

Stochastic Partial Differential Equations in Fluid Mechanics

Lecture 3: Stochastic Navier-Stokes equations

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Summary (and outline)

- In the first lecture we have discussed the origin of *noise from boundary perturbations*
- And we have started the rigorous investigation of *deterministic NS eq. with rough data*
- In the second lecture we have completed the rigorous investigation of deterministic NS eq. with rough data
- And we have introduced *stochastic NS eq., Itô formula, average energy identity*
- This identifies an open problem, namely that *additive noise introduces energy in the average*, so we should subtract it with some additional term.

(Summary and) outline

- Today we discuss *state-dependent noise* from the physical view-point
- and we prove rigorous results.
- The next two lectures will be devoted to *noise of transport type*
- discussing its physical origin from large-small scale decomposition
- and its consequences on turbulence theory.

Summary: Navier-Stokes equations with additive noise

$$\begin{aligned} du + (u \cdot \nabla u + \nabla p) dt &= (v \Delta u + f) dt + \sum_{k \in K} \sqrt{\lambda_k} \sigma_k dW_t^k \\ \operatorname{div} u &= 0 \end{aligned}$$

$$u|_{\partial D} = 0$$

$$u|_{t=0} = u_0$$

where W_t^k are independent Brownian motions.

Theorem

If $\mathbb{E} \|u_0\|_{L^2}^2 < \infty$ then

$$u \in C_{\mathcal{F}}([0, T]; H) \cap L^2_{\mathcal{F}}(0, T; V)$$

and

$$\mathbb{E} \left[\|u(t)\|_{L^2}^2 \right] + 2\nu \int_0^t \mathbb{E} \|\nabla u(s)\|_{L^2}^2 ds = \mathbb{E} \left[\|u_0\|_{L^2}^2 \right] + t \sum_{k \in K} \lambda_k \|\sigma_k\|_{L^2}^2$$

$$\mathbb{E} \left[\sup_{t \in [0, T]} \|u(t)\|_{L^2}^2 \right] \leq \mathbb{E} \left[\|u_0\|_{L^2}^2 \right] + T \sum_{k \in K} \sqrt{\lambda_k} \|\sigma_k\|_{L^2}^2 \\ + C \sum_{k \in K} \lambda_k \mathbb{E} \int_0^T \langle u(s), \sigma_k \rangle^2 ds.$$

- The solution has integrability properties in ω reflecting analogous properties assumed on the data.
- Assume $u(t)$ statistically stationary solution: $\mathbb{E} \|u(t)\|_{L^2}^2 = \mathbb{E} \|u_0\|_{L^2}^2$ and $\mathbb{E} \|\nabla u(s)\|_{L^2}^2$ is independent s . Then:

$$\epsilon := \nu \mathbb{E} \|\nabla u_\nu\|_{L^2}^2 ds = \frac{1}{2} \sum_{k \in K} \lambda_k \|\sigma_k\|_{L^2}^2.$$

The dissipation ϵ of energy due to viscosity remains constant in the inviscid limit $\epsilon \rightarrow 0$ (it is a statement of K41 theory), if the energy injection is constant.

Open problem

- Creating vortices from nothing we introduce energy into the system.
- Therefore we should include an extra dissipation mechanism. Possible proposal:

$$du + (u \cdot \nabla u + \nabla p) dt = \left(\nu \Delta u - \underline{\lambda(x) u} \right) dt + \sum_{k \in K} \sqrt{\lambda_k} \sigma_k dW_t^k$$

The energy balance is now

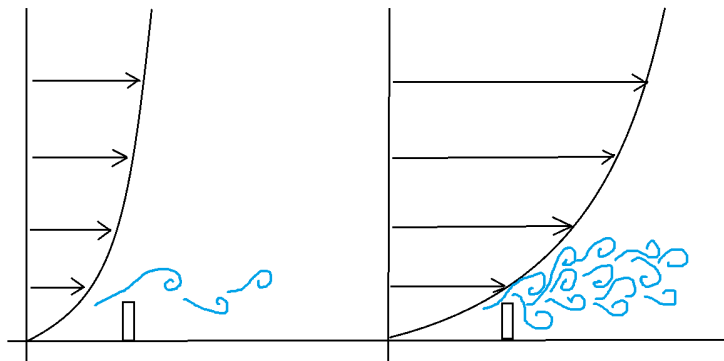
$$\begin{aligned} & \mathbb{E} \left[\|u(t)\|_{L^2}^2 \right] + 2\nu \int_0^t \mathbb{E} \|\nabla u\|_{L^2}^2 ds + 2\mathbb{E} \int_0^t \int_D \lambda |u|^2 dx ds \\ &= \mathbb{E} \left[\|u_0\|_{L^2}^2 \right] + t \sum_{k \in K} \lambda_k \|\sigma_k\|_{L^2}^2. \end{aligned}$$

But we should be able to choose $\lambda(x)$ in such a way that

$$2\mathbb{E} \int_0^t \int_D \lambda(x) |u(s, x)|^2 dx ds \sim t \sum_{k \in K} \lambda_k \|\sigma_k\|_{L^2}^2.$$

We do not know how to reach this target.

