

Well-posedness by noise for scalar conservation laws

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[G., Maurelli.; CPDE 2019+], [Gassiat, G., Lions, Souganidis.; arxiv 2018],
[Smith, G.; JMPA 2019+], [Gassiat, G.; PTRF 2018].

Outline

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- 2 Well-posedness by noise for nonlinear stochastic scalar conservation laws
- 3 Well-posedness by nonlinear noise
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- 5 Speed of propagation for stochastic Hamilton-Jacobi equations

- Classical well-posedness for ODE:

$$\begin{aligned}dX_t^x &= b(X_t^x)dt \\ X_0^x &= x\end{aligned}$$

is well-posed if b is sufficiently smooth, e.g. Lipschitz continuous.

- In contrast, well-posedness for SDE: ($\sigma > 0$)

$$\begin{aligned}dX_t^x &= b(X_t^x)dt + \sigma d\beta_t \\ X_0^x &= x\end{aligned}$$

has a unique solution if b is bounded, measurable. This is called '*well-posedness by noise*'.

- Consider

$$du + b(x) \cdot \nabla u dt = \sigma \nabla u \circ d\beta_t. \quad (\text{STE})$$

- Characteristics for (STE):

$$\begin{aligned} dX_t^x &= b(X_t^x) dt - \sigma d\beta_t \in \mathbb{R}^d \\ X_0^x &= x. \end{aligned}$$

- Two (related) effects: regularization by noise, well-posedness by noise.
- Well-posedness by noise [Flandoli, Gubinelli, Priola; 2010]: Weak solutions to (STE) are unique.
- Regularization by noise [Flandoli, Fedrizzi; 2013]: If u_0 is smooth then $u(t)$ is smooth.

- Left open: What about the nonlinear case, e.g. Burgers?
- Linear multiplicative noise does not help anymore:

$$du + \partial_x u^2 dt = \partial_x u \circ d\beta_t.$$

Then: $v(t, x) := u(t, x - \beta_t)$ is a solution to

$$\partial_t v + \partial_x v^2 = 0.$$

- In particular, weak solutions are non-unique.
- Conclusion in [Flandoli, Gubinelli, Priola; *Invent. Math.*, 2010]:
„It is very easy to produce examples [...] for a stochastic version of Euler equation which show that the particular noise we use does not have any regularizing effect in this case. [...] The generalization to nonlinear transport equations, where b depends on u itself, would be a major next step for applications to fluid dynamics but it turns out to be a difficult problem.“

Well-posedness by noise for stochastic scalar conservation laws

- Consider

$$\partial_t u + b(x, u) \cdot \nabla u = 0.$$

- Scalar conservation laws with irregular flux: Traffic flows, sedimentation processes
[De Philippis et al., *CPDE*, 2015; Andreianov, Karlsen, Risebro, *ARMA*, 2011].
- In this talk: For simplicity consider

$$\partial_t u + b(x) \cdot \nabla(u^2) = 0,$$

for irregular b (in particular $\operatorname{div} b \notin L^\infty$).

- The deterministic problem is ill-posed in general: Entropy solutions are non-unique.

- Model example: Consider

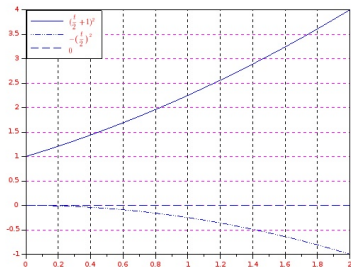
$$\begin{aligned} \partial_t u + b(x) \cdot \nabla(u^2) &= 0 \\ u(0, x) &= 1_{[0,1]}(x) \end{aligned} \quad (*)$$

with $b(x) = 2\operatorname{sgn}(x)(\sqrt{|x|} \wedge K)$.

- Rankine-Hugoniot implies:

$$u^1(t, x) := \begin{cases} 1 & \text{für } 0 \leq x \leq \left(\frac{t}{2} + 1\right)^2 \\ 0 & \text{sonst,} \end{cases}$$

$$u^2(t, x) := \begin{cases} 1 & \text{für } -\left(\frac{t}{2}\right)^2 \leq x \leq \left(\frac{t}{2} + 1\right)^2 \\ 0 & \text{sonst.} \end{cases}$$



- Can we restore well-posedness by adding a linear multiplicative noise term?
- Non-trivial: shocks due to the nonlinearity and shocks due to the irregularity of b may combine in such a way that this noise may be insufficient.
- Stochastic Burgers' equation:

$$du + b(x) \cdot \nabla(u^2)dt = \nabla u \circ d\beta_t.$$

Theorem

Assume $b \in (W^{1,1} \cap L^\infty)(\mathbb{R}^d)$ and $\operatorname{div} b \in L^p(\mathbb{R}^d)$ for some $p > d$. For $u_0 \in (L^1 \cap L^\infty)(\mathbb{R}^d)$, the stochastic Burgers' equation admits a unique entropy solution.

