

# Stability Results for Advection-Diffusion Equations with Deterministic and Random Vector Fields

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# Introduction: Advection-Diffusion Equations

# Advection-Diffusion Equations

Let  $D \subseteq \mathbb{R}^d$ , and consider the continuity equation for a passive scalar  $\theta(t, x)$  under the action of a vector field  $u(t, x)$  with  $\kappa \geq 0$ ,

$$\begin{cases} \partial_t \theta + \nabla \cdot (u\theta) = \kappa \Delta \theta & \text{in } (0, T) \times D, \\ \theta(0, \cdot) = \theta^0 & \text{in } D. \end{cases} \quad (\text{AD})$$

If  $D$  has a boundary:  $(u - \kappa \nabla \theta) \cdot n = 0$  on  $(0, T) \times \partial D$ .



Figure: Action of an alternating shear flow

# Advection-Diffusion Equations

Different viewpoint: follow the particle trajectories, given by the flow map

$$\begin{cases} dX_t &= u(t, X_t)dt + \sqrt{2\kappa}dB_t - n(X_t)dL_t, \\ X_0 &= \text{id}, \end{cases} \quad (\text{SDE})$$

where  $(B_t)_{t \geq 0}$  is a Brownian motion in  $\mathbb{R}^d$ , and  $(L_t)_{t \geq 0}$  is a local time of the process  $X_t$  that only activates when  $X_t$  touches the boundary  $\partial D$ .

- Solutions to (AD) and (SDE) are related through **Feynman-Kac**:

$$\theta(t, \cdot) = \mathbb{E}[(X_t)_{\#} \theta^0].$$

**DiPerna-Lions setting:**  $u \in L^1(W^{1,p})$  with  $p > 1$ , and  $(\nabla \cdot u)^- \in L^1(L^\infty)$ .

## Theorem [DiPerna, Lions (1989)]

For  $\kappa = 0$ , let  $\theta^0 \in L^q$  with  $1/p + 1/q \geq 1$ . Then there exists a unique distributional solution  $\theta \in L^\infty(L^q)$ .

# Control over the Gradient of $\theta$

$\kappa > 0$  yields a control over the gradient of  $\theta$ :

## Theorem [Le Bris, Lions (2008)]

If  $\kappa > 0$ , let  $p = 2$  and  $\theta^0 \in L^2 \cap L^\infty$ . Then there exists a unique distributional solution  $\theta \in L^\infty(L^2 \cap L^\infty) \cap L^2(\dot{H}^1)$ .

For our estimates with  $\kappa > 0$ , we want  $\nabla\theta$  to be controlled in  $L^1(L^1)$ . Consider initial data with **finite entropy**,

$$\int_D \theta^0 \log \theta^0 dx < \infty \quad \Rightarrow \quad \iint_{(0,T) \times D} |\nabla\theta| dx dt \lesssim \sqrt{\frac{T}{\kappa}}.$$

How to achieve finite entropy?

- **Bounded domain:**  $\theta^0 \in L^q$ ,  $q > 1$ .
- **Unbounded domain:**  $\theta^0 \in L^1 \cap L^q$ ,  $q > 1$  and finite first moments.

# Optimal Transport Distances

Let  $\mu, \nu \in L^1$  measures of equal mass,  $\Pi(\mu, \nu)$  the set of all transport plans between them, and  $c : [0, \infty) \rightarrow [0, \infty)$  a nondecreasing **cost function**. The optimal transport distance is defined via the minimisation problem

$$\mathcal{D}_c(\mu, \nu) = \inf_{\pi \in \Pi(\mu, \nu)} \iint_{D \times D} c(|x - y|) d\pi(x, y).$$

- $c(z)$  is a distance:  $\mathcal{D}_c$  metrizes **weak convergence** of measures.
- $c(z)$  is concave: the OT problem admits a dual formulation,

$$\mathcal{D}_c(\mu, \nu) = \sup_{|\xi(x) - \xi(y)| \leq c(|x - y|)} \int_D \xi(z) d(\mu - \nu)(z).$$

We consider a logarithmic cost function,

$$c(z) = \log \left( \frac{z}{\delta} + 1 \right) \quad \text{with} \quad \delta > 0.$$

# Stability for Distributional Solutions



# Stability in the DiPerna-Lions Setting

In the DiPerna-Lions setting, we find bounds on the the distance between two distributional solutions to (AD) given by different data.

