = \theta_0(X_{t,0}^x) \text{ is not true}$$

## Regularization

- Introducing viscosity ( $\kappa$ -limit):

$$\frac{\partial \theta^\kappa(t, x)}{\partial t} = \kappa \Delta \theta^\kappa(t, x) - \mathbf{v}(t, x) \cdot \nabla \theta^\kappa(t, x), \quad t > 0, \quad x \in \mathbb{R}^d;$$

$$dX_{s,t}^{\kappa,x} = \mathbf{v}(t, X_{s,t}^{\kappa,x}) dt + \sqrt{2\kappa} dw(t), \quad t > s, \quad X_{s,s}^{\kappa,x} = x.$$

- Smoothing out  $\mathbf{v}$  ( $\varepsilon$ -limit):

$$\mathbf{v}^\varepsilon(t, x) = \frac{1}{\varepsilon^d} \int_{\mathbb{R}^d} \mathbf{v}(t, y) \psi\left(\frac{x-y}{\varepsilon}\right) dy$$

(Gawędzki and Vergassola (2000), E and Vanden Eijnden (2000))

## Kraichnan's Model of Turbulence

### Physical Model for $v$ :

- $\mathbf{v}$  is a statistically homogeneous, isotropic, and stationary Gaussian vector field with zero mean and covariance

$$E(v^i(t, x)v^j(s, y)) = \delta(t - s)C^{ij}(x - y).$$

- For small  $x$ ,  $C^{ij}(x) \sim C^{ij}(0)(1 - |x|^\gamma)$ ,  
 $0 < \gamma < 2$ .

### Mathematical Model for $v$ (Le Jan and Raimond (2002), Baxendale, Harris (1988)):

- The matrix  $C$  is characterized by its Fourier transform:

$$\widehat{\mathbf{C}}(z) = \frac{A_0}{(1 + |z|^2)^{(d+\gamma)/2}} \left( a \frac{zz^*}{|z|^2} + \frac{b}{d-1} \left( I - \frac{zz^*}{|z|^2} \right) \right),$$

- $a = 0 \Rightarrow \nabla \cdot \mathbf{v} = 0$ ;
- $b = 0 \Rightarrow \mathbf{v} = \nabla V$  for some scalar  $V$ .
- $\zeta = b/(a + b)$ — degree of incompressibility.

- Representation of  $\mathbf{v}$ :

$$v^i(t, x) = \sum_{k \geq 1} \sigma_k^i(x) \dot{w}_k(t),$$

$\dot{w}_k(t)$  are independent standard Gaussian white noises;  $\{\sigma_k, k \geq 1\}$  is a CONS in  $H_C$ .

- $H_C = H^{(d+\gamma)/2}(\mathbb{R}^d; \mathbb{R}^d)$ ,  $a, b > 0$ ;
- $\sigma_k^i$  is Hölder  $\gamma/2$ .

$$C^{ij}(x - y) = \sum_k \sigma_k^i(x) \sigma_k^j(y),$$

**Thm** (Le Jan, Raimond, 2002) For a suitable class of initial conditions  $\theta_0, \theta(t, x)$

$$\theta(t, x) = \int \theta_0(y) P(X_{0,t}^x \in dy | \mathcal{F}_t^W) \quad (1)$$

where

$$X_{t,x}(s) = x + \int_s^t \sigma^k(X_{t,x}(r)) \circ \overleftarrow{dw}_k(r).$$

Le Jan and Raimond have also derived an equation for the measure  $P(X_{0,t}^x \in dy | \mathcal{F}_t^W)$ , similar to the Zakai equation of nonlinear filtering.

