

Densities for the Navier–Stokes equations with noise

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Summary

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The equations

Consider the Navier–Stokes equations,

$$\begin{cases} \dot{\mathbf{u}} + (\mathbf{u} \cdot \nabla)\mathbf{u} + \nabla p = \nu \Delta \mathbf{u} + \dot{\eta}, \\ \operatorname{div} \mathbf{u} = 0, \\ \mathbf{u}(0) = \mathbf{x}, \end{cases}$$

either on the torus \mathbb{T}_3 with periodic boundary conditions (and zero mean) or on a bounded domain with Dirichlet boundary conditions. Here $\dot{\eta}$ is Gaussian noise which is white in time and coloured in space.

For instance assume periodic boundary conditions,

$$\dot{\eta}(t, \mathbf{y}) = S dW = \sum_{\mathbf{k} \in \mathbb{Z}_*^3} \sigma_{\mathbf{k}} \dot{\beta}_{\mathbf{k}}(t) e_{\mathbf{k}}(\mathbf{y}),$$

and $e_{\mathbf{k}}(\mathbf{y}) = e^{i\mathbf{k} \cdot \mathbf{y}}$ are the Fourier exponential.

What we look for and why

Theorem

Under **suitable** assumptions on the covariance of the driving noise, **suitable** finite-dimensional projections of any martingale weak solution of the 3D Navier-Stokes equations have a density with respect to the Lebesgue measure.

- Why densities?
 - probabilistic form of regularity (hot issue for NSE),
 - better understanding of the evolution of the flow,
 - ideas related to uniqueness in law,
 - quantifying uncertainty.
- Why do we consider finite dimensional projections:
 - there is a reference measure,
 - easier: curse of regularity solved,
 - ideas related to long time behaviour [\[MatPar2006\]](#) [\[HaiMat2006\]](#)

Existence of densities

We focus on the following problem

$$dx = b(x) dt + dB$$

How to prove the existence of a density:

- Girsanov transformation,
- the Fokker-Planck equation,
- Malliavin calculus,
- ??????

A standard method

The idea is to use integration by parts. Having a smooth density would give,

$$\mathbb{E}\left[\frac{\partial\phi}{\partial h}(x_t)\right] = \int \frac{\partial\phi}{\partial h}(y) f_t(y) dy = - \int \phi(y) \frac{\partial f_t}{\partial h}(y) dy.$$

So, if we know that

$$\left| \mathbb{E}\left[\frac{\partial\phi}{\partial h}(x_t)\right] \right| \leq c|h| \|\phi\|_\infty$$

by duality there is a density.

How to prove this?

Integration by parts

Assume that we can find a variation of the noise H that compensates the variation h , that is $h = \mathcal{D}_H x_t$. Then

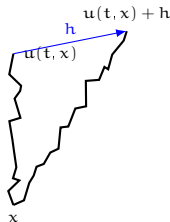
$$\begin{aligned} \mathbb{E}\left[\frac{\partial \phi}{\partial h}(x_t)\right] &= \mathbb{E}[\mathcal{D}\phi(x_t)h] = \mathbb{E}[\mathcal{D}\phi(x_t)\mathcal{D}_H x_t] = \\ &= \mathbb{E}[\mathcal{D}_H \phi(x_t)] = \mathbb{E}\left[\phi(x_t) \int H dW\right] \end{aligned}$$

by the **chain rule** and **integration by parts**.

How to find H ?: essentially one needs to (pseudo)invert the map

$$\Psi H = \int_0^t H(s) \mathcal{D}_s x_t ds$$

that is $H = \Psi^* \mathcal{M}_t^{-1} h$.



