

The mixing and stabilization by transport noise

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- 1 Transpot noise
- 2 Inviscid mixing by transport noise
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Consider the following stochastic equation on torus $\mathbb{T}^d := \mathbb{R}^d / \mathbb{Z}^d$ ($d \geq 2$),

$$du = \lambda u dt + \nu \Delta u dt + \circ dW_t \cdot \nabla u; \quad u(\cdot, 0) = u_0(\cdot), \quad (\text{SHE})$$

where $\lambda, \nu \geq 0$, $\circ d$ means the Stratonovich stochastic differential and $W(x, t)$ is a divergence-free space-time noise. We call this kind of noise as **transport noise**.

The **physical backgrounds** of transport noises in stochastic fluid dynamics:

- Mechanics: (Mikulevicius and Rozovskii, 2004), (Holm, 2015) ...
- Separation of scales: (Flandoli and Pappalettera, 2021), (Flandoli and Pappalettera, 2022), (Debussche and Pappalettera, 2023) ...

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This talk will specifically focus on the effects of transport noise on

- inviscid mixing ($\lambda = 0, \nu = 0$),
- stabilization ($\lambda \geq 0, \nu > 0$).

The form of noise

Consider transport noise as follows,

$$W(x, t) := \sqrt{C_d \kappa} \sum_{k \in \mathbb{Z}_0^d} \sum_{i=1}^{d-1} \theta_k \sigma_{k,i}(x) W_t^{k,i}, \quad (\text{Noise})$$

where $C_d = d/(d-1)$, noise intensity $\kappa > 0$, coefficient $\theta := \{\theta_k\}_k \in \ell^2(\mathbb{Z}_0^d)$ is symmetric, $\{\sigma_{k,i}\}_{k \in \mathbb{Z}_0^d, i=1, \dots, d-1}$ are divergence-free fields:

$$\sigma_{k,i}(x) := \vec{a}_{k,i} \exp\{2\pi i k \cdot x\}, \quad k \cdot \vec{a}_{k,i} = 0,$$

and $\{W^{k,i}\}_{k,i}$ are standard complex Brownian motions:

$$\overline{W_t^{k,i}} = W_t^{-k,i}, \quad [W^{k,i}, W^{l,j}]_t = 2t \delta_{k,-l} \delta_{i,j}.$$

Specific case: [Kraichnan noise](#),

$$\theta_k = \frac{c_\alpha}{|k|^{(d+\alpha)/2}}, \quad \forall k \in \mathbb{Z}_0^d; \quad c_\alpha = \left(\sum_k \frac{1}{|k|^{(d+\alpha)}} \right)^{1/2},$$

where $\alpha > 0$. [Kraichnan noise](#) is important in studying turbulence.

The form of noise

Under the above assumptions, (SHE) has the following Itô form

$$du = \lambda u dt + (\kappa + \nu)\Delta u dt + \sqrt{C_d \kappa} \sum_{k \in \mathbb{Z}_0^d} \sum_{i=1}^{d-1} \theta_k(\vec{a}_{k,i} \cdot \nabla u) e_k(x) dW_t^{k,i}.$$

The operator $\kappa\Delta$ is a "fake" dissipation term, since it is nullified by the martingale component in energy computations.

- 1 Transpot noise
- 2 **Inviscid mixing by transport noise**
 - Backgrounds of mixing
 - Our and previous works
 - Sketch of the Proof of mixing
 - Application: Regularized 2D-Euler equation
- 3 From mixing to stabilization
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Transport equation

Consider the transport equation on torus \mathbb{T}^d ,

$$\frac{d}{dt}u = b(x, t) \cdot \nabla u; \quad u(\cdot, 0) = u_0(\cdot), \quad (\text{TE})$$

where u_0 is mean zero and $b : \mathbb{T}^d \times [0, +\infty) \rightarrow \mathbb{R}^d$ is a divergence-free vector field. There is a corresponding Lagrangian flow $\varphi_{s,t} : \mathbb{T}^d \rightarrow \mathbb{T}^d$ as

$$\frac{d}{dt}\varphi_{s,t}(x) = b(\varphi_{s,t}(x), t); \quad \varphi_{s,s} = \text{Id}. \quad (\text{Flow})$$