- *Statistical ("weak" in probabilistic sense) solution of the flow equation.*
- *Still very little info about  $\theta$  (Transport equation is solved in the space of measures, no uniqueness was established)*

## Transport Equation as an SPDE

- $\mathbb{F} = (\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{t \geq 0}, \mathbb{P})$ , a stochastic basis with the usual assumptions.
- $(w_k(t), k \geq 1, t \geq 0)$ , independent standard Wiener processes on  $\mathbb{F}$ .
- $\mathbf{v}$  divergence-free (incompressible flow  $\implies \operatorname{div} \sigma_k = 0$ ).

Since the Kraichnan velocity  $\mathbf{V}$ :

$$V^i(t, x) = \sum_{k \geq 1} \sigma_k^i(x) \dot{w}_k(t),$$

the transport equation is given by

$$d\theta(t, x) = - \sum_k \sigma_k(x) \cdot \nabla \theta(t, x) \circ dw_k(t).$$

or

$$\boxed{d\theta(t, x) = \frac{1}{2} C^{ij}(0) D_i D_j \theta(t, x) dt - \sigma_k^i(x) D_i \theta(t, x) dw_k(t)}$$

**Notation:**  $D_i = \frac{\partial}{\partial x^i}$ .

**Summation convention:** summation over a pair of repeating indices.

Stochastic PDEs with Multiplicative Noise;  
Basic facts

Consider a stochastic evolution equation

$$du(t, x) = \mathcal{A}(t, x) u(t, x) dt + \sum_{k=1}^{\infty} \mathcal{M}^k(t, x) u(t, x) dw_k(t), u(0, x) = u_0(x)$$

where  $\mathcal{A}$  and  $\mathcal{M}$  are differential operators, and  $w_k$  are independent standard Wiener processes .

$$(H) : \mathcal{A} - \frac{1}{2} \mathcal{M} \mathcal{M}^* \text{ is elliptic}$$

If (H) does not hold, one could not guarantee that a solution of this equation is square integrable, i.e.  $E \|u(t, \cdot)\|_{\mathbb{R}^d}^2 < \infty$  for all  $t$ ;

**Examples:**

1.  $\mathcal{A} = \frac{1}{2} \Delta$ ,  $\mathcal{M} = \varepsilon \nabla$ ,  $\varepsilon < 1$  – elliptic;
2.  $\mathcal{A} = \frac{1}{2} \Delta$ ,  $\mathcal{M} = \nabla$  – degenerate elliptic;
3.  $\mathcal{A} = \frac{1}{2} \Delta$ ,  $\mathcal{M} = \varepsilon \nabla$ ,  $\varepsilon > 1$  – non-elliptic;

The transport equation

$$d\theta(t, x) = \frac{1}{2} C^{ij}(0) D_i D_j \theta(t, x) dt - \sigma_k^i(x) D_i \theta(t, x) dw_k(t)$$

is degenerate elliptic!

(Krylov-R., P. Chow, J. Potthoff, B. Øksendal, etc.  $\sigma$ -smooth)



## A Wiener Chaos Approach to Solving the Stochastic Transport Equation

### Wiener chaos:

$$W(t) = (w_k(t), k \geq 1, 0 < t < T),$$

$$\{m_i(s), i \geq 1\} \text{ — CONS in } L_2([0, T]),$$

$$\xi_i^k = \int_0^T m_i(s) dw_k(s).$$

$$\mathcal{J} = \left( \alpha = (\alpha_i^k, i, k \geq 1) \mid |\alpha| = \sum_{i,k} \alpha_i^k < \infty \right);$$

$$\xi_\alpha = \prod_{i,k} \left( \frac{H_{\alpha_i^k}(\xi_{ik})}{\sqrt{\alpha_i^k!}} \right), \text{ where}$$

$$H_n(x) = (-1)^n \exp\left\{\frac{x^2}{2}\right\} \frac{d^n}{dx^n} \exp\left\{-\frac{x^2}{2}\right\}.$$

**Theorem.** (Cameron and Martin, 1947)

The collection  $\{\xi_\alpha, \alpha \in \mathcal{J}\}$  is an orthonormal basis in  $L_2(\Omega, \mathcal{F}_T^W, \mathbb{P})$ :

