

Path-by-path regularization by noise for scalar conservation laws

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joint work with: Panagiotis E. Souganidis, Benoît Perthame, Paul Gassiat,
Mario Maurelli, Khalil Chouk

[G., Souganidis; CMS, 2014], [G., Souganidis; CPAM, 2017],
[G., Perthame, Souganidis; SINUM, 2016], [Gassiat, G.; PTRF, 2018+],
[Maurelli, G.; arXiv:1701.05393], [Chouk, G.; arXiv:1512.06056].

Outline

- 1 Introduction
- 2 Regularization by nonlinear noise
- 3 Path-by-path regularization by noise
- 4 A path-by-path scaling condition

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- Uniqueness for the stochastic 3d-Navier-Stokes equations remains open.
- More recent: Linear multiplicative noise

$$du + b(x) \cdot \nabla u \, dt = \nabla u \circ d\beta_t \quad [\text{Flandoli, Gubinelli, Priola; } \textit{Invent. Math.}, 2010].$$

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- Side-remark: In certain cases [Delarue, Flandoli, Vincenzi; CPAM, 2014] space-time linear multiplicative noise

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- Different forms of noise?

Regularization by nonlinear noise

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Regularity of solutions for stochastic SCL

- Consider mean field equations

$$dX_t^i = \sigma^L \left(X_t^i, \frac{1}{L} \sum_{j=1}^L \delta_{X_t^j} \right) \circ d\beta_t \quad \text{in } \mathbb{R}^d$$

Taking $L \rightarrow \infty$ and $\sigma^L \rightarrow \sigma$ leads to stochastic scalar conservation laws

$$du + \operatorname{div} \left(\underbrace{\sigma(x, u)u}_{=: A(x, u)} \circ d\beta \right) = 0 \quad \text{on } (0, T) \times \mathbb{R}^d.$$

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- Methods apply to general spatially homogeneous and truly nonlinear flux A , general dimension.
- For simplicity, in this talk restrict to

$$du + \frac{1}{2} \partial_x u^2 \circ d\beta_t = 0.$$

Consider

$$\begin{aligned}\partial_t u + \frac{1}{2} \partial_x u^2 &= 0, \quad \text{on } (0, T) \times \mathbb{R}^d \\ u(0) &= u_0 \in L^\infty(\mathbb{R}^d).\end{aligned}$$

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For

$$\chi(t, x, v) = \chi(u(t, x), v) = 1_{v < u(t, x)} - 1_{v < 0}$$

we get the kinetic form

$$\partial_t \chi + v \partial_x \chi = \partial_v m \quad \text{on } (0, T) \times \mathbb{R}^d \times \mathbb{R}.$$

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Dissipation-dispersion approximations lead to

Definition (De Lellis, Otto, Westdickenberg, 2003)

A function $u \in L^\infty([0, T] \times \mathbb{R}^d)$ is said to be a quasi-solution if $\chi(t, x, v) = \chi(u(t, x), v)$ satisfies

$$\partial_t \chi + v \partial_x \chi = \partial_v m \quad \text{on } (0, T) \times \mathbb{R}^d \times \mathbb{R}$$

for some finite (signed) measure m .

Theorem (De Lellis, Westdickenberg, 2003; Jabin, Perthame 2002)

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Then

- ① *Each quasi-solution satisfies, for all $\lambda \in (0, \frac{1}{3})$,*

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$$u \in L^1([0, T]; W^{\lambda, 1}(\mathbb{R})).$$

- 2 For each $\lambda > \frac{1}{3}$ there exists a quasi-solution u such that

$$u \notin L^1([0, T]; W^{\lambda, 1}(\mathbb{R})).$$

Theorem (G., Souganidis; CPAM, 2016)

Let u be a quasi-solution to

$$du + \frac{1}{2} \partial_x u^2 \circ d\beta_t = 0 \quad \text{on } (0, T) \times \mathbb{T}.$$

Then,

$$u \in L^1([0, T]; W^{\lambda, 1}(\mathbb{T})) \quad \text{for all } \lambda \in (0, \frac{1}{2}), \mathbb{P}\text{-a.s..}$$

If u is an entropy solution, then

$$u(t) \in W^{\lambda, 1}(\mathbb{T}) \quad \text{for all } t > 0, \lambda \in (0, \frac{1}{2}), \mathbb{P}\text{-a.s..} \quad (\star)$$

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- 2 Characterize the properties of Brownian paths leading to (\star) .

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Framework

- Model example:

$$\partial_t u + \frac{1}{2} \partial_x u^2 \circ dw_t = 0 \quad \text{on } (0, T) \times \mathbb{T},$$

with $w \in C([0, T]; \mathbb{R})$.

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- Again: Results are given for general truly nonlinear flux A .

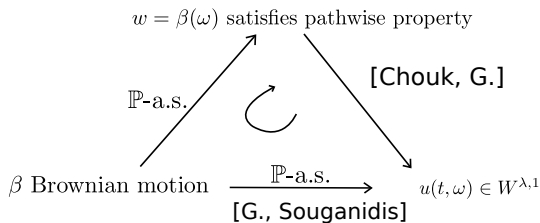
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- Again: Results are given for general truly nonlinear flux A .
- How to classify pathwise properties of w leading to improved regularity?