Definition of mixing

In hydrodynamic systems, vector field $b(x, t)$ is said **exponential mixing**, if there exists $\beta > 0$ and $c_\beta, C_\beta > 0$ such that

$$\|u(t)\|_{H^{-\beta}} \leq C_\beta e^{-c_\beta t} \|u_0\|_{H^\beta}, \quad \forall u_0 \in H^\beta(\mathbb{T}^d).$$

It is equivalent to the Lagrangian flow $\varphi_{s,t}$ satisfies

$$|\langle f \circ \varphi_{s,t}, g \rangle| \leq C_\beta e^{-c_\beta(t-s)} \|f\|_{H^\beta} \|g\|_{H^\beta}, \quad \forall f, g \in H^\beta(\mathbb{T}^d),$$

which means the strong mixing we usually use.

Remark: The inviscid mixing roughly means that the frequency of fluid towards higher.

There are a lot of works on exponential mixing. We only list some of them,

- Deterministic fields: (Iyer, Kiselev, and Xu, 2014), (Yao and Zlatoš, 2017), (Alberti, Crippa, and Mazzucato, 2019), (Elgindi and Zlatoš, 2019) ...
- Random fields: (Bedrossian, Blumenthal, and Punshon-Smith, 2022), (Pappalettera, 2022), (Cooperman, 2023) ...
- Stochastic fields: (Gess and Yaroslavtsev, 2021), (Coti Zelati, Drivas, and Gvalani, 2024), (Luo, Tang, and Zhao, 2024) ...

(Alberti, Crippa, and Mazzucato, 2019) shows that the regularity (in space) of vector field $b(x, t)$ is closely related to the exponential mixing.

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(Bedrossian, Blumenthal, and Punshon-Smith, 2022): $b(x, t)$ is a solution of the Stochastic Navier-Stokes equation driven by an additive noise.

(Pappalettera, 2022): $b(x, t)$ is an Ornstein-Uhlenbeck velocity field.

(Cooperman, 2023): $b(x, t)$ is a random shear flow.

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- Stochastic fields: (Gess and Yaroslavtsev, 2021), (Coti Zelati, Drivas, and Gvalani, 2024), (Luo, Tang, and Zhao, 2024) ...

When $b(x, t)$ is a stochastic field, then (TE) rewrites as

$$du = \circ dW_t \cdot \nabla u; \quad u(\cdot, 0) = u_0(\cdot). \quad (\text{STE})$$

Our work: Mixing by transport noise

Consider the (SHE) with $\lambda = 0$ and $\nu = 0$:

$$du = \kappa \Delta u dt + \sqrt{C_{d\kappa}} \sum_{k \in \mathbb{Z}_0^d} \sum_{i=1}^{d-1} \theta_k (\vec{a}_{k,i} \cdot \nabla u) e_k(x) dW_t^{k,i}. \quad (\text{STE-1})$$

Theorem 2.1 (Luo, T., Zhao, 2024⁺; arXiv:2402.07484)

If initial data $u_0 \in L^2(\mathbb{T}^d)$ and noise coefficients $\{\theta_k\}_k \in \ell^2(\mathbb{Z}_0^d)$, then the solution u of Eq. (STE-1) is exponential mixing.

(i) $\forall \beta > 0$, there exists a constant $C_{\beta,d} > 0$ such that

$$\mathbb{E} \|u(t)\|_{H^{-\beta}}^2 \leq C_{\beta,d} e^{-\kappa C(\theta,d) t/4} \|u_0\|_{L^2}^2,$$

where $C(\theta, d) \simeq_d \|\theta\|_{h^{-1}}^2 := \sum_k |k|^{-2} \theta_k^2$.

