

Course 2: (Data-driven) turbulence

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or "a Gaussian representation of NS turbulence"

① Steady-state NS turbulence. (A Hell of a dissipative machine)

- Turbulent measure.
- Observations.

② Random functions - Short survey.

- Random variables, Law, Equivalence in Law
- Independence
- Processes - FDD - Equivalence in Laws
- Gaussian process.
- Stationary (Homogeneous) processes.

③ Gaussian HIT representation of NS turbulence

- Two-point correlator.
- Increments, power law, self-similarity

④ NS vs HIT: Kolmogorov-Frisch scenario.

- Statistical restoring geometry.
- Mono scaling geometry.

⑤ $\nu > 0$: probing HIT with NS.

- Running Reynolds number
- Inertial range.

⑥ Openings: Gaussian HIT faithful

① Steady-state (NS) turbulence.

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• The NS equation provides an efficient way of generating "turbulent state" as a limit $t \rightarrow \infty$ $D \rightarrow \infty$ of physically reasonable solutions.

• "Turbulent state": ~~stationary long time average~~

φ observable of the space of suitable velocity field.

$$\langle \varphi[v] \rangle_{t,D} = \frac{1}{T} \int_0^T \varphi[v(s)] ds.$$

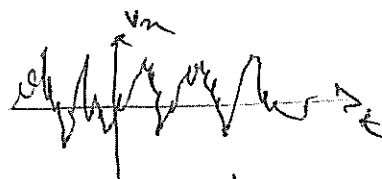
"Turbulent state" = $\lim_{D \rightarrow \infty} \lim_{t \rightarrow \infty} \langle \varphi \rangle_{t,D} = \langle \varphi \rangle_*$

with the proviso: $\langle \varepsilon \rangle = \langle \dot{I} \rangle$ $I = \langle v^2 \rangle$

• Numerical observations from triple periodic numerical simulations.

For sufficiently high Reynolds, the velocity field behaves as a random function

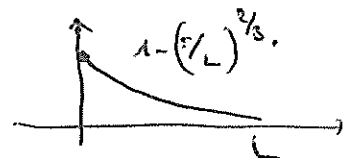
(i) stationary.



(ii) Power-law correlations on a certain range of scale.

$$\langle C(r) \rangle \sim \left[1 - \left(\frac{r}{L} \right)^{2/3} \right]$$

$y \ll r \ll L$



(iii) Finite-variance: $\langle v^2 \rangle \rightarrow \langle v_{rms}^2 \rangle > 0$.

(iv) Finite-dissipation: $\langle \varepsilon \rangle \rightarrow \varepsilon_* > 0$.

OBS: (iii) & (iv) are non-trivial!

D = number of the "turbulent state"

② Random functions, Vocabulary

Jack. Koiv. Probability
 Examples, Jack Koiv. Applications

Probability Spaces:

(Ω, \mathcal{F}, P)
 Ω : outcomes / realization
 \mathcal{F} : events
 P : Probability

\mathcal{F} σ -algebra, $P: \mathcal{F} \rightarrow \mathbb{R}^+$
 $P[\Omega] = 1$ $P[\emptyset] = 0$
 $P[\cup_{i \in \mathbb{N}} A_i] = \sum_{i \in \mathbb{N}} P[A_i]$ (Mutually disjoint)
 ~~$P[\mathbb{R}] = 1$~~

Event with probability 1: "almost sure"

Random variable:

$X: \Omega \rightarrow \mathbb{R}^m$ such that $\forall I$ interval of \mathbb{R}^m $X^{-1}(I) \in \mathcal{F}$

$I \subseteq \mathbb{R}^m$
 $P_X [I] = P[X(\omega) \in I] = P[X^{-1}(I)]$

P_X : pushforward of P by X , sometimes written $X\#P$.

Averages:

Let $\varphi[X]$ a functional of X [observable].

Then $\langle \varphi[X] \rangle_P = \int_{\Omega} \varphi[X(\omega)] P(d\omega)$

$P_X = X\#P$ "pushforward of P by X "

$\langle \varphi \rangle_{P_X} = \int_{\mathbb{R}^m} \varphi[x] P_X[dx]$

$\langle \varphi[X] \rangle_P = \int_{\mathbb{R}^m} \varphi[x] (X\#P)(dx)$

$\langle \varphi \rangle_{X\#P} = \langle 1 \rangle_{\varphi \circ X\#P}$

Notations: $P[X \in dx] = P[X \in]x, x+dx[$, $m=1$


or: $P[X_1 \in]x_1+dx_1[, \dots, X_m \in]x_m+dx_m[$

Examples:

Average: $P_X = \langle X \rangle$
 Correlator: $\langle Y; X \rangle = \langle X \rangle \langle X \rangle$
 Characteristic function: $\varphi(x) = \langle e^{it \cdot x} \rangle$, etc.

