Stochastic transport models of turbulence

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Stochastic equations and particle systems

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Joint work with Marco Bagnara, Lucio Galeati and Mario Maurelli

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An Unassuming Laminar Flow



Shear flow past an obstacle at Re = 0.16 (aluminum dust in water, *An Album of Fluid Motion*, van Dyke 1982).

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Turbulence emerges: Chaos ensues



Turbulence past an obstacle at Re= 2000 (air bubbles in water, *An Album of Fluid Motion*, van Dyke 1982).

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- A linear, stochastic PDE model for turbulence.
- Anomalous Regularization.
- Some integral asymptotics.

Advection of Passive Scalars

Features of (Nonlinear, 3D) Turbulence

Kolmogorov-Obukhov predictions on structure functions and energy cascades in isotropic turbulence:

$$E\left[|u(x+r) - u(x)|^2\right] \sim \varepsilon^{2/3} r^{2/3} \quad r \to 0,$$
$$E\left[E(k)\right] = E\left[\left|\int_{|\underline{k}|=k} |\hat{u}(\underline{k})|^2 d\underline{k}\right|\right] \sim \varepsilon^{2/3} k^{-5/3} \quad \text{(inertial range)}$$

(ε the energy dissipation rate, assumed scale-independent) Anomalous dissipation of energy:

$$\lim_{\nu \to 0} \nu \int_0^T \|\nabla u_t^\nu\|_{L^2}^2 \, \mathrm{d}t > 0$$

These phenomena should be reproduced by sol. of 3D Navier-Stokes eqs.

$$\begin{cases} \partial_t u + (u \cdot \nabla)u = \nu \Delta u + \nabla p \\ \nabla \cdot u = 0 \end{cases}$$

with $u \ll 1$, however simpler phenomenological models are sufficient.

Advection of a scalar field ρ by a (decoupled) random vector field $u : \mathbb{R}^d \to \mathbb{R}^d$ (and possibly small viscosity ν),

$$\partial_t \rho + \mathbf{u} \cdot \nabla \rho = \nu \Delta \rho,$$

is able to replicate many features of turbulence.

(first proposals by Obukhov, Corrsin, Batchelor, see Sreenivasan '18 for a survey).

Kraichnan considered the case where u is a isotropic Gaussian field, delta-correlated in time, with power-law covariance spectrum.

(Kraichnan Phys. Fluids '68, '70, J. Fluid Mech. '74, '76, '76, PRL '89)

The model allows explicit (although sometimes only formal) computations and can replicate the energy cascade, anomalous dissipation, spontaneous stochasticity.

Kraichnan's Model

Consider $u = \frac{\partial}{\partial t}W(t,x)$ as a sample of a divergence-free, isotropic Gaussian random velocity field Gaussian velocity field, study the Stratonovich transport SPDE

$$d\rho + \circ dW \cdot \nabla \rho = 0, \quad \rho : [0, \infty) \times \mathbb{R}^d, \ d \ge 2.$$

We choose covariance with power law spectrum,

$$E[W(t,x)\otimes W(s,y)]=(t\wedge s)Q(x-y), \quad \hat{Q}(\xi)=rac{C_d}{\left(1+|\xi|^2
ight)^{d/2+lpha}}P_\xi^\perp,$$

that is

$$Q(0)-Q(x)=|x|^{2lpha}\left(I-rac{2lpha}{d-1}P_x^{\perp}
ight)+o\left(|x|^{2lpha}
ight),\quad |x| o 0.$$

Regular Kraichnan's Model

When $\alpha > 1$, W(t, x) is continuously differentiable in x, and

- there exists a C¹-regular, measure preserving stochastic flow X_t(x) for the underlying SDE dX_t = W (◦dt, X_t);
- $\rho_t(x) = \rho_0(X_t^{-1}(x))$ and

$$\|
ho_t\|_{L^2} = \|
ho_0\|_{L^2} \quad \forall t \ge 0, \quad \mathbb{P} ext{-a.s.}$$

- Initial data δ_x evolve as $\delta_{X_t(x)}$.
- Explicitly computable Lyapunov exponents, strictly positive top Lyapunov $\lambda = \lim_{t \to \infty} \frac{1}{t} \log |DX_t(x)|$.

cf. Le Jan '85, Baxendale-Harris '86

• On compact manifolds: exponential mixing of passive scalars.

cf. Dolgopyat-Kaloshin-Koralov '04, Gess-Yaroslavtsev '21.