(ii) If $\|\theta\|_{h^1}^2 := \sum_k |k|^2 \theta_k^2 < +\infty$, then for any $0 < \lambda_0/\kappa < D(\theta, d) \simeq_d \|\theta\|_{h^{-1}}^2$ and $q \in (0, \frac{D(\theta,d)\kappa}{\lambda_0} - 1)$, there is $C_{\kappa,\theta,d}(\omega)$ with finite q -moment such that

$$\mathbb{P}\text{-a.s.}, \quad \|u(t)\|_{H^{-1}}^2 \leq C_{\kappa,\theta,d}(\omega) e^{-\lambda_0 t} \|u_0\|_{L^2}^2, \quad \forall t \geq 0.$$

Comparison with previous works

Consider the following equation on manifold \mathcal{M}^1

$$du = \kappa \sum_n \langle \sigma_k(x), \nabla u \rangle_{T\mathcal{M}} \circ dW_t^{k,i}. \quad (\text{STE-2})$$

Theorem 2.2 (Gess and Yaroslavtsev, 2021; arXiv:2104.03949.)

Let $\{\sigma_k(x)\}_k$ satisfy the Hörmander condition, elliptic conditions and some regularity conditions. Let the initial data $u_0 \in H^\beta(\mathcal{M}) \cap H^1(\mathcal{M})$, $\beta > 0$. Then the solution u to (STE-2) is exponential mixing:

$$\mathbb{P}\text{-a.s.}, \quad \|u(t)\|_{H^{-\beta}} \leq C(\omega) e^{-\lambda_0 t} \|u_0\|_{H^\beta}, \quad \forall t \geq 0.$$

This result is coming from the exponential ergodicity of the two-point Lagrangian flow $(\varphi_t(x), \varphi_t(y))$ on off-diagonal set \mathcal{D}^c and the fact

$$\mathbb{E}|\langle f \circ \varphi_t, g \rangle|^2 = \int_{\mathcal{M} \times \mathcal{M}} P_t(f \times f)(x, y) g(x)g(y) \mu(dx)\mu(dy).$$

¹ \mathcal{M} is a d -dimensional C^∞ -smooth compact Riemannian-manifold.

Comparison with previous works

The divergence-free fields $\{\sigma_k(x)\}$ specifically need satisfy

- Hörmander condition: $Lie(\sigma_1, \dots, \sigma_K)(x) = T_x \mathcal{M}$.
- Elliptic condition for $(\sigma_k(\cdot), \sigma_k(\cdot))_k$ on \mathcal{D}^c :

$$\sum_k \left| \langle \sigma_k(x_1), v_1 \rangle_{x_1} + \langle \sigma_k(x_2), v_2 \rangle_{x_2} \right|^2 \geq C(|v_1|^2 + |v_2|^2),$$

where $v_i \in T_{x_i} \mathcal{M}$ for $i = 1, 2$.

- Elliptic condition for the normalized tangent flow: ...
- Regularity conditions: there exists $\alpha \in (0, 1]$ such that

$$(x, y) \mapsto \sum_k D\sigma_k(x) \otimes D\sigma_k(y), \quad (x, y) \mapsto \sum_k D^2\sigma_k(x) \otimes \sigma_k(y)$$

are α -Hölder continuous, and ...

Note: If $\mathcal{M} = \mathbb{T}^d$, the above conditions are **more difficult to verify and stronger** than those in our work.

Comparison with previous works

Consider the following equation on \mathbb{R}^d ,

$$du = \nabla u \circ dW_t. \quad (\text{STE-3})$$

$W(x, t)$ is a specific Kraichnan noise on \mathbb{R}^d , where

$$\mathbb{E}W(x, t) = 0, \quad \mathbb{E}[W(x, t), W(y, s)] = D(x - y)(t \wedge s),$$

with the isotropic covariance matrix D ,

$$D(0) - D(r) = D_1 \left[\text{Id} + \left(\frac{2}{d-1} \right) (\text{Id} - r \otimes r) \right] |r|^2.$$

Let $I_\beta(r) := \frac{1}{r^{d-2\beta}}$. Due to $\|h\|_{H^{-\beta}}^2 = \langle h, I_\beta * h \rangle_{L^2}$ and

$$(D(0) - D(r)) : \nabla_r \otimes \nabla_r I_\beta = -\lambda_{d,\beta} I_\beta,$$

Itô formula gives the exponential mixing in the average.

Comparison with previous works

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Theorem 2.3 (Coti Zelati, Drivas, and Gvalani, 2024; J. Stat. Phys.)