Example of state-dependent noise



Example of state-dependent noise

Jumps (creation of new structures) by non-homogeneous Poisson process with instantaneous rate

$$\lambda_k(u(t)) = \chi^2 \left(\frac{1}{|B(x_k, r)|} \int_{B(x_k, r)} |u(t, y)| dy \right)$$

where χ^2 is a nondecreasing non-negative function, equal to zero in zero and $r > 0$ is a length scale relevant to the problem. Then we introduce the cumulative rate

$$\Lambda_k(t) = \int_0^t \lambda_k(u(s)) ds$$

Example of state-dependent noise

Finally we modify the Poisson process N_t^k by this rate, namely we consider the process

$$N_{\Lambda_k(t)}^k.$$

The case previously considered was simply

$$\lambda_k(u(t)) = \lambda_k, \quad \Lambda_k(t) = \lambda_k t, \quad N_{\lambda_k t}^k.$$

$$\partial_t u + u \cdot \nabla u + \nabla p = \nu \Delta u + f + F(u) + \sum_{k \in K} \frac{1}{\sqrt{2}} \sigma_k \partial_t \left(N_{\Lambda_k(t)}^{k,1} - N_{\Lambda_k(t)}^{k,2} \right)$$

Example of state-dependent noise

$$\sum_{k \in K} \frac{1}{n\sqrt{2}} \sigma_k(x) \left(N_{n^2 \Lambda_k(t)}^{k,1} - N_{n^2 \Lambda_k(t)}^{k,2} \right). \quad (1)$$

$$B_{\Lambda_k(t)}^k = \int_0^t \sqrt{\lambda_k(u(s))} dW_s^k$$

(jointly in k). This result is undoubtedly advanced and not trivial even at the heuristic level but notice at least the analogy with the coefficients $\sqrt{\lambda_k}$ in the case of constant rate: when $\lambda_k(u(s)) = \lambda_k$, $\Lambda_k(t) = \lambda_k$, the previous identity reads

$$B_{\lambda_k t}^k = \int_0^t \sqrt{\lambda_k} dW_s^k = \sqrt{\lambda_k} W_t^k$$

$$\partial_t u + u \cdot \nabla u + \nabla p = \nu \Delta u + f + F(u) + \sum_{k \in K} \sigma_k(u) \partial_t W_t^k$$

by introducing the maps $\sigma_k : H \rightarrow H$ given by

$$\sigma_k(u)(x) = \sigma_k(x) \sqrt{\lambda_k(u)}.$$

The Wong-Zakai corrector

The right corrector is:

$$F(u) = \frac{1}{2} \sum_{k \in K} D\sigma_k(u) \sigma_k(u).$$

Here by $D\sigma_k(u)$ we mean the Frechét Jacobian of $\sigma_k(u)$, which is a linear bounded operator from H to H , under suitable assumptions, and $D\sigma_k(u) \sigma_k(u)$ is the application of the linear map $D\sigma_k(u)$ to the element $\sigma_k(u)$ of H .

$$\begin{aligned} \partial_t u + u \cdot \nabla u + \nabla p &= v \Delta u + f \\ &+ \frac{1}{2} \sum_{k \in K} D\sigma_k(u) \sigma_k(u) + \sum_{k \in K} \sigma_k(u) \partial_t W_t^k. \end{aligned}$$