Density: $f(x_1, \dots, x_n) = \lim_{dx_1, \dots, dx_n \rightarrow 0} \frac{P[X \in]x_i; x_i + dx_i[]}{dx_1 \dots dx_n}$

Equality in law: $P_X = P_Y \iff X \stackrel{L}{=} Y$

- Example:
- Discrete density: $f(x) = \frac{1}{2} \delta(x) + \delta(x-1) \frac{1}{2}$ 
 - Gaussian density 1D: $f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$
 $X \sim \mathcal{N}(\mu, \sigma^2)$
 - nD: $f(\vec{x}) = \frac{1}{[(2\pi)^n \det C]^{1/2}} e^{-\frac{1}{2}(\vec{x}-\mu)^T C^{-1}(\vec{x}-\mu)}$
 \hookrightarrow symmetrisch u.a. negativ definit.
 - Characteristic function of

1D gaussian variable $X \sim \mathcal{N}(0, \sigma^2)$

$\varphi(t) = \langle e^{-itx} \rangle = e^{-\frac{t^2}{2} \sigma_x^2}$

nD $\varphi(t) = e^{-\frac{t_i t_j}{2} \langle x_i x_j \rangle}$ $X \sim \mathcal{N}(0, C)$

Conditional probability: X, Y two (\mathbb{R}) Random variables.

$P_{X|Y} = \frac{P_{(X,Y)} [X \in]x; x+dx[; Y \in]y; y+dy[]}{P[Y \in]y; y+dy[]}$

Notations: $P_{X|Y} [dx|y] = \frac{P_X [dx, dy]}{P_Y [dy]}$] Not needed

Independence: $P_{X|Y} [dx|y] = P_X [dx]$ $\iff X \perp\!\!\!\perp Y$

$P_{X,Y} [dx, dy] = P_X [dx] P_Y [dy]$

• Stochastic process

collection of random variables parametrized by

continuous parameter ex: $(X(t, \omega), t \in \mathbb{R})$

$(\mu(t, \vec{x}, \omega), t \in \mathbb{R}, \vec{x} \in \mathbb{R}^3)$

For fixed realization $\omega \in \Omega$, $t \mapsto \mu(t)$ is the "path".

• Equivalence in Law

$$(X_t, t \in \mathbb{R}) \stackrel{L}{\sim} (Y_t, t \in \mathbb{R})$$

iff

$$FDD(X) = FDD(Y)$$

with $FDD(X) = \left\{ P_{X_{t_1}, \dots, X_{t_n}}, t_i \in \mathbb{R}, n \in \mathbb{N} \right\}$

• Gaussian process

$(X_t, t \in \mathbb{R})$ is gaussian

iff

$\forall t_1, \dots, t_n \in \mathbb{R} \exists \mu, C_k$ (non-negative definite)

$$(X_{t_1}, \dots, X_{t_n}) \sim CN(\mu, C_n)$$

OBS gaussian process is fully specified by

the knowledge of $C(t_1, t_2), (t_1, t_2) \in \mathbb{R}, \mu(t), t \in \mathbb{R}$.

How
(Stationary) process.

A process whose finite-dimensional distributions are invariant under arbitrary time shift of index points.

$$X_{t_1} \dots X_{t_n} \stackrel{L}{=} X_{t_1+z} \dots X_{t_n+z}, \quad z \in \mathbb{R}.$$

or
$$a_z : X_t \rightarrow X_{t+z}.$$

$$\underbrace{X_{t_1, \dots, t_n}}_X \stackrel{L}{=} a_z[X]$$

obs: t "no role" $P_X = P_{a_z \circ X}$
 \rightarrow stationary.

Observation: For stationary process, the average is constant over time and the covariances / second order structure function only depend on the time lag.

$$\langle X(t_1) \rangle = \langle X(0) \rangle = \mu_0.$$

$$\begin{aligned} \langle X(t_1) X(t_2) \rangle &= \langle X(t_1 - t_2) X(0) \rangle = c(t_1 - t_2) \\ &= c(t_2 - t_1). \end{aligned}$$

$C: \mathbb{Z} \rightarrow \mathbb{R}$: Correlation function of the process.

