# Averaging vs Chaos in Turbulent Transport?

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## The Model

PDE in  $\mathbb{R}^d$ 

$$\partial_t T + v \cdot \nabla T = \kappa \Delta T$$

 $\kappa > 0. v \in (C(\mathbb{R}^d))^d$ ,  $\operatorname{div}(v) = 0.$ 

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## The Model

PDE in  $\mathbb{R}^d$ 

$$\begin{split} \partial_t T + \underbrace{v} \cdot \nabla T &= \kappa \Delta T \\ \kappa > 0. \ v \in (C(\mathbb{R}^d))^d, \ \operatorname{div}(v) = 0. \\ \end{split}$$

 $\Gamma \in (C^1(\mathbb{R}^d))^{d \times d}, \ \Gamma_{i,j} = -\Gamma_{j,i}, \ v_i = \sum_{j=1}^d \partial_j \Gamma_{i,j}.$ 

# Motivations: turbulent flows



#### M. Rutgers

$$\Gamma := \sum_{n=0}^{\infty} \gamma^n \quad E_n \ (\frac{x}{\rho^n}).$$



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$$\Gamma := \sum_{n=0}^{\infty} \gamma^n \left( \underbrace{E_n}_{\rho^n} \right) \left( \frac{x}{\rho^n} \right).$$

Eddies geometrical parameters  $\mathbb{T}^d :=$  Torus of Dimension d and side 1.  $\forall n, E_n \in (C^1(\mathbb{T}^d))^{d \times d}; E_{n;i,j} = -E_{n;j,i}; E_n(0) = 0.$ 

$$\sup_{n \in \mathbb{N}} \sup_{m, i, j \in \{1, \dots, d\}} \|\partial_m E_{n; i, j}\|_{\infty} \le 1$$

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$$\Gamma := \sum_{n=0}^{\infty} (\gamma^n) E_n (\frac{x}{\rho^n}).$$

#### **Circulation rates**

 $\gamma \in \mathbb{R}^{+,*}$ ;



# The Spectrum is not Kolmogorov

- v(l) velocity of the eddies of size l
- $\mathcal{E}(k)$  The kinetic energy distribution in the Fourier modes
- Kolmogorov

$$v(l) \sim l^{\frac{1}{3}} \qquad \mathcal{E}(k) \sim k^{-\frac{5}{3}}$$

Our Model

$$v(l) \sim l^{\frac{\ln \gamma}{\ln \rho} - 1} \qquad \mathcal{E}(k) \sim k^{1 - 2\frac{\ln \gamma}{\ln \rho}}$$

• Kolmogorov  $ightarrow \gamma = 
ho^{\frac{4}{3}}$ 

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Our model  $\gamma < \rho$ 

#### One scale.





Stream lines of the flow  $\gamma^n E_n(\frac{x}{\rho^n})$ 

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Stream lines of  $\gamma^n E_n(\frac{x}{\rho^n})$  and  $\gamma^{n+1} E_{n+1}(\frac{x}{\rho^{n+1}})$ .





Stream lines of  $\gamma^n E_n(\frac{x}{\rho^n})$  +  $\gamma^{n+1} E_{n+1}(\frac{x}{\rho^{n+1}})$ .



 $\gamma^{n} E_{n}(\frac{x}{\rho^{n}}) + \gamma^{n+1} E_{n+1}(\frac{x}{\rho^{n+1}}) + \gamma^{n+2} E_{n+2}(\frac{x}{\rho^{n+2}})$ 

 $\rho^{n+2}$  $o^{n+1}$ sum

 $\frac{\gamma^{n}E_{n}\left(\frac{x}{\rho^{n}}\right)}{\text{H}}\gamma^{n+1}E_{n+1}\left(\frac{x}{\rho^{n+1}}\right)} + \frac{\gamma^{n+2}E_{n+2}\left(\frac{x}{\rho^{n+2}}\right)}{\text{Modification of the ratio }\rho}$ 

 $\rho^{n+2}$  $o^{n+1}$ sum

 $\frac{\gamma^{n}E_{n}\left(\frac{x}{\rho^{n}}\right)}{\text{Modification of the ratio }\rho} + \frac{\gamma^{n+2}E_{n+2}\left(\frac{x}{\rho^{n+2}}\right)}{\gamma^{n+2}E_{n+2}\left(\frac{x}{\rho^{n+2}}\right)}$ 