Explanation in one-dimension

Consider the one dimensional equation, with $\sigma(x) \geq \nu > 0$,

$$\frac{dX_t^\epsilon}{dt} = \sigma(X_t^\epsilon) \frac{dW_t^\epsilon}{dt}$$

where W_t^ϵ is an approximation of a Brownian motion W_t . It is an equation with separated variables. Then

$$\int_0^T \frac{\frac{dX_t^\epsilon}{dt}}{\sigma(X_t^\epsilon)} dt = \int_0^T \frac{dW_t^\epsilon}{dt} dt$$

$$\Phi(X_T^\epsilon) - \Phi(x_0) = W_T^\epsilon, \quad \Phi'(x) = \frac{1}{\sigma(x)}$$

$$X_t^\epsilon = \Phi^{-1}(\Phi(x_0) + W_t^\epsilon)$$

Hence X_t^ϵ converges weakly to X_t given by

$$X_t = \Phi^{-1}(\Phi(x_0) + W_t).$$

From Ito formula, since

$$\begin{aligned} D\Phi^{-1}(x) &= \frac{1}{\Phi'(\Phi^{-1}(x))} = \sigma(\Phi^{-1}(x)) \\ D^2\Phi^{-1}(x) &= \sigma'(\Phi^{-1}(x)) \sigma(\Phi^{-1}(x)) \\ dX_t &= \sigma(X_t) dW_t + \frac{1}{2} \sigma'(X_t) \sigma(X_t) dt. \end{aligned}$$

(Link with Stratonovich integrals)

Definition

Given $u_0 \in L^2_{\mathcal{F}_0}(\Omega, H)$ and $f \in L^2_{\mathcal{F}}(0, T; V')$, we say that

$$u \in C_{\mathcal{F}}([0, T]; H) \cap L^2_{\mathcal{F}}(0, T; V)$$

is a weak solution if

$$\begin{aligned} & \langle u(t), \phi \rangle - \int_0^t b(u(s), \phi, u(s)) ds \\ = & \langle u_0, \phi \rangle + \int_0^t \langle u(s), A\phi \rangle ds + \int_0^t \langle f(s), \phi \rangle ds \\ & + \int_0^t \langle F(u(s)), \phi \rangle ds + \sum_{k \in K} \int_0^t \langle \sigma_k(u(s)), \phi \rangle dW_s^k \end{aligned}$$

for every $\phi \in D(A)$.

Theorem

For every $u_0 \in L^2_{\mathcal{F}_0}(\Omega, H)$ and $f \in L^2_{\mathcal{F}}(0, T; V')$, there exists a unique weak solution. It satisfies

$$\begin{aligned} & \mathbb{E} \left[\|u(t)\|_{L^2}^2 \right] + 2\nu \mathbb{E} \int_0^t \|\nabla u(s)\|_{L^2}^2 ds \\ &= \mathbb{E} \left[\|u_0\|_{L^2}^2 \right] + 2\mathbb{E} \int_0^t \langle u(s), f(s) + F(u(s)) \rangle ds \\ &+ \sum_{k \in K} \mathbb{E} \int_0^t \|\sigma_k(u(s))\|_{L^2}^2 ds. \end{aligned}$$

Proof of uniqueness

Let $u^{(i)}$ be two solutions. Then $w = u^{(1)} - u^{(2)}$ satisfies

$$\begin{aligned} & \langle w(t), \phi \rangle - \int_0^t (b(w(s), \phi, w(s))) ds \\ = & \int_0^t \langle w(s), A\phi \rangle ds + \int_0^t \langle F(u^{(1)}(s)) - F(u^{(2)}(s)), \phi \rangle ds \\ & + \sum_k \int_0^t \langle \sigma_k(u^{(1)}(s)) - \sigma_k(u^{(2)}(s)), \phi \rangle dW_s^k \\ & - \int_0^t (b(u^{(2)}, \phi, w) + b(w, \phi, u^{(2)}))(s) ds. \end{aligned}$$

No problem to estimate the F and σ_k terms, assumed to be Lipschitz.
The problem is the last term.

From Itô formula:

$$\begin{aligned} & \|w(t)\|_H^2 + \nu \int_0^t \|\nabla w(s)\|_H^2 ds \\ & \leq C \int_0^t \|w(s)\|_H^2 ds + C \int_0^t \|w(s)\|_H^2 \left(1 + \|u^{(2)}(s)\|_{\mathbb{L}^4}^2\right) ds + M_t \end{aligned}$$

$$M_t := \sum_k \int_0^t \left\langle \sigma_k(u^{(1)}(s)) - \sigma_k(u^{(2)}(s)), w(s) \right\rangle dW_s^k.$$

The difficult term coming from

$$\int_0^t \left(b(u^{(2)}, w, w) + b(w, w, u^{(2)}) \right) (s) ds.$$

Taking expected values, the cubic term "does not close".