If  $\eta \in L_2(\Omega, \mathcal{F}_T^W, \mathbb{P})$  and  $\eta_\alpha = \mathbb{E}(\eta \xi_\alpha)$ , then

$$\eta = \sum_{\alpha \in \mathcal{J}} \eta_\alpha \xi_\alpha$$

and

$$\mathbb{E}|\eta|^2 = \sum_{\alpha \in \mathcal{J}} \eta_\alpha^2.$$

## The Propagator System

$$d\theta(t, x) = \frac{1}{2}C^{ij}(0)D_iD_j\theta(t, x)dt - \sigma_k^i(x)D_i\theta(t, x)dw_k(t)$$

If  $\sigma_k^i, \theta_0$  are smooth, then the STE has a nice square integrable solution, moreover  $\theta(t, x) = \sum_{\alpha \in \mathcal{J}} \theta_\alpha(t, x)\xi_\alpha(t)$ .

**Define:**  $\xi_\alpha(t) = \mathbb{E}(\xi_\alpha | \mathcal{F}_t^W)$ ;  $\xi_\alpha(0) = I(|\alpha| = 0)$ .

**Fact:**

$$d\xi_\alpha(t) = \mathcal{D}\xi_\alpha(t)dW(t),$$

where

$\mathcal{D}\xi_\alpha(t) = m_i(t) \sqrt{\alpha_i^k} \xi_{\alpha^-(i,k)}(t) \ell_k$  is the Malliavin derivative, and  $\alpha^-(i, k)$  is the multi-index with the components

$$(\alpha^-(i, k))_j^l = \begin{cases} \max(\alpha_i^k - 1, 0), & \text{if } i = j \text{ and } k = l, \\ \alpha_j^l, & \text{otherwise.} \end{cases}$$

By the Itô formula

$$\begin{aligned} \frac{\partial \theta_\alpha(t, x)}{\partial t} &= \frac{1}{2}C^{ij}(0)D_iD_j\theta_\alpha(t, x) \\ &\quad - \sum_{i,k} \sqrt{\alpha_i^k} \sigma_k^j(x) D_j \theta_{\alpha^-(i,k)}(t, x) m_i(t); \\ \theta_\alpha(0, x) &= \theta_0(x) I(|\alpha| = 0) \end{aligned}$$

## Solving the Propagator

**Good news:**  $\sigma_k^i$  do not have to be smooth or even continuous.  
 $|\alpha| = 0$ :

$$\frac{\partial \theta_{(0)}(t, x)}{\partial t} = \frac{1}{2} C^{ij}(0) D_i D_j \theta_{(0)}(t, x)$$

$$\theta_{(0)}(0, x) = \theta_0(x) \Rightarrow \theta_{(0)}(t, x) = \mathbb{T}_t \theta_0(x).$$

$\alpha = \delta_{ik}$ :

$$\begin{aligned} \frac{\partial \theta_{(ik)}(t, x)}{\partial t} &= \frac{1}{2} C^{ij}(0) D_i D_j \theta_{(ik)}(t, x) \\ &\quad - \sigma_k^j(x) D_j \theta_{(0)}(t, x) m_i(t); \quad \theta_{(ik)}(0, x) = 0. \end{aligned}$$

$$\theta_{ik}(t, x) = - \int_0^t m_i(s) \mathbb{T}_{t-s} \sigma_k^j D_j \mathbb{T}_s \theta_0(x) ds.$$

In fact, with  $\mathcal{M}_k = -\sigma_k^j D_j$ ,

$$\sum_{|\alpha|=N} |\theta_\alpha(t, x)|^2 = \sum_{k_1, \dots, k_N=1}^{\infty} \int_0^t \int_0^{s_N} \dots \int_0^{s_2} |\mathbb{T}_{t-s_N} \mathcal{M}_{k_N} \dots \mathbb{T}_{s_2-s_1} \mathcal{M}_{k_1} \mathbb{T}_{s_1} \theta_0(x)|^2 ds_1 \dots ds_N.$$