OBS.

• Continuity of sample path is guaranteed by continuity at 0 for the sample function.

• Ergodic theorem holds for C that decays fast enough.

Ex: if $\int_0^{\infty} |c| < \infty$ then.

$$\mathbb{E} \left| \frac{1}{T} \int_0^T X_t - \mu \right|^2 \xrightarrow{T \rightarrow \infty} 0.$$

For stationary process, we may identify long-run average with ensemble averages!

• Homogeneity: stationarity when the ^{index} parameter is "space".

③ Gaussian representation of turbulence.

A simple representation of the spatial structure of a gaussian field is to prescribe it as a gaussian random field, homogeneous in space,

fully prescribe by the two-point correlator:



$$\langle v_i(x) v_j(x') \rangle = \langle v_i(0) v_j(r) \rangle, \quad i, j \in \{1, 2, 3\}$$

$$= C_{ij}(r)$$

with C_{ij} prescribed by:

$$C_{ij}(r) = \frac{f(r) - g(r)}{r^2} r_i r_j + g(r) \delta_{ij}$$

with $f(r) = U_0^2 \left[1 - \left(\frac{r}{L} \right)^{2/3} \right] \cdot \mathbb{1}_{r \leq L}$

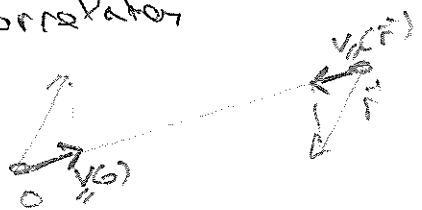
$$g(r) = f(r) + \frac{r}{2} f'(r)$$

$$= U_0^2 \left[1 - \left(\frac{4}{3} \right) \left(\frac{r}{L} \right)^{2/3} \right] \cdot \mathbb{1}_{r \leq L}$$

Piphen.

• $f(r)$ represents a parallel correlator

$$f(r) = \langle v_{\parallel}(0) v_{\parallel}(r) \rangle$$
$$= \hat{r}_i C_{ij} \hat{r}_j$$



• $g(r)$ — a \perp correlator.

$$g(r) = \hat{m}_i C_{ij} \hat{m}_j$$

• The relation between f and g is mediated through incompressibility $\partial_i C_{ij} = 0$.

• Beyond homogeneity, the gaussian vector field is isotropic, i.e. the distribution is invariant under arbitrary $O \in SO_3(\mathbb{R})$ and parity invariant!

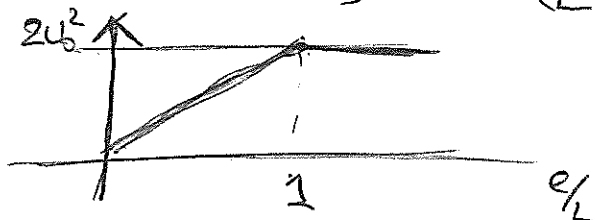
in the sense

$$O_{ie} O_{je} C_{ee'} [O^{-1} \hat{r}] = C(\hat{r})$$

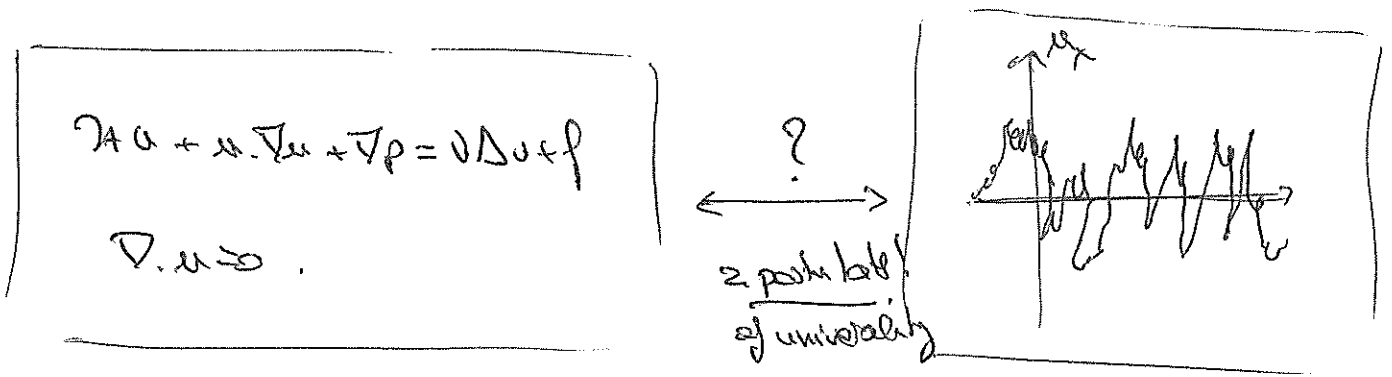
$$O \in (SO_3(\mathbb{R}))$$

• It produces scaling laws in the form

$$\langle \delta v_{\parallel}^2 \rangle = 2[C(0) - C(r)] = 2v_0^2 \left(\frac{r}{L}\right)^{2/3}$$