Malliavin calculus and Navier-Stokes

We are dealing with a process u such that

$$du - \nu \Delta u \, dt + (u \cdot \nabla)u \, dt + \nabla p = S \, dW,$$

whose Malliavin derivative $\mathcal{D}_H u$ satisfies

$$\frac{d}{dt} \mathcal{D}_H u - \nu \Delta \mathcal{D}_H u + (u \cdot \nabla) \mathcal{D}_H u + (\mathcal{D}_H u \cdot \nabla)u + \nabla q = S H,$$

- NO!** {
- Trying to prove invertibility of the Malliavin matrix forces to use the dynamics of the **linearisation**.
 - A decent estimate on the Malliavin derivative is the same as an estimate on the difference of two solutions!
- MAYBE** {
- the dynamics is well defined and good for short times,
 - an invertible covariance allows to prove invertibility without using the dynamics.

It falls in the class of equations with non-regular coefficients,

- Fournier-Printemps (starting point for what we'll explain here),
- Bally and Caramellino (based on interpolation),
- Kohatsu-Higa and co-workers.

Primer on Besov spaces

Set

$$(\Delta_h^1 f)(x) = f(x+h) - f(x), \quad (\Delta_h^n f)(x) = \Delta_h^1 (\Delta_h^{n-1} f)(x)$$

and given any integer $n > s$,

$$\|f\|_{B_{p,q}^s} = \|f\|_{L^p} + \left(\int_{|h| \leq 1} \frac{\|\Delta_h^n f\|_{L^p}^q}{|h|^{sq}} \frac{dh}{|h|^n} \right)^{\frac{1}{q}} \quad \left[+ \sup_{|h| \leq 1} \frac{\|\Delta_h^n f\|_{L^p}}{|h|^s} \right]$$

Some nice properties of Besov spaces

- different (more general) definition in terms of Littlewood-Paley decomposition,
- $B_{p,p}^s = W^{s,p}$, $B_{\infty,\infty}^s = C^s$, $s \geq 0$ non-integer,
- $(B_{p,q}^s)' = B_{p',q}^{-s}$, $q < \infty$,
- $(I - \Delta) : B_{p,q}^s \longrightarrow B_{p,q}^{s-2}$ isomorphism,

Smoothing lemma

Theorem (deterministic fractional integration by parts)

If μ is a finite measure on \mathbf{R}^d and there are $s > 0$, an integer $m > s$, and $\alpha \in (0, 1)$ such that for every $\phi \in C_c^\infty(\mathbf{R}^d)$ and every $h \in \mathbf{R}^d$,

$$\left| \iint_{\mathbf{R}^d} \Delta_h^m \phi(x) \mu(dx) \right| \leq c|h|^s \|\phi\|_{C_b^\alpha},$$

then μ has a density $f \in B_{1,\infty}^r$ for every $r < s - \alpha$.

We use the above lemma to prove existence of a density for the solution x of our simple model problem at time $t = 1$

$$dx = b(x) dt + dB.$$

We look for an estimate

$$\mathbb{E}[\phi(x_1 + h) - \phi(x_1)] \approx |h|^s \|\phi\|_{C_b^\alpha}$$

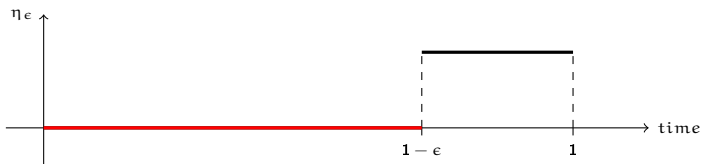
Fact 1: exploiting the short times

Recall: wish to estimate

$$\mathbb{E}[\phi(x_1 + h) - \phi(x_1)] \approx |h|^s \|\phi\|_{C_b^\alpha}$$

Consider

$$dx^\epsilon = \eta_\epsilon(t)b(x^\epsilon) + dB.$$



$$\begin{cases} dx = b(x) + dB, \\ dx^\epsilon = b(x^\epsilon) + dB, \end{cases}$$

[fournier-printemps (2010)]