Fix $\beta \in [0, d/2)$, the solution u to Eq. (STE-3) satisfies

$$\mathbb{E} \|u(t)\|_{H^{-\beta}}^2 \leq e^{-\lambda_{d,\beta} t} \|u_0\|_{H^{-\beta}}^2, \quad t \geq 0,$$

where $\lambda_{d,\beta} = 2D_1\beta(d - 2\beta)$.

Comparison with previous works – Summary from results

$$du = \sum_k \sigma_k(x) \cdot \nabla u \circ dW_t^k. \quad (\text{abstract-STE})$$

Literature & Result	Space	Initial data	Noise
A: Mixing in \mathbb{P} -a.s.	\mathcal{M}	$H^\beta(\mathcal{M}) \cap H^1(\mathcal{M})$	Hörmander conditions and some regularity conditions
B: Mixing in average	\mathbb{R}^d	Not specified ¹	Kraichnan noise
C: Mixing in average	\mathbb{T}^d	$L^2(\mathbb{T}^d)$	$\{\theta_k\}_k \in \ell^2(\mathbb{Z}_0^d)$
C: Mixing in \mathbb{P} -a.s.	\mathbb{T}^d	$L^2(\mathbb{T}^d)$	$\{\theta_k\}_k \in h^1(\mathbb{Z}_0^d)$

A : (Gess and Yaroslavtsev, 2021)

B : (Coti Zelati, Drivas, and Gvalani, 2024)

C : Our results (Luo, Tang, and Zhao, 2024)

¹At least ensuring the well-posedness and belonging to $H^{-\beta}(\mathbb{R}^d)$.

$$du = \sum_k \sigma_k(x) \cdot \nabla u \circ dW_t^k. \quad (\text{abstract-STE})$$

A (Gess and Yaroslavtsev, 2021): Studying the two-point Lagrangian flow $(\varphi_t(x), \varphi_t(y))$ and proving its exponential ergodicity.

B (Coti Zelati, Drivas, and Gvalani, 2024): Using the identity about this special Kraichnan noise. It doesn't work for torus and other noises.

Using the symmetry of noise coefficient $\theta := \{\theta\}_k$ and the particular spectrum of Δ on \mathbb{T}^d , we find an elementary and direct approach to the exponential mixing.

Sketch of the Proof of mixing

For simplicity, we consider $d = 2$. Eq. (STE-1) rewrites as

$$du = \kappa \Delta u dt + \sqrt{2\kappa} \sum_{k \in \mathbb{Z}_0^2} \theta_k \left(\frac{k^\perp}{|k|} \cdot \nabla u \right) e_k(x) dW_t^k, \quad u(0, \cdot) \in L^2(\mathbb{T}^2).$$

Then $\mathbb{E}|\hat{u}_k|^2$ satisfies the infinite dimensional ODE:

$$\begin{aligned} \frac{d}{dt} \mathbb{E}|\hat{u}_k|^2 &= -8\pi^2 \kappa |k|^2 \mathbb{E}|\hat{u}_k|^2 + 16\pi^2 \kappa \sum_{l \in \mathbb{Z}_0^2} \theta_l^2 \frac{|k \cdot l^\perp|^2}{|l|^2} \mathbb{E}|\hat{u}_{k-l}|^2 \\ &= 16\pi^2 \kappa \sum_{l \in \mathbb{Z}_0^2} \theta_l^2 \frac{|k \cdot l^\perp|^2}{|l|^2} \left(\mathbb{E}|\hat{u}_{k-l}|^2 - \mathbb{E}|\hat{u}_k|^2 \right). \end{aligned} \quad (\text{ODE})$$

To prove exponential mixing, we only need to show that the ℓ^p norm of $Y_k := \mathbb{E}|\hat{u}_k|^2$ decays exponentially for some $p \in (1, +\infty)$. By symmetry,

$$\frac{d}{dt} \sum_k Y_k^p = -8\pi^2 \kappa p \sum_{k, l \in \mathbb{Z}_0^2} \theta_l^2 \frac{|k \cdot l^\perp|^2}{|l|^2} (Y_{k+l} - Y_k) (Y_{k+l}^{p-1} - Y_k^{p-1}).$$