④ NS-HIT: Kolmogorov Frisch-Sovano



2

A) Statistical restoring of symmetries ["universality of statistics"]

The NS symmetries broken by the presence of forcing + viscosity are restored at the level of the turbulent measure for "small enough scales"

In other words:

In the limit $t \rightarrow \infty, \nu \rightarrow 0, l \rightarrow 0$
 turbulent measure reflect a stochastic process
 $u(\vec{x}, t)$ such that:

- T T $u(x, t) \stackrel{L}{=} u(x, t + \tau)$
- H Hom. $u(x + \rho, 0) \stackrel{L}{=} u(x, 0)$
- I Rot. $u(\vec{x}, 0) \stackrel{L}{=} O^{-1} u(O \vec{x}, 0)$
- S_q. $u(x, t) \stackrel{L}{=} \lambda^{-h} u(\lambda x, \lambda^{1-h} t)$
- G $u(x + \vec{u} t, t) \stackrel{L}{=} u(x + \vec{u} t, t) - u_0$

obs: HIT = OK.

• G and δu are realized at the level of increments.

ie \Rightarrow

$$* \delta u(\ell) \sim \mathcal{N}(0, \sigma_\ell^2) \quad \sigma_\ell^2 = 2U_0^2 \left(\frac{\ell}{L}\right)^{2/3}$$

$$\sim \left(\frac{\ell}{L}\right)^{2/3} \delta u(L)$$

$$\frac{2}{\delta u(\ell)} \sim \mathcal{N}(0, \sigma_\ell^2) \sim \left(\frac{\ell}{L}\right)^{2/3} \underbrace{\mathcal{N}(0, \sigma_L^2)}_{\sim \left(\frac{\ell}{L}\right)^{2/3} \delta u(L)}$$

$$L = \lambda \ell \quad , \text{ with } \lambda = \frac{L}{\ell}$$

$$h = \frac{1}{3}$$

$$* \delta u(\ell) = u(\vec{\ell}) - u(0) \stackrel{L}{=} u(\vec{\ell} + U_0 t) - u(U_0 t)$$

↑
(statistics)

\Rightarrow Gaussian HIT realizes the statistical testing of universality with $h = \frac{1}{3}$.

ⓑ. Statistical Scaling universality.

Dimensional analysis fully determine the value of h along with statistical properties of $\delta u(\ell)$.

$$\delta u(\ell) = C_K \underbrace{\varepsilon^{2/3}}_{\substack{\uparrow \\ \text{Gaussian}}} \underbrace{\rho^{2/3}}_{\substack{\uparrow \\ L^2 T^{-3}}} \underbrace{\ell^{2/3}}_{\substack{\uparrow \\ L^{2/3}}} \quad \rightarrow \quad \langle u(\ell) u(\ell') \rangle = U_0^2 - \frac{1}{2} C_K^2 \varepsilon^{2/3} \rho^{2/3} \ell^{2/3} \ell'^{2/3}$$

↑
"trans. field" δu

ie $\frac{U_0^2}{1/2} = \varepsilon^{2/3} \rho^{2/3} \Rightarrow U_0^2 = \varepsilon^{2/3} \rho^{2/3} L^{2/3}$

⑤ $\nu > 0$.

L''

Can we probe HIT with NS at finite ν ? $L=1, F=1$

Let us define $Re_e = \frac{S_2^{1/2}(\ell) \ell}{\nu}$. (Rayleigh number)

we define the inertial range of scale as:

$$1 \ll Re_e \ll Re_L.$$

$$\gamma \ll \ell \ll L. \quad \text{with } \gamma = \left(\frac{\nu^3}{\ell^3} \right)^{1/4}.$$

Power laws appear generically for $\gamma \ll \ell \ll L$.