# The circulation rates $\gamma^k$



 $\gamma^{n} \overline{E_{n}(\frac{x}{\rho^{n}})} + \gamma^{n+1} \overline{E_{n+1}(\frac{x}{\rho^{n+1}})} + \gamma^{n+2} \overline{E_{n+2}(\frac{x}{\rho^{n+2}})}$ 

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# The circulation rates $\gamma^k$



 $\gamma^{n} E_{n}(\frac{x}{\rho^{n}}) + \gamma^{n+1} E_{n+1}(\frac{x}{\rho^{n+1}}) + \gamma^{n+2} E_{n+2}(\frac{x}{\rho^{n+2}})$ 

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# $E_k$ : Geometry of the eddies



 $\gamma^{n} E_{n}\left(\frac{x}{\rho^{n}}\right) + \gamma^{n+1} E_{n+1}\left(\frac{x}{\rho^{n+1}}\right) + \gamma^{n+2} E_{n+2}\left(\frac{x}{\rho^{n+2}}\right)$ 

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# $E_k$ : Geometry of the eddies



 $\gamma^n E_n\left(\frac{x}{\rho^n}\right)$  +  $\gamma^{n+1} E_{n+1}\left(\frac{x}{\rho^{n+1}}\right)$  +  $\gamma^{n+2} E_{n+2}\left(\frac{x}{\rho^{n+2}}\right)$ 

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# A simple example of the multi-scale flow

d=2, for all n,

$$E_n(x_1, x_2) = \begin{pmatrix} 0 & h(x_1, x_2) \\ -h(x_1, x_2) & 0 \end{pmatrix}$$

with  $h(x_1, x_2) := \cos(2\pi x_1) \sin(2\pi x_2)$ 



 $h_0^2(x,y) = \sum_{k=0}^2 \gamma^k h(\frac{x}{\rho^k}, \frac{y}{\rho^k})$ 

#### with

 $\begin{array}{l} \rho=2.9\text{,}\\ \gamma=1.25 \end{array}$ 



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#### Another example

$$h_0^2(x,y) = \sum_{k=0}^2 \gamma^k h(\frac{x}{\rho^k}, \frac{y}{\rho^k})$$

with  $\rho = 3$ ,  $\gamma = 1.1$  and

 $h(x, y) = 2\sin(2\pi x + 3\cos(2\pi y - 3\sin(2\pi x + 1)))$  $\sin(2\pi y + 3\cos(2\pi x - 3\sin(2\pi y + 1)))$ 

# Another example



# Motivations: turbulent flows



# Two opposed descriptions?

 The transport is Superdiffusive, Highly mixing, self-averaging (Kolmogorov 41, Richardson 26, Obukhov 41)



# Two opposed descriptions?

 High density gradients, coherent patterns, sensitive to the geometry of the flow, (Poincaré 08, Landau-Lifshitz 42-85, Ruelle-Takens 71)



# Questions for our model

- The transport is
- Superdiffusive or not?
- Sensitive to the particular geometry of the flow or not?

#### Our results: outline

- We can define a parameter λ<sup>−</sup> ∈ ℝ<sup>+</sup> from the characteristics (γ, (E<sub>n</sub>)<sub>n∈ℕ</sub>) of our model, ↔ inverse of a local Peclet (Reynolds) number
- if  $\lambda^- > 0 \to \text{superdiffusive}$  + highly mixing + self-averaging

## Our results: outline

- We can define a parameter λ<sup>−</sup> ∈ ℝ<sup>+</sup> from the characteristics (γ, (E<sub>n</sub>)<sub>n∈ℕ</sub>) of our model, ↔ inverse of a local Peclet (Reynolds) number
- if  $\lambda^- > 0 \rightarrow$  superdiffusive + highly mixing + self-averaging
- if  $\lambda^- = 0 \rightarrow$  self-averaging collapses, highly sensitive, high gradients

## Results

$$\begin{cases} dy_t = \sqrt{2\kappa} d\omega_t + v(y_t) dt \\ y_0 = x \end{cases}$$

 $\omega$  standard BM in  $\mathbb{R}^d$ .