Idea of the proof

- Ideas of the proof of regularity for

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- Kinetic formulation:

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- Change of variables gives

$$\chi(t, x, v) = \chi_0(x + v\beta_t, v) + \int_0^t \partial_v m(s, x + v(\beta_t - \beta_s), v) ds.$$

Idea of the proof

- Averaging over velocity

$$u(t, x) = \int_v \chi = \int_v \chi_0(x + v\beta_t, v) dv + \int_0^t \int_v \partial_v m(s, x + v(\beta_t - \beta_s), v) dv ds.$$

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- Rigorously, this can be seen by Fourier transform, that is,

$$\hat{u}(t, n) = \int_v e^{-iv\beta_t n} \hat{\chi}_0(n, v) dv + \int_0^t \int_v e^{-iv(\beta_t - \beta_s)n} \partial_v \hat{m}(s, n, v) dv ds.$$

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- The oscillatory integrals have a regularizing effect, both in v and in $\beta_t - \beta_s$.

Framework

- For SDE this has been considered by [Catellier, Gubinelli; *SPA*, 2016]: A path $w \in C(\mathbb{R}_+; \mathbb{R}^d)$ is said to be (ρ, γ) -irregular if

$$\left| \int_s^t e^{i w_r \cdot n} dr \right| \lesssim (1 + |n|)^{-\rho} |t - s|^\gamma \quad \forall n \in \mathbb{R}^d, s < t.$$

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- Note:

$$\int_s^t e^{i w_r \cdot n} dr = \int_{\mathbb{R}} e^{i x \cdot n} dL_w^{s,t}(x) = L_w^{\hat{s},t}(n)$$

the Fourier transform of the local time.

Main result

Theorem

Let $w \in C^\eta([0, T], \mathbb{R}^d)$ for some $\eta > 0$ be (ρ, γ) -irregular, u a quasi-solution solution to

$$\partial_t u + \frac{1}{2} \partial_x u^2 \circ dw_t = 0 \quad \text{on } (0, T) \times \mathbb{T}.$$

Then, for all

$$\lambda < \frac{\rho(\eta + 1) - (1 - \gamma)}{(\rho \vee 1)(\eta + 1) + (1 - \gamma)},$$

we have

$$u \in L^1([0, T]; W^{\lambda, 1}(\mathbb{T})).$$

Corollary

Let β^H be a fractional Brownian motion with Hurst parameter $H \in (0, \frac{1}{2}]$ and u be a quasi-solution to

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Then, \mathbb{P} -a.s. for all $\lambda < \frac{1}{1+2H}$,

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- Note: Fully recover the probabilistic result from [G., Souganidis; CPAM, 2016]: For $H = \frac{1}{2}$ get $\lambda < \frac{1}{2}$.

A path-by-path scaling condition

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Discussion of the path classification

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- However: (ρ, γ) -irregularity depends on two parameters, also encoding a time regularity. Hence, does not seem to be optimal.
- Moreover: (ρ, γ) -irregularity not easy to check.
- To avoid the use of oscillatory integrals: Completely avoid Fourier methods in the proof.

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- Rewrite as, for $\lambda > 0$,

$$\partial_t \chi + v \partial_x \chi \circ dw_t + \lambda \chi = \partial_v m + \lambda \chi.$$

Idea of the proof

- Change of variables, take velocity integral and drop i.c.:

$$u(t, x) = \int_v \chi(t, x, v) dv = \int_0^t \int_v e^{-\lambda(t-s)} (\partial_v m)(s, x - vw_{s,t}, v) dv ds \\ + \lambda \int_0^t \int_v e^{-\lambda(t-s)} \chi(s, x - vw_{s,t}, v) dv ds.$$

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- Introduce the random X-ray transform

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where m is a finite measure and $\chi(t, x, v) := 1_{[0, u(t, x)]}(v)$.

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- Strategy: Estimate the regularity of $T(\partial_v m)$, $T\chi$ then use real interpolation.

Path-by-path scaling condition

- This leads to: *Path-by-path scaling condition*: Assume that there is a $\iota \in (0,1]$ such that for every $\sigma \in [0,1)$, $\lambda \geq 1$ we have

$$\int_0^T \int_0^{T-r} e^{-\lambda t} \underbrace{|w_{t+r} - w_r|}_{=: w_{r,r+t}}^{-\sigma} dt dr \lesssim \lambda^{-1+\iota\sigma}.$$

Path-by-path scaling condition

- This leads to: *Path-by-path scaling condition*: Assume that there is a $\iota \in (0,1]$ such that for every $\sigma \in [0,1)$, $\lambda \geq 1$ we have

$$\int_0^T \int_0^{T-r} e^{-\lambda t} \underbrace{|w_{t+r} - w_r|}_{=: w_{r,r+t}}^{-\sigma} dt dr \lesssim \lambda^{-1+\iota\sigma}.$$

- Easy to see: (ρ, γ) -irregularity implies path-by-path scaling.

Theorem

Let u be a quasi-solution to

$$\partial_t u + \frac{1}{2} \partial_x u^2 \circ dw_t = 0 \quad \text{on } \mathbb{R}$$

and suppose that $w \in C^\eta$ satisfies path-by-path scaling. Then, for all $\lambda < \frac{1+\eta-l}{1+\eta+l}$,

$$u \in L^1([0, T]; W^{\lambda,1}(\mathbb{T})).$$