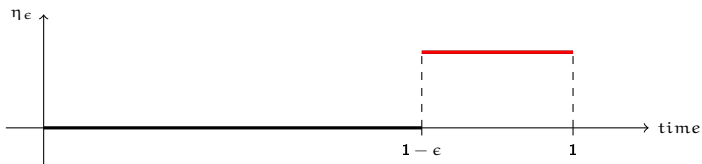
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[fournier-printemps (2010)]

Decomposition of the error

Hence

- $s \leq 1 - \epsilon$: $x^\epsilon(s) = x(s)$,
- $s \geq 1 - \epsilon$: $x^\epsilon(s)$ is the one-step explicit Euler approximation of x , starting at time $1 - \epsilon$ with time step ϵ .

and

$$\begin{aligned}\mathbb{E}[\phi(x_1 + h)] - \mathbb{E}[\phi(x_1)] &= \mathbb{E}[\phi(x_1 + h)] - \mathbb{E}[\phi(x_1^\epsilon + h)] \\ &\quad + \mathbb{E}[\phi(x_1^\epsilon + h)] - \mathbb{E}[\phi(x_1^\epsilon)] \\ &\quad + \mathbb{E}[\phi(x_1^\epsilon)] - \mathbb{E}[\phi(x_1)]\end{aligned}$$

where

- **numerical error**
- **probabilistic error**

The numerical error

There is not much that can be done here,

$$\mathbb{E}[\phi(x_1^\epsilon)] - \mathbb{E}[\phi(x_1)] \leq [\phi]_\alpha \mathbb{E}[\|x_1^\epsilon - x_1\|^\alpha]$$

and

$$|x_1 - x_1^\epsilon| = \left| \int_{1-\epsilon}^1 b(x_s) ds \right| \leq \|b\|_\infty \epsilon,$$

so that

$$\text{numerical error} \approx \epsilon^\alpha$$

How to improve it?

The numerical error can be made smaller using a higher order numerical method.

More regularity of the drift b needed.

Fact 2: inverting the covariance (a.k.a. the probabilistic error)

Absorb the variation in the noise

$$\begin{aligned} x_1^\epsilon(t) + h &= \text{i. c.} + \int_{1-\epsilon}^1 dB_s + h \\ &= \text{i. c.} + \int_{1-\epsilon}^1 (dB_s + H ds) \end{aligned}$$

with an obvious value of $H \approx \frac{h}{\epsilon}$ and by the Girsanov formula (actually Cameron–Martin here!)

$$\text{prob. error} \approx \mathbb{E}[(G_1 - 1)\phi(x_1^\epsilon)] \approx \frac{|h|}{\sqrt{\epsilon}} \|\phi\|_\infty$$

How to improve it? Look for higher order increments $\Delta_h^n \phi(x_1)$ that yield the same numerical error but

$$\text{prob. error} \approx \left(\frac{|h|}{\sqrt{\epsilon}}\right)^n.$$

A balance of regularity

Putting all together,

$$\text{total error} \approx \text{probabilistic} + \text{numeric} \approx \frac{|h|}{\sqrt{\epsilon}} \|\phi\|_{\infty} + [\phi]_{\alpha} \epsilon^{\alpha}$$

and by optimizing in ϵ ,

$$\mathbb{E}[\phi(x_1 + h)] - \mathbb{E}[\phi(x_1)] \approx |h|^{\frac{2\alpha}{2\alpha+1}} \|\phi\|_{C^{\alpha}}$$

- “ α derivatives” given for the estimate,
- “ $\frac{2\alpha}{2\alpha+1}$ derivatives” obtained
- use the smoothing lemma with $n = 1$, $\alpha \rightarrow \alpha$ and $s = \frac{2\alpha}{2\alpha+1}$.

Other functionals

The same idea can be used on quantities that **derive** from an equation (although might not solve one), such as in Navier–Stokes the balance of energy:

$$\mathcal{E}_t = \frac{1}{2} \int |u(t, x)|^2 dx + \nu \int_0^t \int |\nabla u(s, x)|^2 dx ds$$

To keep it simple, let us think again at $x \in \mathbf{R}^d$,

$$dx = b(x) dt + dB, \quad \text{and} \quad \mathcal{E}_t = \sum_i (x_t^i)^2$$

It is immediate to check that the corresponding energy computed for x^ϵ has a generalized χ^2 distribution.