Sketch of the Proof of mixing – Continue

To prove the exponential decay rate of $\sum_k Y_k^p(t)$, we need

$$\sum_k Y_k^p \leq c \sum_{k,l} \theta_l^2 \frac{|k \cdot l^\perp|^2}{|l|^2} (Y_{k+l} - Y_k) (Y_{k+l}^{p-1} - Y_k^{p-1})$$

for some $c > 0$. We have the following inequality in one dimension

$$\sum_{n \in \mathbb{N}} a_n^p \leq \frac{2p^2}{p-1} \sum_{n \in \mathbb{N}} (n+1)^2 (a_{n+1}^{p-1} - a_n^{p-1}) (a_{n+1} - a_n).$$

Let $\Gamma(k_0, n) = k_0 + \lfloor \frac{n+1}{2} \rfloor l + \lfloor \frac{n}{2} \rfloor l^\perp$ with $k_0 \cdot l^\perp > 0$. Noticing

$$\frac{|\Gamma(k_0, n) \cdot l^\perp|^2}{|l|^2} = \left| \frac{k_0 \cdot l^\perp}{|l|} + \lfloor \frac{n}{2} \rfloor \right|^2 \approx \frac{(n+1)^2}{4},$$

we can decompose \mathbb{Z}_0^2 into countable orbits as the above form and apply the inequality to obtain the exponential decay of $\sum_k Y_k^p(t)$.

Sketch of the Proof of mixing – Continue

We give a diagram of this decomposition,

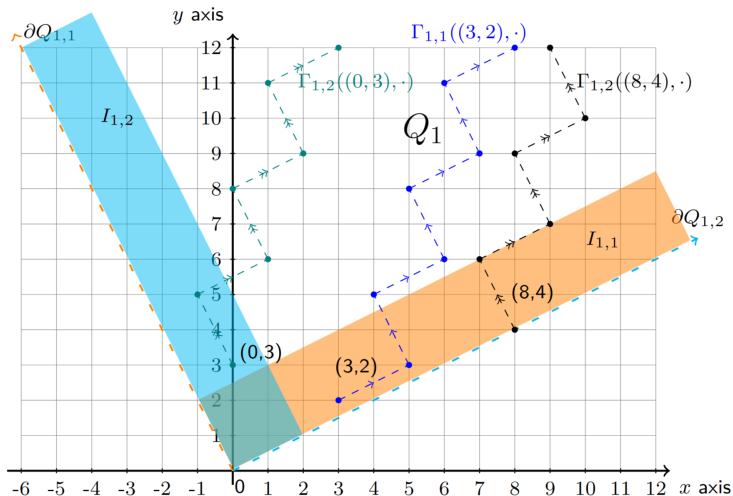


Figure: The decomposition of Q_1 for $l = (2, 1)$

Combining the above estimates, we obtain

$$\|Y(t)\|_{\ell^p} \leq e^{-\kappa\pi^2 \frac{1}{p}(1-\frac{1}{p})\|\theta\|_{h^{-1}}^2 t} \|Y(0)\|_{\ell^p}, \quad \forall t > 0,$$

where $Y(t) := \{Y_k(t)\}_k$. By Hölder inequality, for $\beta > 1/2$,

$$\begin{aligned} \mathbb{E}\|u(t)\|_{H^{-\beta}}^2 &= \mathbb{E} \sum_k \frac{|\hat{u}_k(t)|^2}{|2\pi k|^{2\beta}} \leq \left(\sum_k \frac{1}{|2\pi k|^{4\beta}} \right)^{1/2} \left(\sum_k (\mathbb{E}|\hat{u}_k(t)|^2)^2 \right)^{1/2} \\ &\leq C e^{-\frac{\kappa\pi^2}{4}\|\theta\|_{h^{-1}}^2 t} \|Y(0)\|_{\ell^2}^4 \leq C e^{-\frac{\kappa\pi^2}{4}\|\theta\|_{h^{-1}}^2 t} \|u_0\|_{L^2}^2. \end{aligned}$$

Similarly, the exponential mixing in average also holds for $0 < \beta \leq 1/2$.

Sketch of the Proof of mixing – Continue

To obtain the exponential mixing in \mathbb{P} -a.s. from mixing in average, we need the Borel–Cantelli lemma and the following boundness lemma.