SDE

## Results

$$\begin{cases} dy_t = \sqrt{2\kappa} d\omega_t + v(y_t) dt \\ y_0 = x \end{cases}$$

#### $\omega$ standard BM in $\mathbb{R}^d$ . Exit Time

SDE

$$\tau(r) := \inf\{t > 0 : |y_t| \ge r\}$$

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#### **Initial Distribution**

$$m_r(dx) := \frac{dx}{\int_{B(0,r)} dx} 1_{B(0,r)}$$

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#### Mean Exit Time

$$\mathbb{E}_{m_r}[\tau(r)] = \frac{1}{\operatorname{Vol}(B(0,r))} \int_{B(0,r)} \mathbb{E}_x[\tau(r)] dx$$

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## **One Point Fast Motion**

Theorem If  $\lambda^- > 0$  then  $\exists C(d, 1/\lambda^-, 1/\ln \gamma) < \infty$  such that for  $\rho > C\gamma$  one has

$$\left|\limsup_{r\to\infty}\frac{1}{\ln r}\ln\left(\mathbb{E}_{m_r}[\tau(r)]\right)<2\right|$$

## **One Point Fast Motion**

#### Theorem For $\lambda^- > 0$ , $\rho > C\gamma$ and $r > \rho$

$$\mathbb{E}_{m_r}[\tau(r)] = r^{2-\nu(r)}$$

$$\nu(r) = \frac{\ln \gamma}{\ln \rho} (1 + \epsilon(r))$$

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Shear flow models: Avellaneda-Majda (91), Glimm-Zhang (92), Gaudron (00), Komorowski-Fannjiang (01), Ben Arous-Owhadi (01). Non shear flow, Kraichnan, Gaussian, annealed, one particle: Piterbarg 97, Komorowski-Olla 02, Fannjiang 02

## Fast Transport as an almost sure event

Fast transport event

$$H(r) := \left\{ \tau(r) \le r^{2-\delta} \right\}$$

with

$$\delta = 0.9 \frac{\ln \gamma}{\ln \rho}$$

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Observe that  $\delta > 0$ 

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Theorem If  $\lambda^- > 0$  then for  $\rho > C\gamma$  one has

$$\lim_{r \to \infty} \mathbb{P}_{m_r} \big[ H(r) \big] = 1$$

# Super-Diffusive two-points motion

 $z_t$  second passive tracer

$$dz_t = \sqrt{2\kappa} \, d\bar{\omega}_t + v(z_t) \, dt.$$

 $\bar{\omega}_t$  standard BM independent of  $\omega_t$ .

$$\begin{split} B(0,r,l) := & \{(y,z) \in \mathbb{R}^d \times \mathbb{R}^d \, : \, |y-z| < r \\ & \text{and} \quad y^2 + z^2 < l^2 \} \end{split}$$

 $\tau(r,l) := \inf\{t > 0 : (y_t, z_t) \notin B(0, r, l)\}$ 

# Super-Diffusive two-points motion

$$m_{r,l}(dy\,dz) := \frac{dy\,dz}{\int_{(y,z)\in B(0,r,l)} dy\,dz} \mathbf{1}_{B(0,r,l)}.$$

Theorem If  $\lambda^- > 0$  then for  $\rho > C\gamma$  one has

$$\limsup_{r \to \infty} \lim_{l \to \infty} \frac{1}{\ln r} \ln \left( \mathbb{E}_{m_{r,l}} [\tau(r, l)] \right) < 2$$

$$\lim_{l \to \infty} \mathbb{E}_{m_{r,l}} \left[ \tau(r, l) \right] = r^{2 - \nu(r)}$$

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Theorem If  $\lambda^- > 0$  then for  $\rho > C\gamma$  one has

$$\lim_{r \to \infty} \lim_{l \to \infty} \mathbb{P}_{m_{r,l}} \{ \tau(r,l) \le r^{2-\delta} \} = 1$$

#### Strong Self Averaging property of the flow

#### $\text{If }\lambda^->0$

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#### $\text{If }\lambda^->0$

- The particular geometry of the eddies  $E_n$  has no influence on the transport
- The transport depends only on the power law  $\frac{\ln\gamma}{\ln\rho}$  of the velocity field.

• Renormalization procedure.

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- S(T<sup>d</sup>) := {skew symmetric d × d matrices with coefficients in L<sup>∞</sup>(T<sup>d</sup>)}.