- $u \in L^1(W^{1,p})$ ,  $(\nabla \cdot u)^- \in L^1(L^\infty)$ ,
- $\theta^0 \in L^1 \cap L^q$  and finite first moments.

## Theorem 1 [NF, Schlichting, Seis]

Let  $\theta_1, \theta_2 \in L^\infty(L^q) \cap L^1(W^{1,1})$  be the unique solutions to (AD) defined by  $(u_1, \kappa_1, \theta_1^0)$  and  $(u_2, \kappa_2, \theta_2^0)$  respectively. Then we find the following stability estimate,

$$\sup_{t \in (0, T)} \mathcal{D}_\delta(\theta_1, \theta_2)(t) \lesssim 1 + \mathcal{D}_\delta(\theta_1^0, \theta_2^0) + \frac{\|u_1 - u_2\|_{L^1(L^p)}}{\delta} + \frac{|\kappa_1 - \kappa_2|}{\delta},$$

for every  $\delta > 0$ .

# Optimality of the Estimate and Zero-Diffusivity Limit

**Rate of convergence:** smallest  $\delta = \delta_n$  for which the RHS is finite,

- 1 Initial data:  $\mathcal{D}_{\delta_n}(\theta^0, \theta_n^0) \sim 1$ . Optimal for weak convergence.
- 2 Vector field:  $\|u - u_n\|_{L^1(L^p)} \sim \delta_n$ . Optimal if  $\kappa = 0$ .
- 3 Diffusivity constant:  $|\kappa - \kappa_n| \sim \delta_n$ . Best known rate. Optimal?

$$\frac{t|\kappa_1 - \kappa_2|}{\sqrt{\kappa_1} + \sqrt{\kappa_2}} \lesssim W_1(\theta_1, \theta_2)(t), \quad \theta_1, \theta_2 \text{ heat kernels.}$$

**The zero-diffusivity limit:** let  $u_1 = u_2$  and  $\theta_1^0 = \theta_2^0$ ,

$$\sup_{t \in (0, T)} \mathcal{D}_\delta(\theta_1, \theta_2)(t) \lesssim 1 + \frac{|\kappa_1 - \kappa_2| \|\nabla \theta_2\|_{L^1(L^1)}}{\delta} \lesssim 1 + \frac{|\kappa_1 - \kappa_2|}{\delta} \sqrt{\frac{T}{\kappa_2}}.$$

- Seis (2018).

In the limit  $\kappa \rightarrow 0$ , the optimal rate is  $\mathcal{D}_\delta(\theta, \theta^\kappa)(t) \lesssim 1 + \sqrt{t\kappa}/\delta$ .

## Stability out of the DiPerna-Lions Setting

We prove well-posedness of the Cauchy problem (AD) out of the DiPerna-Lions setting, see [Bouchut, Crippa \(2013\)](#).

- $\nabla u = K * \omega$  where  $\omega \in L^1(L^1)$  and  $K$  is a singular integral kernel,
- $\theta^0 \in L^1 \cap L^\infty$  mean free.

### Theorem 2 [NF, Schlichting, Seis]

The Cauchy problem (AD) has a unique distributional solution with

$$\theta \in L^\infty(L^1 \cap L^\infty) \quad \text{and} \quad \nabla \theta \in L^1(L^1).$$

Uniqueness is a byproduct of the estimate:  $\forall \varepsilon > 0, \exists C_\varepsilon > 0$  such that

$$\sup_{t \in (0, T)} \mathcal{D}_\delta(\theta, 0)(t) \lesssim \mathcal{D}_\delta(\theta^0, 0) + \varepsilon \left( 1 + \log \frac{\|u\|_{L^{p, \infty}}}{\varepsilon \delta} \right) + C_\varepsilon,$$

for all  $\delta > 0$ .