We need now a very interesting trick that we have learned from Bjorn Schmalfuss: introduced

$$\rho_t = \exp \left(-C \int_0^t \left(1 + \left\| u^{(2)}(s) \right\|_{\mathbb{L}^4}^2 \right) ds \right)$$

we have, from Itô formula again,

$$\|w(t)\|_H^2 \rho_t + \nu \int_0^t \|\nabla w(s)\|_H^2 \rho_s ds \leq C \int_0^t \|w(s)\|_H^2 \rho_s ds + \tilde{M}_t$$

$$\tilde{M}_t := \sum_{k \in K} \int_0^t \left\langle \sigma_k \left(u^{(1)}(s) \right) - \sigma_k \left(u^{(2)}(s) \right), w(s) \right\rangle \rho_s dW_s^k.$$

$$\mathbb{E} \left[\|w(t)\|_H^2 \rho_t \right] + \nu \mathbb{E} \int_0^t \|\nabla w(s)\|_H^2 \rho_s ds \leq C \int_0^t \mathbb{E} \left[\|w(s)\|_H^2 \rho_s \right] ds$$

which leads to $\mathbb{E} \left[\|w(t)\|_H^2 \rho_t \right] = 0$ by Gronwall lemma.

- Compactness criteria
- Strategies to deal with convergence of laws
- The 3D case

Deterministic Aubin-Lions type theorems

Given two Banach spaces $X \subset Y$, we say that the embedding $X \subset Y$ is compact if bounded sets of X are relatively compact in Y .

Theorem

Let $X \subset Y \subset Z$ be three Banach spaces, with continuous dense embeddings. Assume that the embedding $X \subset Y$ is compact. Let $p \in [1, \infty)$ be given. Then the embedding

$$L^p(0, T; X) \cap W^{1,1}(0, T; Z) \subset L^p(0, T; Y)$$

is compact.

Several variants exist (Simon).

Useful in the stochastic case:

Theorem

If $\alpha r > 1 - \frac{r}{p}$ ($p, r \geq 1$) then

$$L^p(0, T; X) \cap W^{\alpha, r}(0, T; Z) \stackrel{\text{compact}}{\subset} L^p(0, T; Y)$$

Here $\alpha \in (0, 1)$ and $W^{\alpha, r}(0, T; Z)$ is the space of functions $f \in L^r(0, T; Z)$ such that

$$\int_0^T \int_0^T \frac{\|f(t) - f(s)\|_Z^r}{|t - s|^{1 + \alpha r}} ds dt < \infty.$$

Given a probability space $(\Omega, \mathcal{F}, \mathbb{P})$ compactness criteria in $L^p(\Omega)$ are not natural (although something can be done by weighted Sobolev spaces and Malliavin calculus, when $(\Omega, \mathcal{F}, \mathbb{P})$ has a special structure).

The natural approach is to consider the laws of the random objects and apply compactness arguments to these laws.

Let (X, d) be a complete metric space and \mathcal{B} the Borel σ -field.

A family \mathcal{G} of probability measures on (X, \mathcal{B}) is *tight* if for every $\epsilon > 0$ there is a compact set $K \subset X$ such that

$$\mu(K) \geq 1 - \epsilon$$

for all $\mu \in \mathcal{G}$.

Theorem (Prohorov)

A family \mathcal{G} of probability measures on (X, \mathcal{B}) is tight if and only if it is relatively compact.

Corollary

Assume (u_N) is a sequence of random functions from $(\Omega, \mathcal{F}, \mathbb{P})$ to $L^p(0, T; Y)$. Assume $\alpha r > 1 - \frac{r}{p}$ and that for every $\epsilon > 0$ there are $R_1, R_2 > 0$ such that

$$\begin{aligned}\mathbb{P}\left(\|u_N\|_{L^p(0, T; X)} \geq R_1\right) &\leq \epsilon \\ \mathbb{P}\left(\|u_N\|_{W^{\alpha, r}(0, T; Z)} \geq R_2\right) &\leq \epsilon.\end{aligned}$$

Then there exists a subsequence (u_{N_k}) which converges in law, in the strong topology of $L^p(0, T; Y)$, to a random function \tilde{u} from a probability space $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{\mathbb{P}})$ to $L^p(0, T; Y)$. Moreover, if $p, r > 1$, we may choose (u_{N_k}) so that \tilde{u} takes also values in $L^p(0, T; X)$ and $W^{\alpha, r}(0, T; Z)$. If u_N are (\mathcal{F}_t) -progressively measurable, there exists $(\tilde{\mathcal{F}}_t)$ such that \tilde{u} is $(\tilde{\mathcal{F}}_t)$ -progressively measurable.