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When $\alpha \in (0, 1)$, W is Hölder continuous and the SPDE is well-posed, and (strong existence, pathwise uniqueness for $\rho_0 \in L^1$, cf. Le Jan-Raimond '02, Maurelli '11, Galeati-Luo '23)

- Spontaneous stochasticity: particle trajectories X_t , Y_t starting at the same position x depart istantaneously.
- SDE dynamics not described by a map x → X_t(x) but a flow of Markovian kernels.
- Initial data δ_x of the SPDE become L^1 -densities at t > 0.
- Solutions to SPDE are limit of vanishing viscosity/smooth approximations.
- Anomalous dissipation: $\|\rho_t\|_{L^2} < \|\rho_0\|_{L^2}$ with positive probability. (related to spontaneous stochasticity, cf. Drivas-Evink '17)

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Evolution of Multiscale Norms

Negative Sobolev norms

$$\|f\|_{\dot{H}^{-s}}^2 = \int_{\mathbb{R}^d} |\xi|^{-2s} |\hat{f}(\xi)|^2 \, \mathrm{d}\xi$$

allow to gauge mixing behaviour and cascade mechanisms. Previous results include:

 α > 1 : [Gess, Yaroslavtsev '21] show pathwise exponential decay of negative Sobolev norms (uniform-in-diffusivity estimates):

$$\|\rho_t\|_{\dot{H}^{-s}} \leq De^{-\gamma t} \|\rho_0\|_{\dot{H}^s} \quad \forall t \geq 0, \quad E\left[|D|^p\right] < \infty.$$

• $\alpha = 1$, statistically self-similar case: [Coti Zelati, Gvalani, Drivas '24]

$$E\left[\|\rho_t\|_{\dot{H}_{-s}}^2\right] = e^{-\lambda_{d,s}t} \|\rho_0\|_{\dot{H}^{-s}}^2, \quad \lambda_{d,s} \sim 2s(d-2s)$$

• $\alpha < 1$, on \mathbb{T}^d [Rowan '23] proves exponential decay of energy:

$$E\left[\left\|\rho_{t}\right\|_{L^{2}}^{2}\right] \leq e^{-Ct}\left\|\rho_{0}\right\|_{L^{2}}^{2}, \quad \forall t \geq \tau(\alpha)$$

Regularization in Rough Kraichnan's Model

Anomalous Regularization for $\alpha < 1$

For $\alpha > 1$, $\rho_t(x) = \rho_0(X_t^{-1}(x))$, reversible dynamics and no regularization.

Theorem (Galeati-G.-Maurelli)

Let $d \ge 2, \alpha \in (0, 1), s \in (0, d/2)$. There exist positive constants $C_1, C_2 > 0$ such that

$$\frac{d}{dt}E\left[\left\|\rho_{t}\right\|_{\dot{H}^{-s}}^{2}\right]+C_{1}E\left[\left\|\rho_{t}\right\|_{\dot{H}^{1-\alpha-s}}^{2}\right]\leq C_{2}E\left[\left\|\rho_{t}\right\|_{\dot{H}^{-s}}^{2}\right]\quad\forall t>0$$

In particular, for T > 0

$$\sup_{t\in[0,T]} E\left[\|\rho_t\|_{H^{-s}}^2 \right] + C_1 E\left[\int_0^T \|\rho_t\|_{H^{1-\alpha-s}}^2 \, \mathrm{d}t \right] \le e^{2C_2 T} E\left[\|\rho_0\|_{H^{-s}}^2 \right].$$

As a consequence, the solution map extends uniquely to any initial data $\rho \in \dot{H}^{-s}$ with $s \in (0, d/2)$ and solutions become istantaneously L^2 -regular.