# Stability for the Implicit Finite Volume Scheme

# Unstructured Meshes

Let  $D \subset \mathbb{R}^d$  be bounded, let  $\partial D$  be  $C^{1,1}$ , and consider

- $\{K\}_{K \in \mathcal{T}} \subset D$  a tessellation with closed, polygonal cells;
- $h = \max \text{diam} K$  size of the mesh.

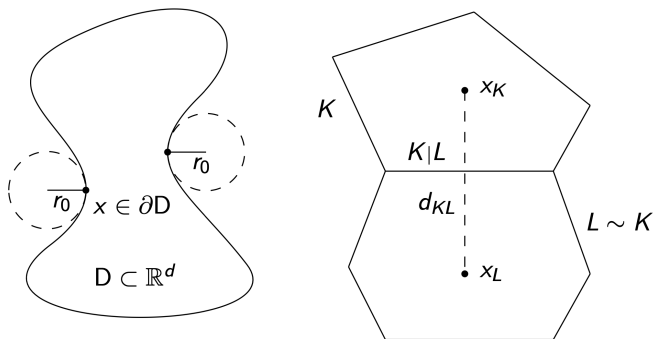


Figure: Exterior ball condition and an example of control cell.

# The Implicit Finite Volume Scheme

Let  $\tau > 0$  be the time step.

- Initial datum averaged on every cell  $\theta_K^0 = \int_K \theta^0 dx$ .
- Discrete normal velocity from control cell  $K$  to neighboring  $L$ ,

$$u_{KL}^n = \int_{n\tau}^{(n+1)\tau} \int_{K|L} u \cdot n_{KL} dH^{d-1} dt.$$

Then the finite volume scheme is given by

$$\frac{\theta_K^{n+1} - \theta_K^n}{\tau} + \sum_{L \sim K} \frac{|K|L|}{|K|} \left( u_{KL}^{n+} \theta_K^{n+1} - u_{KL}^{n-} \theta_L^{n+1} + \kappa \frac{\theta_K^{n+1} - \theta_L^{n+1}}{d_{KL}} \right) = 0.$$

The approximate solution  $\theta_{\tau h}$  is defined by

$$\theta_{\tau h}(t, x) = \theta_K^n \quad \text{a.e. } (t, x) \in [n\tau, (n+1)\tau) \times K. \quad (\text{FV})$$

# Stability for the Implicit Finite Volume Scheme

We study the convergence of the approximate solution towards the distributional solution on the DiPerna-Lions setting:

- $u \in L^1(W^{1,p})$  with  $p \in (1, \infty]$ ,  $(\nabla \cdot u)^- \in L^1(L^\infty)$ ;
- $\theta^0 \in L^q$  with  $q \in (1, \infty]$  and  $1/p + 1/q \leq 1$ .

In addition we assume:  $u \in L^\infty((0, T) \times D)$ .

## Theorem 3 [NF, Schlichting]

Let  $\theta \in L^\infty(L^q) \cap L^1(W^{1,1})$  be the unique distributional solutions to (AD) and  $\theta_{\tau h}$  the unique approximate solution given by (FV). Then, for  $\tau > 0$  small enough, there holds

$$\sup_{t \in (0, T)} \mathcal{D}_\delta(\theta, \theta_{\tau h})(t) \lesssim 1 + \frac{h}{\delta} + \frac{\sqrt{\tau T} \|u\|_\infty}{\delta} + \frac{\sqrt{\tau \kappa}}{\delta},$$

for every  $\delta > 0$ .

# Numerical Diffusion and Optimality

$$\sup_{t \in (0, T)} \mathcal{D}_\delta(\theta, \theta_{\tau h})(t) \lesssim 1 + \frac{h}{\delta} \min \left\{ \frac{1}{\sqrt{h}}, \frac{1}{\sqrt{\kappa}} \right\} + \frac{\sqrt{\tau}}{\delta}$$

- Guo, Stynes (1997), Droniou (2002).  
Rate of convergence with  $\kappa > 0$  and smooth vector field:  $h$ .
- Schlichting, Seis (2018).  
Rate of convergence with  $\kappa = 0$  in DiPerna-Lions:  $\sqrt{h}$ .