Let  $(a, E) \in \mathcal{M} \times \mathcal{S}(\mathbb{T}^d)$ 

Operator

$$\begin{cases} \operatorname{div} \left( a + E(\frac{x}{\epsilon}) \right) \nabla u_{\epsilon}(x) = f(x) & \text{in } \Omega \\ u_{\epsilon} = 0 & \text{in } \partial \Omega \end{cases}$$

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then  $\exists \sigma(a, E) \in \mathcal{M}$  such that as  $\epsilon \downarrow 0$ ,

$$u_{\epsilon} \rightarrow u_0$$

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$$u_{\epsilon} \rightarrow u_0$$

 $\begin{cases} \operatorname{div} \boldsymbol{\sigma}(a, E) \nabla u_0 = f & \text{in } \Omega \\ u_0 = 0 & \text{in } \partial \Omega \end{cases}$ 

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#### SDE

 $dx_t = 2^{\frac{1}{2}} a^{\frac{1}{2}} d\omega_t + \operatorname{div} E(x_t) dt,$ 

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As  $\epsilon \downarrow 0$ ,

SDE

$$\epsilon x_{t/\epsilon^2} \to B_t$$

 $B_t$  Brownian Motion with covariance matrix  $2\sigma(a, E)$ .

#### Literature.

- $2\sigma(a, E)$ : Effective diffusivity
- $\sigma(a, E)$ : Effective conductivity Eddy viscosity. Dispersion matrix.

#### Effective conductivity as a mapping

$$\sigma : \mathcal{M} \times \mathcal{S}(\mathbb{T}^d) \to \mathcal{M}$$
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Effective conductivity as a mapping

$$\sigma : \mathcal{M} \times \mathcal{S}(\mathbb{T}^d) \to \mathcal{M}$$
$$(a, E) \to \sigma(a, E)$$

for all  $(a, E) \in \mathcal{M} \times \mathcal{S}(\mathbb{T}^d)$ 

$$a \le \sigma(a, E) \le a + \int_{\mathbb{T}^d} {}^t E(x) a^{-1} E(x) dx$$

Renormalization sequence  $(A_n)_{n \in \mathbb{N}}$ For all  $n \in \mathbb{N}$ ,  $A_n \in \mathcal{M}$ 

$$egin{array}{ccc} A_0 = rac{\kappa}{\gamma} I_d & ext{and} & A_{n+1} = rac{1}{\gamma} \sigma(A_n, E_n) \end{array}$$

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 $(A_n)_{n\in\mathbb{N}}$  does not depend on  $\rho$ 

$$\lambda^{-} := \liminf_{n \to \infty} \lambda_{\min}(A_n)$$

$$\Gamma^{n-1}(x) := \sum_{k=0}^{n-1} \gamma^k E_k(\frac{x}{\rho^k})$$

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Assume  $\rho \in \mathbb{N}$ . Then  $\Gamma^{n-1}$  periodic.

 $\operatorname{div}(\kappa I_d + \Gamma^{n-1}) \nabla \xrightarrow{Homogen} \operatorname{div} \sigma(\kappa I_d, \Gamma^{n-1}) \nabla$ 

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 $\operatorname{div}(\kappa I_d + \Gamma^{n-1}) \nabla \xrightarrow{Homogen} \operatorname{div} \sigma(\kappa I_d, \Gamma^{n-1}) \nabla$ 

$$\lim_{\rho \to \infty} \sigma(\kappa I_d, \Gamma^{n-1}) = \gamma^n A_n$$

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Magnitude of the velocity vector field at scale n:

 $V_n \sim \frac{\gamma^n}{\rho^n}$ 

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Local Peclet tensor

$$\mathbf{Pe}^{n} := V_{n} \rho^{n} \big( \lim_{\rho \to \infty} \sigma(\kappa I_{d}, \Gamma^{n-1}) \big)^{-1}$$

Magnitude of the velocity vector field at scale *n*:

$$V_n \sim \frac{\gamma^n}{\rho^n}$$

Local Reynolds tensor

$$\mathbf{Re}^{n} := V_{n} \rho^{n} \left( \lim_{\rho \to \infty} \sigma(\kappa I_{d}, \Gamma^{n-1}) \right)^{-1}$$