Sufficient conditions:

$$\begin{aligned}\mathbb{E} \left[\|u_N\|_{L^p(0,T;X)} \right] &\leq C \\ \mathbb{E} \left[\|u_N\|_{W^{\alpha,r}(0,T;Z)} \right] &\leq C.\end{aligned}$$

Indeed, by Markov inequality,

$$\mathbb{P} \left(\|u_N\|_{L^p(0,T;X)} \geq R_1 \right) \leq \frac{C}{R_1}$$

and similarly for the second inequality, hence given $\epsilon > 0$ we can find $R_1, R_2 > 0$ with the required properties.

Galerkin approximations: 2D and 3D case

Until now we have described only the 2D case. Now it is convenient to unify something with the 3D case.

A main difference is that, after the common inequality

$$b(u, v, w) \leq \|v\|_V \|u\|_{\mathbb{L}^4} \|w\|_{\mathbb{L}^4}$$

we have different Sobolev embeddings::

$$\|f\|_{L^4} \stackrel{d=2}{\leq} \|f\|_{W^{\frac{1}{2},2}} \leq \|f\|_{L^2}^{1/2} \|f\|_{W^{1,2}}^{1/2}$$

$$\|f\|_{L^4} \stackrel{d=3}{\leq} \|f\|_{W^{\frac{3}{4},2}} \leq \|f\|_{L^2}^{1/4} \|f\|_{W^{1,2}}^{3/4} .$$

Notations in 2D and 3D

- (e_i) complete orthonormal system in H made of eigenvectors of A , with eigenvalues $(-\lambda_i)$,
- H_n and π_n are consequently defined
- bilinear operator $B_n : H_n \times H_n \rightarrow H_n$ defined as

$$B_n(u, v) = \pi_n P(u \cdot \nabla v)$$

- equation

$$du_n + B_n(u_n, u_n) dt = (Au_n + f_n + F_n(u_n)) dt + \sum_k \sigma_k^n(u_n) dW_t^k$$

where $f_n = \pi_n f$, $F_n(u) = \pi_n F(u)$, $\sigma_k^n(u_n) = \pi_n \sigma_k(u_n)$; with initial condition $u_0^n = \pi_n u_0$

- It is easy to check that

$$\langle B_n(u_n, u_n), u_n \rangle = 0.$$

Common estimates in 2D and 3D

From

$$\begin{aligned} \|u_n(t)\|_H^2 + 2\nu \int_0^t \|\nabla u_n(s)\|_H^2 ds &= 2 \int_0^t \langle f_n(s) + F(u_n(s)), u_n(s) \rangle ds \\ &\quad + \sum_{k \in K} \int_0^t \|\sigma_k^n(u_n(s))\|_{L^2}^2 ds + M_t^n \end{aligned}$$

we easily deduce

$$\begin{aligned} \mathbb{E} \int_0^T \|u_n(s)\|_V^2 ds &\leq C \\ \mathbb{E} \left[\sup_{t \in [0, T]} \|u_n(t)\|_H^2 \right] &\leq C. \end{aligned}$$

With more work (just technical, see the notes), assuming

$$\mathbb{E} \int_0^t \|f(s)\|_{V'}^r ds < \infty$$

for some $r > 2$, we prove

$$\mathbb{E} \left[\sup_{t \in [0, T]} \|u_n(t)\|_H^r \right] \leq C.$$

With more work (just technical, see the notes), we can prove

$$\mathbb{E} \int_0^T \int_0^T \frac{\|u_n(t) - u_n(s)\|_{V'}^r}{|t-s|^{1+\alpha r}} ds dt \leq C$$

if

$$1 - \frac{r}{2} < \alpha r < \frac{r}{2}.$$

Given $r > 1$ there exists $\alpha \in (0, \frac{1}{2})$ with such property.

Thus we have:

Theorem

There exist (α, r) with $\alpha r > 1 - \frac{r}{2}$ and $C > 0$ such that

$$\mathbb{E} \left[\|u_n\|_{W^{\alpha,r}(0,T;V')} \right] \leq C.$$

Form the previous results:

Corollary

The family of laws of u_n is tight in $L^2(0, T; H)$. Hence there exist subsequences which converge in law.

- We have seen above that the estimates on (u_n) are similar in the 2D and 3D case.
- The problem is uniqueness: in the 2D case we have proved it. In the 3D case it is an open problem (see below).
- When we have uniqueness, we can upgrade the convergence in law to convergence in probability (see below).
- Without uniqueness, we can pass to the limit "in law" and find solutions on an auxiliary probability space $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{\mathbb{P}})$.