How do we improve the rate of convergence? **BV estimates:**

$$\tau \sum_n \sum_K \sum_{L \sim K} |K|L \|\theta_K^{n+1} - \theta_L^{n+1}\| \lesssim \min \left\{ \frac{1}{\sqrt{h}}, \frac{1}{\sqrt{\kappa}} \right\}.$$

The discretization of  $D \subset \mathbb{R}^d$  generates numerical diffusion that heuristically corresponds to a second diffusion with coefficient  $h > 0$ ,

$$\partial_t \theta + u \cdot \nabla \theta = (\kappa + h) \Delta \theta.$$



# Ergodicity and Mixing with Random Vector Fields

## Transport by Random Vector Fields

Let  $\kappa = 0$  and consider the transport of the passive scalar  $\theta(t, x)$  by a divergence free vector field  $u(t, x)$ ,

$$\begin{cases} \partial_t \theta + u \cdot \nabla \theta = 0 & \text{in } (0, \infty) \times D, \\ \theta(0, \cdot) = \theta^0 & \text{in } D. \end{cases} \quad (\text{T})$$

If  $D$  has a boundary, we impose  $u \cdot n = 0$  on  $(0, \infty) \times \partial D$ . Then,

- the total mass of  $\theta(t, \cdot)$  is conserved;
- $\|\theta(t)\|_{L^p} = \|\theta^0\|_{L^p}$  for all  $p \in [1, \infty]$ .

**Question:** Can we find examples of vector fields  $u$  that make

$$\theta(t, \cdot) \rightarrow \int_D \theta^0 dx \stackrel{\text{WLOG}}{=} 0$$

as  $t \rightarrow \infty$  in some sense? How fast can the convergence be?

# Random Vector Fields

Let  $D, D' \subset \mathbb{R}^d$  be bounded, we choose vector fields the form

$$u = u(x, y) \quad \text{with } (x, y) \in D \times D'.$$

In  $D'$ , define a Brownian motion  $(Y_t)_{t \geq 0}$  of *intensity*  $\nu > 0$ ,

$$dY_t = \sqrt{2\nu} dB_t - n(Y_t) dL_t, \quad Y_0 = \text{id}.$$

The **random vector field** in  $(T)$  is defined by  $u(x, Y_t)$ , an depends implicitly on the noise realisation and initial point  $Y_0 \in D'$ .

- We look for examples of vector fields  $u(\cdot, Y_t)$  that make the passive scalar  $\theta(t, \cdot)$  being **exponentially ergodic**,

$$\|\mathbb{E}\theta(t)\|_{L^2} \rightarrow 0 \quad \text{exponentially fast.}$$

# Ergodicity and Annealed Mixing

- We obtain a coupled SDE in  $D \times D'$ :

$$\begin{pmatrix} dX_t \\ dY_t \end{pmatrix} = \begin{pmatrix} u(X_t, Y_t) \\ 0 \end{pmatrix} dt + \begin{pmatrix} 0 \\ \sqrt{2\nu} \end{pmatrix} dB_t - \begin{pmatrix} 0 \\ n(Y_t) \end{pmatrix} dL_t$$

To study the long-time behaviour of this system there are two perspectives.

- 1 **ODE:** random dynamics + Markov process

$$P_t(x, A) = \mathbb{P}[X_t \in A \mid X_0 = x], \quad A \in \mathcal{B}(D),$$

see [Bedrossian, Blumenthal, Punshon-Smith \(2018, 2019, 2020\)](#).