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Local Reynolds tensor

$$\mathbf{Re}^{n} := V_{n} \rho^{n} \big( \lim_{\rho \to \infty} \sigma(\kappa I_{d}, \Gamma^{n-1}) \big)^{-1}$$

$$A_n = (\mathbf{P}\mathbf{e}^n)^{-1} = (\mathbf{R}\mathbf{e}^n)^{-1}$$

$$\lambda^+ := \limsup_{n \to \infty} \lambda_{\max}(A_n)$$

$$\mu := \limsup_{n \to \infty} \frac{\lambda_{\max}(A_n)}{\lambda_{\min}(A_n)}$$

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Theorem

$$\lambda^+ \leq rac{C_d}{\lambda^-} \quad ext{and} \quad \mu \leq rac{C_d}{(\lambda^-)^2}$$

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$$\mu := \limsup_{n \to \infty} \frac{\lambda_{\max}(A_n)}{\lambda_{\min}(A_n)}$$

Theorem

$$egin{array}{ccc} \lambda^+ \leq rac{C_d}{\lambda^-} & ext{and} & \mu \leq rac{C_d}{(\lambda^-)^2} \end{array} \end{array}$$

 $\lambda^- > 0 \Rightarrow \lambda^+ < \infty$  and  $\mu < \infty$ 

#### for $\zeta > 0$

$$V(\zeta) := \liminf_{n \to \infty} \frac{\lambda_{\min}(\sigma(\zeta I_d, E_n))}{\zeta}$$
# When is $\lambda^- > 0$ ?

for  $\zeta > 0$ 

$$V(\zeta) := \liminf_{n \to \infty} \frac{\lambda_{\min}(\sigma(\zeta I_d, E_n))}{\zeta}$$

V is decreasing and  $V \ge 1$  thus

$$V(0) := \lim_{\zeta \downarrow 0} V(\zeta)$$

is well defined.

#### When is $\lambda^- > 0$ ?

for  $\zeta > 0$ 

$$V(\zeta) := \liminf_{n \to \infty} \frac{\lambda_{\min}(\sigma(\zeta I_d, E_n))}{\zeta}$$

Theorem If  $\mu < \infty$  and  $\gamma < V(0)$  then  $\lambda^- > 0$  and

$$C_1 \le \lambda^- \lambda^+ \le C_2.$$

# Idea of the Proof.

#### • Ball B(0,r)



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- Ball B(0,r)
- We want to compute  $\mathbb{E}[\tau(0,r)]$

#### Idea of the Proof.



- Ball B(0,r)
- We want to compute  $\mathbb{E}[\tau(0,r)]$

• Main feature of the flow: Infinite number of scales  $0, 1, \ldots, \infty$ 

• Scale  $n(r) = [\ln r / \ln \rho]$ 

 $\rho^{n(r)} \le r < \rho^{n(r)+1}$ 





• Scale  $n(r) = [\ln r / \ln \rho]$ 



• Small scales  $0, \ldots, n(r) - 1$ 



• Scale  $n(r) = [\ln r / \ln \rho]$ 



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- Small scales  $0, \ldots, n(r) 1$
- Intermediate scales n(r), n(r) + 1

• Large scales  $n(r) + 2, \dots, \infty$ 

# All Scales.



• Medium with infinite number of scales  $0, 1, \ldots, \infty$ .

# All Scales.



- Medium with infinite number of scales  $0, 1, \ldots, \infty$ .
- Transport of a drop of dye ?

# Transport by Large Scales



#### Large scales

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# Transport by Large Scales



Large scales

 Their influence on the transport of the drop of dye is

# Transport by Large Scales



Large scales

 Their influence on the transport of the drop of dye is negligible

# Transport by Small Scales



# • Small Scales $\rightarrow$ homogenized.

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# Transport by Small Scales



 Transport = Diffusion with Effective diffusivity

 $\sigma(\kappa I_d, \Gamma^{n(r)}) \sim \gamma^{n(r)} A_{n(r)}$ 

• Exit Time:

$$au_D(0,r) \sim rac{r^2}{\gamma^{n(r)}\lambda(A_{n(r)})}$$

# Transport by Small Scales



 Transport by mixing, density gradients smoothed.

# Transport by Intermediate Scales



 Intermediate Scales: not homogenized, not negligible.