About Regularization

- Result motivated by [Coghi, Maurelli '23] who obtained the result for d = 2, s = 1 and applied it to regularization by Kraichnan noise for 2D Euler.
- Solutions become regular: \mathbb{P} -a.s. $\rho \in L^2_t H^{1-\alpha-}$.
- Optimality of threshold s < d/2, up to equality: white-noise (which is supported on $H^{-d/2-}$) is formally invariant.
- Constant C₁ dictated by the local Hölder behaviour of W, while constant C₂ by how much W deviates from self-similarity. In the (extremely formal!) statistically self-similar case, the balance becomes

$$\frac{d}{dt}E\left[\left\|\rho_{t}\right\|_{\dot{H}^{-s}}^{2}\right]+C_{1}E\left[\left\|\rho_{t}\right\|_{\dot{H}^{1-\alpha-s}}^{2}\right]=0$$

Heuristics

Given a smoothing kernel G, ρ solution as above,

$$\begin{aligned} \frac{d}{dt} E\left[\langle G * \rho_t, \rho_t \rangle\right] + 2\nu E\left[\langle G * \nabla \rho_t, \nabla \rho_t \rangle\right] \\ &= E\left[\int_{\mathbb{R}^d \times \mathbb{R}^d} \operatorname{Tr}\left((Q(0) - Q(x - y))D^2 G(x - y)\right)\rho_t(x)\rho_t(y)dx \, \mathrm{d}y\right] \\ &=: E\left[\langle H * \rho_t, \rho_t \rangle\right]. \end{aligned}$$

Consider the (ill-defined) self-similar case $\hat{Q}(\xi) = c_d |\xi|^{-d-2\alpha} P_{\xi}^{\perp}$. Take $R_s(z) = |z|^{d-2s}$, so that $\langle R_s * \rho_t, \rho_t \rangle \sim \|\rho_t\|_{\mathcal{H}^{-s}}^2$. By an explicit computation:

$$H(z) = 2(2s-d)(s+\alpha-1)|z|^{2(s+\alpha-1)-d} = -2(d-2s)(s+\alpha-1)R_{s+\alpha-1}(z)$$

(consistent with $\alpha = 1$ examined in Coti Zelati-Gvalani-Drivas '24)

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Lemma

Let $d \ge 2, \alpha \in (0, 1)$; let $\rho_0 \in L^1 \cap L^2 \cap \dot{H}^{-s}, \rho$ associated solution. Then

$$\frac{d}{dt}E\left[\left\|\rho_{t}\right\|_{\dot{H}^{-s}}^{2}\right] = \int_{\mathbb{R}^{d}}F(\xi)E\left[\left|\hat{\rho}_{t}(\xi)\right|^{2}\right]d\xi - 2\nu E\left[\left\|\rho_{t}\right\|_{\dot{H}^{-s}}^{2}\right]$$

for the flux function $F = F(\alpha, s, d)$ defined by

$${\sf F}(\xi):=\int_{\mathbb{R}^d}rac{1}{(1+|\xi-\eta|^2)^{rac{d}{2}+lpha}}\left|{\sf P}_{\xi-\eta}^{ot}\xi
ight|^2\left(rac{1}{|\eta|^{2s}}-rac{1}{|\xi|^{2s}}
ight)d\eta$$

We are therefore reduced to study the high-frequency asymptotics of $F(\xi)$.

Beyond Regularization in Kraichnan's model

- Rigorous derivation of intermittency in the Kraichnan model.
- Study more realistic and complicated models (e.g. *W* not white in time, replaced by a solution to an SPDE).
- Rigorously derivation of anomalous dissipation and anomalous regularization in nonlinear SPDEs driven by "Kraichnan noise".

(recent results in Coghi-Maurelli '23, Bagnara-Galeati-Maurelli '24)

Results in the nonlinear case are restricted to $\alpha \in (0, 1/2)$, but a physically motivated choice would be $\alpha = 2/3$.

• Regularization by Kraichnan's transport noise for vector fields, aiming to an application to 3D Navier-Stokes.

Theorem (Bagnara-G.-Maurelli)

The induction equation

$$\partial_t B + (u \cdot \nabla) B = (B \cdot \nabla) u, \quad \nabla \cdot B = 0,$$

driven by the random field u as in Kraichnan's Model, has a unique solution in $L^2_{t,\omega}(H^{-s+(1-\alpha)})$ if the initial datum is in H^{-s} , for $d \ge 3$, small enough $\alpha \in (0, 1)$ and an appropriate negative s.