Apparently odd: it turns out that the regularity depends on the number of degrees of freedom of the χ^2 (here d), and only in infinite dimension the unconditioned result is recovered.

Time regularity

Let f be the density of the solution of

$$dx = b(x) dt + dB$$

Theorem

$$\|f(t) - f(s)\|_{B_{1,\infty}^\alpha} \lesssim |t - s|^{\beta/2}, \text{ for all } \alpha + \beta < 1$$

- $\|f(t) - f(s)\|_{L^1} \lesssim |t - s|^{\frac{1}{2}-}$, more challenging, through Girsanov transformation,
- $\|f(t) - f(s)\|_{B_{1,\infty}^\alpha} \lesssim |t - s|^{\beta/2}$ easier:

$$\|\Delta_h^n(f(t) - f(s))\|_{L^1} \lesssim \begin{cases} \|\Delta_h^n f(t)\|_{L^1} + \|\Delta_h^n f(s)\|_{L^1} & |h| \ll |t - s|, \\ \|f(t) - f(s)\|_{L^1} & |t - s| \ll |h| \end{cases}$$

The associated Fokker-Planck equation

Consider again

$$dx = b(x) dt + dB, \quad x(0) = x_0$$

with $b \in L^\infty$. We know that the density $f \in B_{1,\infty}^{1-}$. In fact we can show that $f \in B_{\infty,\infty}^{1-}$. The density solves

$$\begin{cases} \partial_t f = \frac{1}{2} \Delta f - \nabla \cdot (bf), \\ f(0) = \delta_{x_0}, \end{cases}$$

that is

$$f(t, x) = p_t(x - x_0) + \int_0^t \nabla p_{t-s} \star (bf)(x) dt,$$

where p_t is the heat kernel.

Besov bounds with the Fokker-Planck equation

$$f(t, x) = p_t(x - x_0) + \int_0^t \nabla p_{t-s} \star (bf)(x) dt,$$

hence

$$\begin{aligned} \|\Delta_h^2 f(t, x)\|_{L^1} &= \left\| \Delta_h^2 p_t(x - x_0) + \int_0^t (\Delta_h^2 \nabla p_{t-s}) \star (bf)(x) dt \right\|_{L^1} \\ &\leq \|\Delta_h^2 p_t\|_{L^1} + \|b\|_\infty \|f\|_{L^1} \int_0^t \|\Delta_h^2 \nabla p_{t-s}\|_{L^1} ds \end{aligned}$$

and conclude with estimates for the heat kernel

Hölder bounds with the Fokker-Planck equation

To see the Hölder bound, consider for simplicity the equation for the stationary solution (density of the invariant measure):

$$\frac{1}{2}\Delta f - \nabla \cdot (bf) = 0,$$

that is, if g is the Poisson kernel ($g(x) = |x|^{2-d}$, $d \geq 3$),

$$\begin{aligned} f(x) &= \nabla g \star (bf) = (\nabla g \mathbb{1}_{B_\epsilon(0)}) \star (bf) + (\nabla g \mathbb{1}_{B_\epsilon(0)^c}) \star (bf) \\ &\lesssim \|b\|_\infty \|f\|_\infty \epsilon^d + \epsilon^{1-d} \|b\|_\infty \|f\|_{L^1} \end{aligned}$$

and choose $\epsilon = (2\|b\|_\infty)^{-1/d}$. The Hölder norm follows likewise.

Watch out! Needs to know that already $\|f\|_\infty < \infty$.

Some open problems

- Local estimate (see the problem with the energy).
- Probabilistic proof of the Fokker-Planck approach.
- Degenerate noise completely open (at least in this way).