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# **Transport by Intermediate Scales**



- Transport by convection through particular geometry.
- Exit time



# Transport by Intermediate Scales



 Transport by advection, density gradients increased.

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Local Peclet number

$$\mathbf{Pe}(r) := \frac{\tau_C(r)}{\tau_D(r)} \sim \left(\lambda(A_{n(r)})\right)^{-1}$$

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 $\lambda^- > 0 \Rightarrow \mathbf{Pe}(r) < \infty \Rightarrow \text{At every scale } r$ , advection (irregularities, high gradients) is compensated by averaging (smoothing, dissipating).

Local Peclet number

$$\mathbf{Pe}(r) := \frac{\tau_C(r)}{\tau_D(r)} \sim \left(\lambda(A_{n(r)})\right)^{-1}$$

 $\lambda^- > 0 \Rightarrow \mathbf{Pe}(r) < \infty \Rightarrow$  influence of the intermediate scales on the transport comparable to the influence of the small scales.

$$au(0,r) \sim au_D(0,r) \sim rac{r^2}{\gamma^{n(r)}}$$

Local Peclet number

$$\mathbf{Pe}(r) := \frac{\tau_C(r)}{\tau_D(r)} \sim \left(\lambda(A_{n(r)})\right)^{-1}$$



$$\tau(0,r) \sim \frac{r^2}{\gamma^{n(r)}} \sim r^{2-\nu}$$

with 
$$\nu = \frac{\ln \gamma}{\ln \rho} > 0$$

#### As $r \to \infty$

 $\mathbf{Pe}(r) \to \infty$ 

# $\overline{\mathrm{If}\,\lambda^{-}}=0$

As  $r \to \infty$ 

 $\mathbf{Pe}(r) \to \infty$ 

At every scale transport dominated by advection.

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 $\mathbf{Pe}(r) \to \infty$ 

- At every scale transport dominated by advection.
- Collapse of the self-averaging property of the flow towards chaos.

As  $r \to \infty$ 

 $\mathbf{Pe}(r) \to \infty$ 

- At every scale transport dominated by advection.
- Collapse of the self-averaging property of the flow towards chaos.
- The particular geometry of the eddies can not be neglected even if  $\rho$  is large.

Definition  $A_n$  is self-similar and isotropic iff

•  $orall n \in \mathbb{N}$ ,  $E_n = E$  ; .

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$$\forall n \in \mathbb{N}, E_n = E$$
 ;

• 
$$\forall \zeta > 0$$
,  $\sigma(\zeta I_d, E) = \lambda(\zeta) I_d$ .

Definition

 $A_n$  is self-similar and isotropic iff

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• 
$$\forall \zeta > 0$$
,  $\sigma(\zeta I_d, E) = \lambda(\zeta) I_d$ .

If  $A_n$  is self-similar then it is a low order dynamical system.

$$A_{n+1} = \frac{1}{\gamma}\sigma(A_n, E)$$

Definition  $A_n$  is self-similar and isotropic iff

• 
$$\forall n \in \mathbb{N}, E_n = E$$
 ;

•  $\forall \zeta > 0$ ,  $\sigma(\zeta I_d, E) = \lambda(\zeta) I_d$ .

$$V(0) := \lim_{\zeta \downarrow 0} \frac{\lambda_{\min}(\sigma(\zeta I_d, E))}{\zeta}$$

**Theorem If**  $A_n$  is self-similar and isotropic then

• If 
$$\gamma < V(0)$$
 then  $\lambda^- > 0$  and  
 $\lim_{n\to\infty} A_n = \zeta_0 I_d$  where  $\zeta_0$  is the unique solution of  $V(\zeta_0) = \gamma$ .

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• If  $\gamma = V(0)$  and  $(V(0) - V(x))x^{-p}$  admits a non ero limit as  $x \downarrow 0$  with p > 0 then  $\lambda^{-} = 0$  and  $\lim_{n\to\infty} \frac{\ln \lambda(A_n)}{\ln n} = -\frac{1}{p}$ .

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- If  $\gamma > V(0)$  then  $\lambda^- = 0$  and  $\lim_{n \to \infty} \frac{1}{n} \ln \lambda(A_n) = \ln \left(\frac{V(0)}{\gamma}\right)$
# Bifurcation



- Flow self similar and isotropic. for all n,  $E_n = E$ .
- Shape of the eddy *E* over a period.
- $1 < V(0) < \infty$ .