Asymptotics

A rather tough integral

The asymptotics at $|\xi| \to \infty$ of the flux function

$$F(\xi) := \int_{\mathbb{R}^d} \frac{1}{(1+|\xi-\eta|^2)^{\frac{d}{2}+\alpha}} \left| P_{\xi-\eta}^{\perp} \xi \right|^2 \left(\frac{1}{|\eta|^{2s}} - \frac{1}{|\xi|^{2s}} \right) d\eta$$

can be reduced to that of

$$J(\lambda) = \int_0^\infty h(\lambda t) f(t) dt,$$

at $\lambda(=|\xi|) o \infty$, with

$$f(r) = r^{d-1} \int_0^{\pi} \frac{\sin^d \theta d\theta}{|1 - 2r \cos \theta + r^2|^s}, \quad h(r) = \frac{1}{(1 + r^2)^{d/2 + \alpha}}.$$

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The Mellin transform of f is defined by

$$M[f,z] = \int_0^\infty t^{z-1} f(t) dt$$

for $z \in \mathbb{C}$ for which the integral is absolutely convergent.

$$J(\lambda) = \int_0^\infty h(\lambda t) f(t) dt = \frac{1}{2\pi i} \int_{r-i\infty}^{r+i\infty} \frac{M[h,z]M[f,1-z]}{\lambda^z} dz,$$

if the line $r + i\mathbb{R}$ is entirely included in the intersection of the fundamental strips of the involved Mellin transforms.

Taking the difference of Parseval formulas for two lines

$$J(\lambda) = \frac{1}{2\pi i} \int_{r-i\infty}^{r+i\infty} \frac{M[h, z]M[f, 1-z]}{\lambda^z} dz$$

= $\sum_{r < \operatorname{Re} z < r'} \operatorname{Res} \left\{ -\lambda^{-z} M[h, z] M[f, 1-z] \right\}$
+ $\frac{1}{2\pi i} \int_{r'-i\infty}^{r'+i\infty} \frac{M[h, z] M[f, 1-z]}{\lambda^z} dz,$

so we can compute an asymptotic expansion by evaluating residues. The relevant pole (producing the leading coefficient) is at $z = d + 2\alpha$, and we need the residue to be a **positive** real number.

Computing Coefficients

One therefore needs to evaluate

$$M[h,z] = \frac{\Gamma(z/2)\Gamma(d/2 + \alpha - z/2)}{2\Gamma(d/2 + \alpha)},$$

and the crucial

$$\begin{split} M[f,1-z] &= \int_0^\infty r^{d-1-z} \int_0^\pi \frac{\sin^d \theta d\theta}{|1-2r\cos\theta+r^2|^s} \\ &= \frac{\sqrt{\pi}\Gamma\left(\frac{d-2s+2}{2}\right)\Gamma\left(\frac{d+1}{2}\right)}{2\Gamma(s)} \cdot \frac{\Gamma\left(\frac{2s-d+z}{2}\right)\Gamma\left(\frac{d-z}{2}\right)}{\Gamma\left(\frac{z+2}{2}\right)\Gamma\left(\frac{2d-2s+2-z}{2}\right)}. \end{split}$$

Why the explicit computation? The relevant pole comes from M[h, z], and it is located **outside of the analyticity region of** M[f, 1-z].

Theorem

Let $\alpha \in (0,1)$ and $s \in (0, d/2)$. There exists a finite constant $C_{d,\alpha,s} > 0$ such that

$$|F(\xi)+K_{d,lpha,s}|\xi|^{2-2lpha-2s}|\leq C_{d,lpha,s}|\xi|^{-2s}\quad orall\,\xi\in\mathbb{R}^d\setminus\{0\}.$$

where

$$\mathcal{K}_{d,\alpha,s} = -\frac{2^{d/2-1}(d+1)\Gamma(s+\alpha)\Gamma(-\alpha)\Gamma\left(\frac{d-2s+2}{2}\right)}{\Gamma(s)\Gamma\left(\frac{d+2\alpha+2}{2}\right)\Gamma\left(\frac{d-2s+2-2\alpha}{2}\right)} > 0.$$

Thank you!

(arXiv:2407.16668, arXiv:2411.09482)