- 2 **PDE:** we can use Feynman-Kac to transform the SDE into a PDE

$$\begin{cases} \partial_t f + u(x, y) \cdot \nabla_x f &= \nu \Delta_y f, & \text{in } D \times D', \\ n_y \cdot \nabla_y f &= 0, & \text{on } \in D \times \partial D', \end{cases} \quad (\text{PDE})$$

where  $f(t, x, y) = \mathbb{E}[(X_t, Y_t)_{\#} f^0(x, y)]$ .

# Hypoocoercivity and Ergodicity

Let  $u \in C^2(D \times D')$ , and  $f(0, x, y) = \theta^0(x)\rho(y)$  with

- law  $Y_0 = \rho \in (\mathcal{P} \cap W^{1,\infty})(D')$ ;
- $\theta^0$  mean free, and  $\theta^0 \in H^1(D)$ .

## Theorem 4 [NF, Schlichting, Seis]

If  $\exists \gamma > 0$  such that  $\|f(t)\|_{H^1} \lesssim e^{-\gamma t}$ , then  $\|\mathbb{E}\theta(t)\|_{L^2} \lesssim e^{-\gamma t}$ .

Namely, we found a *sufficient condition* for exponential ergodicity:

**hypoocoercivity** of  $\mathcal{L} = u(x, y) \cdot \nabla_x - \nu \Delta_y$  (+ BC), see [Villani \(2009\)](#).

- **Example 1:** Shear flows with random phases in  $\mathbb{T}^2$ ,

$$u_{\text{shear}}(x, y) = \begin{pmatrix} \sin(x_2 + y_1) \\ \sin(x_1 + y_2) \end{pmatrix}, \quad (x, y) \in \mathbb{T}^2 \times \mathbb{T}^2.$$

## Examples of Ergodicity with a Random Vector Field

- **Example 2:** Randomly moving vortex in  $B_1 \subset \mathbb{R}^2$ ,

$$u_{\text{vortex}}(x, y) = -\frac{e^{-2\pi\psi(x,y)}}{1 - |y|^2} \nabla_x^\perp \psi(x, y), \quad (x, y) \in B_1 \times B_r,$$

where  $r < 1$  and  $\psi(x, y)$  is the streamfunction of a point vortex in  $y$ .

### Theorem 5 [NF, Schlichting, Seis]

Let  $\nu \gg 1$  be sufficiently large, and  $f$  be a solution to (PDE).

- ① Given the vector field  $u_{\text{shear}}(\cdot, Y_t)$ ,  $\exists \alpha > 0$  such that

$$\|f(t)\|_{H^1(\mathbb{T}^2 \times \mathbb{T}^2)} \lesssim \|f^0\|_{H^1(\mathbb{T}^2 \times \mathbb{T}^2)} e^{-\alpha t}, \quad \forall t \geq 0.$$

- ② Given the vector field  $u_{\text{vortex}}(\cdot, Y_t)$ ,  $\exists \beta > 0$  such that

$$\|f(t)\|_{H_\delta^1(B_1 \times B_r)} \lesssim \|f^0\|_{H_\delta^1(B_1 \times B_r)} e^{-\beta t}, \quad \forall t \geq 0.$$

# Examples of Ergodicity with a Random Vector Field

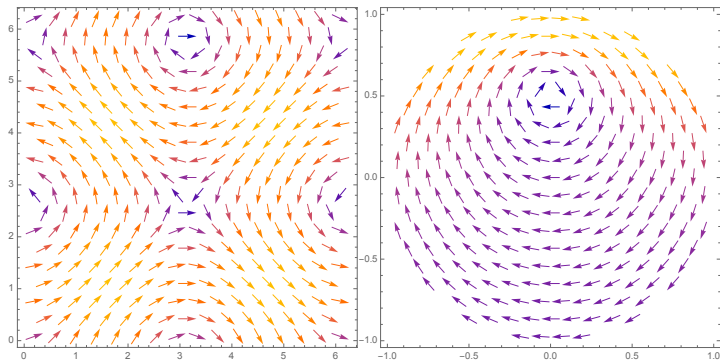


Figure: Streamlines of the two examples of vector fields with  $y = (0, 0.5)$ .

# Some Key References

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