# Bifurcation



- $\gamma < V(0)$
- $\lambda^- > 0$
- The flow is self averaging

# Bifurcation



- $\gamma \ge V(0)$
- $\lambda^- = 0$

• The self averaging property collapses.

Assume  $A_n$  to be self-similar and isotropic. Spatial scales  $\rho^n \to R_n$ 

$$\Gamma(x) = \sum_{n=0}^{\infty} \gamma^n E(\frac{x}{R_n})$$

$$\rho_{\min} := \inf_{n \in \mathbb{N}} \frac{R_{n+1}}{R_n}$$

 $2 \leq \rho_{\min}$ 

### For $y \in [0,1]^d$

$$\sigma(n, y) := \sigma(\kappa I_d, \Gamma^{n-1} + \gamma^n E(y + \frac{x}{R_n}))$$

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- Averaging paradigm  $\Rightarrow$  Relative translation by y has little influence on  $\sigma(n, y)$
- for all y,

$$\lim_{\rho_{\min} \to \infty} \sigma(n, y) = \lim_{\rho_{\min} \to \infty} \sigma(n, 0)$$

$$\sigma(n, y) := \sigma(\kappa I_d, \Gamma^{n-1} + \gamma^n E(y + \frac{x}{R_n}))$$

$$\sigma(n, y, \rho) := \lim_{\frac{R_1}{R_0}, \dots, \frac{R_{n-1}}{R_{n-2}} \to \infty; \frac{R_n}{R_{n-1}} = \rho} \sigma(n, y)$$

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 $\sigma(n, y, \rho) = \gamma^{n-1} \sigma(A_{n-1}, E(\rho x) + \gamma E(x+y))$ If  $\lambda^- > 0$  then  $l \in (R^d)^*$  and  $y \in [0, 1]^d$ ,

$$\limsup_{n \to \infty} \frac{{}^{t} l \sigma(n, y, \rho) l}{{}^{t} l \sigma(n, 0, \rho) l} < 1 + C_d(\rho \lambda^{-})^{-\frac{1}{2}}$$

$$\sigma(n, y) := \sigma(\kappa I_d, \Gamma^{n-1} + \gamma^n E(y + \frac{x}{R_n}))$$

$$\sigma(n, y, \rho) := \lim_{\frac{R_1}{R_0}, \dots, \frac{R_{n-1}}{R_{n-2}} \to \infty; \frac{R_n}{R_{n-1}} = \rho} \sigma(n, y)$$

 $\sigma(n, y, \rho) = \gamma^{n-1} \sigma(A_{n-1}, E(\rho x) + \gamma E(x+y))$ If  $\lambda^- = 0$  then for any  $\rho > 1$  there exists E and  $y \in [0, 1]^d$  such that for any  $l \in (R^d)^*$ ,

$$\lim_{n \to \infty} \frac{{}^{t} l \sigma(n, y, \rho) l}{{}^{t} l \sigma(n, 0, \rho) l} = \infty.$$

# Two scale flows



 $S_{\rho}E(x) = E(\rho x)$  $\Theta_y E(x) = E(x - y)$ As  $\zeta \downarrow 0$ 

 $\sigma(\zeta I_d, S_\rho E + E) \sim C_1 \zeta I_d.$ 

# Two scale flows



$$S_{\rho}E(x) = E(\rho x)$$
  

$$\Theta_{y}E(x) = E(x - y)$$
  
As  $\zeta \downarrow 0$ 

 $\sigma(\zeta I_d, S_\rho E + \Theta_y E) \sim C_2 \zeta^{\frac{1}{2}} I_d$ 



The flow at scale r is laminar. Viscosity  $\kappa$ , velocity  $V(r) = V_0$ .

# 

A small perturbation is introduced.

Averaging vs Chaos in Turbulent Transport? - p. 53/



 $\tau_D(r)$ : exit time of the perturbation by diffusion  $\tau_C(r)$ : exit time of the perturbation by convection



Averaging vs Chaos in Turbulent Transport? – p. 54/



If  $\tau_D(r) < \tau_C(r)$  the perturbation exits by diffusion and is smoothed before going out of B(0, r).