Lemma 2.4 (Luo, T., Zhao, 2024⁺; arXiv:2402.07484)

If initial data $u_0 \in L^2(\mathbb{T}^d)$ and noise coefficient θ satisfies $\|\theta\|_{h^1}^2 < +\infty$, then the solution u to the stochastic transport equation (STE-1) satisfies

$$\mathbb{E} \left[\sup_{t \in [0, t_0]} \|u(t)\|_{H^{-1}}^2 \right] \leq 2 \|u_0\|_{H^{-1}}^2,$$

where $t_0 = \left(\frac{\sqrt{11}-3}{16}\right)^2 \frac{1}{\pi^2 d \kappa \|\theta\|_{h^1}^2} > 0$.

Application: Regularized 2D-Euler equation

We apply our results to the nonlinear equations, such as the regularized stochastic 2D Euler equation with $\alpha > 0$ on \mathbb{T}^2 :

$$\begin{cases} dw + u \cdot \nabla w dt = \sqrt{2\kappa} \sum_k \theta_k \sigma_k \cdot \nabla w \circ dW_t^k, \\ u = \operatorname{curl}^{-1}(-\Delta)^{-\alpha/2} w, \quad w(0, \cdot) \in L^2(\mathbb{T}^2), \end{cases} \quad (\text{R-Euler})$$

where curl^{-1} is Biot–Savart operator.

Note: similar to the 2D Euler equation, the regularized equation

$$\frac{d}{dt} w + u \cdot \nabla w = 0, \quad u = \operatorname{curl}^{-1}(-\Delta)^{-\alpha/2} w,$$

has the following identity when w is smooth,

$$\|w(t)\|_{H^{-1-\alpha}} = \|w(0)\|_{H^{-1-\alpha}}, \quad \forall t \geq 0.$$

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$$\begin{cases} dw + u \cdot \nabla w dt = \sqrt{2\kappa} \sum_k \frac{\sigma_k \cdot \nabla w}{K_\alpha |k|^{1+\alpha}} \circ dW_t^k, \\ u = \operatorname{curl}^{-1} (-\Delta)^{-\alpha/2} w, \quad w(0, \cdot) \in L^2(\mathbb{T}^2), \end{cases} \quad (\text{R-Euler})$$

where curl^{-1} is Biot–Savart operator and K_α is a normalizing constant.

Let $\tilde{W}_t^k := W_t^k - \int_0^t K_\alpha |k|^{1+\alpha} \frac{\langle u(s), \sigma_k \rangle}{\sqrt{2\kappa}} ds$, then by Girsanov theorem, \tilde{W}_t^k is a Brownian motion under $\tilde{\mathbb{P}}$. Then (R-Euler) rewrites

$$dw = \kappa \Delta w dt + \sqrt{2\kappa} \sum_k \frac{\sigma_k \cdot \nabla w}{K_\alpha |k|^{1+\alpha}} d\tilde{W}_s^k.$$

Thus, w is exponential mixing in average with $\tilde{\mathbb{P}}$.

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where curl^{-1} is Biot–Savart operator and K_α is a normalizing constant.

Theorem 2.5 (Luo, T., Zhao, 2024⁺; arXiv:2402.07484)

For fixed $R > 0$ and initial data $w_0(\cdot) := w(0, \cdot) \in L^2(\mathbb{T}^2)$. If $\|w_0\|_{L^2}^2 \leq R$, then for any $\lambda > 0$, there exist a noise intensity $\kappa(\lambda, R)$ such that

$$\mathbb{E}\|w(t)\|_{H^{-1-\alpha}}^2 \leq C e^{-\lambda t} \|w_0\|_{L^2}^2, \quad \forall t \geq 0.$$

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Backgrounds of stabilization by noise

The trivial solution $X_t \equiv 0$ of ODE

$$\dot{X}_t = A_0 X_t, \quad X_0 \in \mathbb{R}^d,$$

is unstable if matrix $A_0 \in \mathbb{R}^{d \times d}$ has positive eigenvalues.