Rationale: Rescaling of the NS:

$$2u' + u' \cdot \nabla' u' + (\nabla' p)' = \nu_e \partial_{xx}^2 u' + f_e.$$

$$\text{with } \nu_e = \frac{\nu}{Re_e} \quad f_e = \frac{\ell}{\ell_e^2} F = \frac{\ell}{S_2(\ell)} F.$$

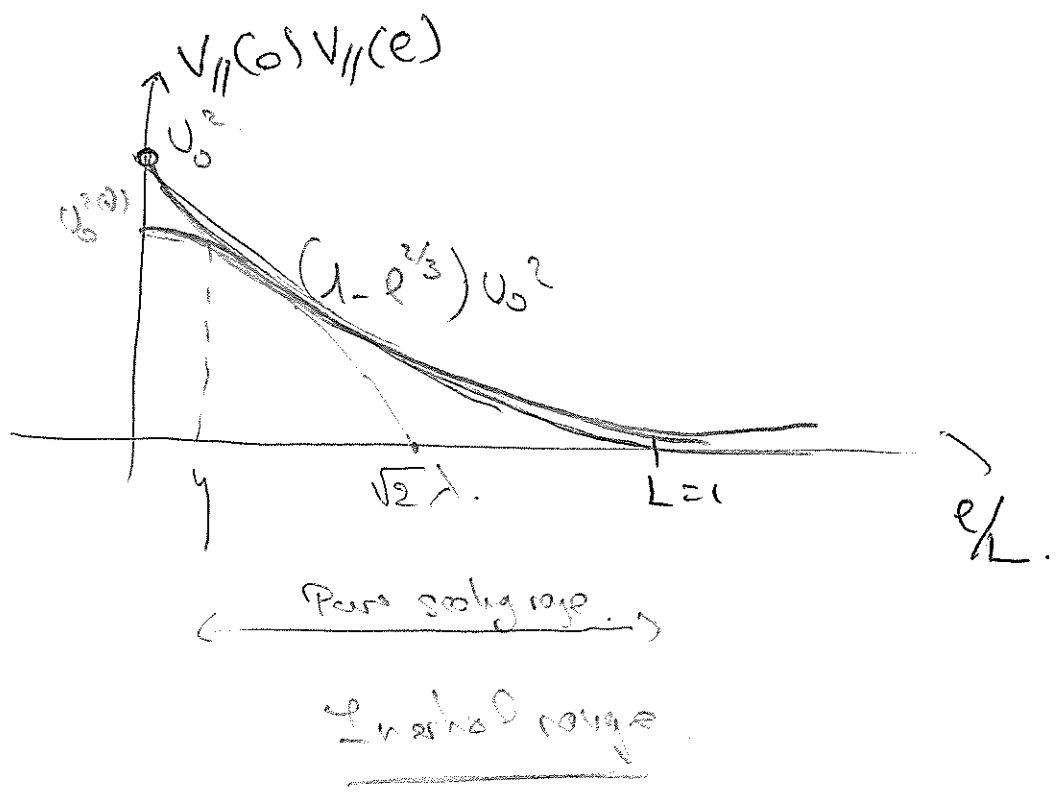
with the scaling $S_2(\ell) \sim \ell^{2/3}$

then, in the inertial range:

$$\left| \begin{aligned} \nu_e &= \frac{\nu}{S_2^{1/2}(\ell) \ell} = \frac{\nu}{\ell^{1/2} \ell^{1/2} \ell} = \frac{\nu}{\ell^{3/2} \ell^{1/2}} = \left(\frac{\nu^3}{\ell^4 \ell^2} \right)^{1/3} = \left(\frac{\nu}{\ell} \right)^{4/3} \ll 1 \\ f_e &= \frac{\ell}{\ell^{2/3} \ell^{2/3}} F = \frac{\ell^{1/3}}{\ell^{2/3}} F \ll \frac{F L^{1/3}}{\ell^{2/3}} = O(\epsilon) \end{aligned} \right.$$

we formally recover the Euler!

↑ cst.



Obs : $Re = \left(\frac{L}{y}\right)^{4/3}$.

$Re_\lambda = \frac{U_\lambda}{\nu} = Re^{1/2}$.

Calculation: $\langle V_{||}(0) V_{||}(e) \rangle = \left(1 - \frac{e^2}{2L^2}\right) U_0^2$

↑
Taylor's exp.

$\epsilon = 15 \nu \frac{U_0^2}{L^2}$.

Question is the Gaussian HIT indeed realized in "true" steady state turbulence ??

→ same "which is which"