### $\Rightarrow$ the laminar flow is stable at the scale r



$$\tau_D(r) < \tau_C(r) \Leftrightarrow \frac{r^2}{\kappa} < \frac{r}{V_0} \Leftrightarrow \mathbf{Re} = \frac{rV_0}{\kappa} < 1$$

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If  $\tau_D(r) > \tau_C(r)$  the perturbation exits by convection and propagates.



 $\Rightarrow$  The flow is unstable at the scale r and fluctuates and this scale.



$$\tau_D(r) > \tau_C(r) \Leftrightarrow \frac{r^2}{\kappa} > \frac{r}{V_0} \Leftrightarrow \mathbf{Re} = \frac{rV_0}{\kappa} > 1$$

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change of scale

Let's look at the flow at the scale  $r/\rho$ 

The flow is laminar at this scale  $r/\rho$  with velocity  $V(r/\rho) = V_0/\gamma$ 

change of scale

A small perturbation is introduced at the scale  $r/\rho$ 

change of scale



This self similar process is iterated, until the dissipation scale l is reached.



At the dissipation scale l ,  $au_C(l) \sim au_D(l)$ 

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 $\sigma(r) \sim rV(r)$ 

The multi-scale structure creates an effective viscosity  $\sigma(r) \sim rV(r)$ 

Averaging vs Chaos in Turbulent Transport? – p. 60/



 $\sigma(r) \sim rV(r)$  $\tau_C(r) \sim r/V(r)$  $au_D(r) \sim r^2 / \sigma(r)$ 

The multi-scale structure stabilizes the flow since  $\sigma(r) \sim rV(r) \Rightarrow \tau_D(r) \sim \tau_C(r)$ 



 $V(x) = V_0 \overline{(x/r)^{\alpha}}$ 

We write  $\gamma = \rho^{1+\alpha}$ 

Averaging vs Chaos in Turbulent Transport? – p. 61/0



$$V(x) = V_0 (x/r)^{\alpha}$$
$$\epsilon(x) \sim \frac{(V(x))^3}{x}$$

Kolmogorov  $\alpha = 1/3$ 

The energy dissipation  $\epsilon(x)$  at scale x is  $\sigma(x)(\frac{V(x)}{x})^2 \sim \frac{(V(x))^3}{x}$ 



$$V(x) = V_0 (x/r)^{\alpha}$$
$$\epsilon(x) \sim \frac{(V(x))^3}{x}$$

Kolmogorov  $\alpha = 1/3$ 

If  $\alpha > 1/3$  the larger eddies are dissipated before the smaller ones



$$V(x) = V_0 (x/r)^{\alpha}$$
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Kolmogorov  $\alpha = 1/3$ 

If  $\alpha < 1/3$  the smaller eddies are dissipated before the larger ones



$$V(x) = V_0 (x/r)^{\alpha}$$
$$\epsilon(x) \sim \frac{(V(x))^3}{x}$$

Kolmogorov  $\alpha = 1/3$ 

The relation  $\sigma(x) \sim x V(x)$  is at the core of the Kolmogorov law



$$V(x) = V_0 (x/r)^{\alpha}$$
$$\epsilon(x) \sim \frac{(V(x))^3}{x}$$

Kolmogorov  $\alpha = 1/3$ 

In the anisotropic case, the relation  $\lambda_{\max}(\sigma(x))\lambda_{\min}(\sigma(x)) \sim x^2(V(x))^2$  restores the isotropy of the flow.

Averaging vs Chaos in Turbulent Transport? - p. 61/



$$V(x) = V_0 (x/r)^{\alpha}$$
$$\epsilon(x) \sim \frac{(V(x))^3}{x}$$

Kolmogorov  $\alpha = 1/3$ 

At the dissipation scale,  $\tau_C(l) \sim \tau_D(l) \Leftrightarrow l/V(l) \sim l^2/\kappa$  $\Leftrightarrow r/l \sim (\frac{V_0 r}{\kappa})^{\frac{3}{4}}$ 

Averaging vs Chaos in Turbulent Transport? – p. 61/6
## On the nature of Turbulence.



$$V(x) = V_0 (x/r)^{\alpha}$$
$$\epsilon(x) \sim \frac{(V(x))^3}{x}$$

Kolmogorov  $\alpha = 1/3$ 

## intermittency at the smaller scales