$\exists!$  in  $L_2$

$$d\theta(t, x) = \frac{1}{2} C^{ij}(0) D_i D_j \theta(t, x) dt - \sigma_k^i(x) D_i \theta(t, x) dw_k(t)$$

**Thm** (Lototsky and R., Russian Math. Surveys, 2003)

If  $\theta_0 \in L_2(\mathbb{R}^d)$ , then:

- For every  $\varphi \in C_0^\infty(\mathbb{R}^d)$ , the random field  $\theta(t, x) = \sum_{\alpha \in \mathcal{J}} \theta_\alpha(t, x) \xi_\alpha$  is a unique strong solution of the transport equation in that for any test-function  $\varphi$ ,

$$\begin{aligned} (\theta, \varphi)(t) &= (\theta_0, \varphi) + \frac{1}{2} \int_0^t C^{ij}(0) (\theta, D_i D_j \varphi)(s) ds \\ &\quad + \int_0^t (\theta, \sigma_k^i D_i \varphi) dw_k(s) \end{aligned}$$

- For  $t > 0$ ,

$$\|\theta(t)\|_{L_2(\mathbb{R}^d)}^2 = \sum_{\alpha \in \mathcal{J}} \|\theta_\alpha(t)\|_{L_2(\mathbb{R}^d)}^2 < \|\theta_0\|_{L_2(\mathbb{R}^d)}^2.$$

- For

$$\begin{aligned} d_s X_t^{x,i}(s) &= -\sigma_k^i(X_t^x(s)) \overleftarrow{dw}_k(s), s \in [0, t), \\ X_t^x(t) &= x \end{aligned}$$

(martingale solution), and

$$\theta(t, x) = \mathbb{E}(\theta_0(X_t^x(0)) | \mathcal{F}_t^W)$$

## Remarks

- $\mathbb{E}\theta(t, x) = \theta_\emptyset(t, x)$ . ( $\emptyset$  is a multi-index with zero entries)
  - $\mathbb{E}(\theta(t, x)\theta(s, y)) = \sum_{\alpha \in \mathcal{J}} \theta_\alpha(t, x)\theta_\alpha(s, y)$ .
  - By interpolation:  $\mathbb{E}\|\theta\|_{L_p(\mathbb{R}^d)}^p < \|\theta_0\|_{L_p(\mathbb{R}^d)}^p$ ,  $2 < p < \infty$ . Weighted  $L_p$  (e.g.  $\theta_0(x) = |x|$ ) are also OK.
  - Conservation of energy,  $\mathbb{E}\|\theta(t)\|_{L_2(\mathbb{R}^d)}^2 = \|\theta_0\|_{L_2(\mathbb{R}^d)}^2$ ,
- $\Updownarrow$
- Pathwise solution of the flow equation.

## Totally Turbulent Transport

$$d\theta(t, x) = \nu \Delta \theta(t, x) dt - \sigma_k^i(x) D_i \theta(t, x) dw_k(t),$$

$\sigma_k$ ,  $k \geq 1$  — CONS in  $L_2(\mathbb{R}^d; \mathbb{R}^d) \Leftrightarrow \dot{W}$  is space-time white noise

**Note:**  $\sum_{k \geq 1} \sigma_k^i(x) \sigma_k^j(x)$  diverges.

S-system:

$$\begin{aligned} \frac{\partial \theta_\alpha(t, x)}{\partial t} &= \nu \Delta \theta_\alpha(t, x) \\ &\quad - \sum_{i, k} \sqrt{\alpha_i^k} \sigma_k^j(x) D_j \theta_{\alpha^-(i, k)}(t, x) m_i(t); \end{aligned}$$

Still solvable, but now

$$\sum_{\alpha \in \mathcal{J}} \|\theta_\alpha\|_{L_2(\mathbb{R}^d)}^2 = \infty$$

## Weighted Wiener Chaos

Let  $Q := \{q_1, q_2, \dots\}$ ,  $q_k > 0$ , and  $q^\alpha := \prod_{k \in \mathcal{J}} q_k^{\alpha_k}$ .