(Arnold, Crauel, and Wihstutz, 1983) showed that suitable linear noise induce **asymptotic stability**. If $\text{Tr}A_0 < 0$, then there exists skew-symmetric matrices A_k such that the solution to

$$dX_t = A_0 X_t dt + \kappa \sum_{k=1}^{d-1} A_k X_t \circ dW_t^k$$

is \mathbb{P} -a.s. exponentially decays with an exponential rate arbitrarily close to $\frac{1}{d} \text{Tr}A_0 < 0$ for κ large enough.

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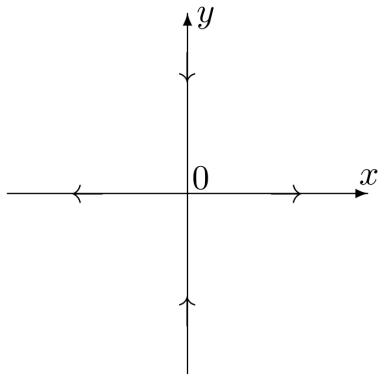
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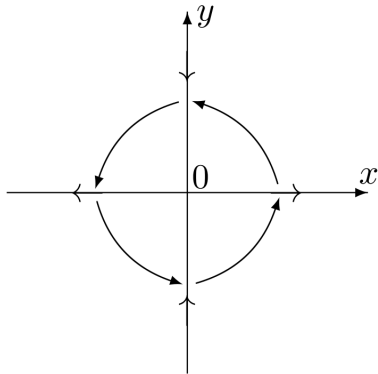
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Backgrounds of stabilization by noise

Consider $A_0 = \begin{pmatrix} 1 & 0 \\ 0 & -2 \end{pmatrix}$ and $\text{Tr}A_0 = -1 < 0$.



$$\dot{X}_t = A_0 X_t$$



$$dX_t = A_0 X_t dt + \kappa \sum_{k=1}^{d-1} A_k X_t \circ dW_t^k$$

Can this result be extended to the infinite dimension?

Let $\lambda \geq 0$ and $\nu > 0$. Recall Eq. (SHE) as

$$du = \lambda u dt + \nu \Delta u dt + \circ dW_t \cdot \nabla u; \quad u(\cdot, 0) = u_0(\cdot).$$

Due to $\text{Tr}(\nu \Delta + \lambda) = -\infty$, Capiński in the late 80s formulated the conjecture that under suitable noise, the solution of (SHE) exponentially decays with an exponential rate arbitrarily close to $-\infty$.

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Theorem 3.1 (Luo, T., Zhao, 2024⁺; arXiv:2402.07484)

Given initial datum $u_0 \in L^2(\mathbb{T}^d)$ and $\theta \in \ell^2(\mathbb{Z}_0^d)$, the solution u satisfies

$$\mathbb{E} \|u(t)\|_{L^2}^2 \leq C_0 \frac{8\pi^2\nu + D(\theta, d)\kappa}{2\nu} e^{-(2\lambda + 8\pi^2\nu + D(\theta, d)\kappa)t} \|u_0\|_{L^2}^2,$$

where $C_0 \geq \frac{1}{4\pi^2}$ and $D(\theta, d) \simeq_d \|\theta\|_{h^{-1}}^2$. There are noise parameters (κ, θ) such that $2\lambda < 8\pi^2\nu + D(\theta, d)\kappa$, then $\mathbb{E} \|u(t)\|_{L^2}^2$ decays exponentially.

The similar result also holds in \mathbb{P} -a.s. sense.

We follow the idea of (Gess and Yaroslavtsev, 2021) to obtain the stabilization from the exponential mixing. The key point is






$$\frac{d}{dt} \frac{e^{2\lambda t}}{\mathbb{E}\|u(t)\|_{L^2}^2} = \frac{2\nu e^{2\lambda t}}{(\mathbb{E}\|u(t)\|_{L^2}^2)^2} \mathbb{E}\|u(t)\|_{H^1}^2 \geq \frac{2\nu e^{2\lambda t}}{\mathbb{E}\|u(t)\|_{H^{-1}}^2}.$$






Using the exponential mixing:





$$\mathbb{E}\|u(t)\|_{H^{-1}}^2 \leq C e^{-(2\lambda + 8\pi^2\nu + D(\theta, d)\kappa)t} \|u_0\|_{L^2}^2,$$



we obtain the stabilization for $\kappa > 0$ big enough.

Thank you !

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