**Definition.** The  $Q$ -weighted Wiener Chaos space  $L_{2,Q}(\mathcal{F}_T^W; L_2(\mathbb{R}^d))$  is

$$L_{2,Q}(\mathcal{F}_T^W; L_2(\mathbb{R}^d)) = \left\{ (u_\alpha) : \sum_{\alpha \in \mathcal{J}} q^{-2\alpha} \|u_\alpha\|_{L_2(\mathbb{R}^d)}^2 < \infty \right\}.$$

Still write  $u = \sum_{\alpha \in \mathcal{J}} u_\alpha \xi_\alpha$

Where this series converges?

Examples:

1. (Obvious) If  $u(t) = 1 + \sum_{k \geq 1} \int_0^t u(s) dw_k(s)$ , then  $u \in L_{2,Q}(\mathcal{F}_T^W; \mathbb{R})$  for every  $Q = (q_1, q_2, \dots)$  so that  $\sum_{k \geq 1} q_k^2 < \infty$ .
2. (Nualart-R., JFA, 1997) If

$$du(t, x) = \Delta u(t, x) dt + u(t, x) dw(t, x), \quad t > 0, x \in \mathbb{R}^d, \quad d \geq 2,$$

then  $u \in L_{2,Q}(\mathcal{F}_T^W; L_2(\mathbb{R}^d))$  for some  $Q$ .

**Theorem** (Lototsky-R., Annals of Prob. 2005)

Assume that  $\theta_0 \in L_2(\mathbb{R}^d)$  and  $|\sigma_k^i(x)| \leq C_k$ . Let  $Q$  be a sequence with  $q_k = \frac{\sqrt{\delta\nu}}{d2^k C_k}$  for some  $0 < \delta < 2$ . If

$$\begin{aligned} \frac{\partial \theta_\alpha(t, x)}{\partial t} &= \nu \Delta \theta_\alpha(t, x) \\ &\quad - \sum_{i,k} \sqrt{\alpha_i^k} \sigma_k^j(x) D_j \theta_{\alpha^-(i,k)}(t, x) m_i(t); \end{aligned}$$

then  $\sum_{\alpha \in \mathcal{J}} q^{2\alpha} \|\theta_\alpha(t)\|_{L_2(\mathbb{R}^d)}^2 < \infty$   
and  $\theta(t, x) = \sum_{\alpha \in \mathcal{J}} \theta_\alpha(t, x) \xi_\alpha$  satisfies

$$\theta \in L_{2,Q}(\mathcal{F}_T^W; \mathbf{C}((0, T); L_2(\mathbb{R}^d))).$$

This  $\theta$  is called the *Wiener Chaos solution* of the totally turbulent transport equation  $d\theta(t, x) = \nu \Delta \theta(t, x) dt - \sigma_k^i(x) D_i \theta(t, x) dw_k(t)$ .



## Wiener Chaos Approach

- Computable expressions for the solution and its moments from the S-system.
- New regularity results.
- Possibilities for generalization.

### Further Directions

- Anticipating equations.
- Elliptic equations.
- Nonlinear equations.

### **The main results can be found in:**

1. S. Lototsky, B. L. Rozovskii. Passive Scalar Equation in a Turbulent Incompressible Gaussian Velocity Field. *Russian Math. Surv.* 59 (2004), No.2,
2. S. Lototsky, B. L. Rozovskii. Wiener chaos solutions of linear stochastic evolution equations. *Ann. Probab.* (2006, to appear).