Model example: $b(x) = 2\operatorname{sgn}(x)(\sqrt{|x|} \wedge K)$.

- Kinetic solutions: Given $u = u(t, \omega, x)$, introduce a new (velocity) variable $\xi \in \mathbb{R}$ and define the kinetic function:

$$f = f[u](t, \omega, x, \xi) = 1_{\xi < u(t, x)} - 1_{\xi < 0}.$$

- Then u is an entropy solution iff

$$\partial_t f + 2b(x)\xi \cdot \nabla_x f - \nabla_x f \circ d\beta_t = \partial_\xi m.$$

Definition

$f = f(t, \omega, x, \xi)$ is a generalized entropy solution if f solves the kinetic equation

$$\partial_t f + 2b(x)\xi \cdot \nabla_x f - \nabla_x f \circ d\beta_t = \partial_\xi m$$

and, for some nonnegative measure ν ,

$$|f| = \text{sgn}(\xi)f \leq 1 \quad \partial_\xi f = \delta_0 - \nu.$$

- Set of generalized entropy solutions is convex and closed under weak limits.

Proposition

Let $u_0 \in (L^1 \cap L^\infty)(\mathbb{R}^d)$ and $b, \text{div}b \in L^1(\mathbb{R}^d)$. Then there exists a generalized entropy solution.

- Proof of uniqueness:
 - First step (deterministic): Via renormalization arguments derive an inequality for $|f| - f^2$ (similar to the kinetic equation).
 - Second step (stochastic): Take the expectation and use *parabolic* theory.
- Kinetic equation

$$\partial_t f + 2b(x)\xi \cdot \nabla_x f - \nabla_x f \circ d\beta_t = \partial_\xi m$$

rewritten in Itô form:

$$\partial_t f + 2b(x)\xi \cdot \nabla_x f - \nabla_x f d\beta_t - \frac{1}{2} \Delta_x f = \partial_\xi m.$$

A Laplacian appears, which suggests regularization. Note: the equation is hyperbolic (not parabolic: no regularization of initial datum).

- Taking the expectation yields

$$\partial_t \mathbb{E}f + 2b(x)\xi \cdot \nabla_x \mathbb{E}f - \frac{1}{2} \Delta_x \mathbb{E}f = \partial_\xi \mathbb{E}m,$$

i.e. a parabolic PDE.

Proposition

Fix $T > 0$ and assume $b \in L^\infty(\mathbb{R}^d)$, $\operatorname{div} b$ in $L^p(\mathbb{R}^d)$ for some $p > d$. Then, there exists a $\varphi \geq 0$, independent of ξ , with $\varphi_T \sim 1$, such that

$$\partial_t \varphi - 2 \operatorname{div}_x (b(x) \xi \varphi) + \frac{1}{2} \Delta_x \varphi \leq C$$

for some $C > 0$ (independent of T).

Theorem

Assume $b \in (W^{1,1} \cap L^\infty)(\mathbb{R}^d)$ with $\operatorname{div} b \in L^p$ for some $p > d$. Then

- 1 Every generalized entropy solution is an entropy solution.
- 2 For $u_0 \in (L^1 \cap L^\infty)(\mathbb{R}^d)$ there is an entropy solution.

Entropy solutions are unique

Well-posedness by nonlinear noise

- Consider

$$\partial_t u + \nabla \cdot (b(t, x)u) + \nabla B(u) \circ d\beta_t = 0$$

on $[0, T] \times \mathbb{R}^d$.

- The noise $\nabla B(u) \circ d\beta_t$ may degenerate: $b(u) := B'(u)$ may vanish for some values of u .

We assume

- [Renormalization] $b \in L_t^1(W_x^{1,1})$ and $\operatorname{div} b \in L_{t,x}^r$ for some $r > 1$.
- [Sub-criticality] $b \in L_{t,x}^\infty$ and $(\operatorname{div} b)_- \in L_{t,x}^q$ for some $q > d + 2$. Moreover, there exists an $R > 0$ such that $\operatorname{supp} b \subseteq [0, T] \times B_R$.
- [Asymptotic ellipticity] $B \in C^3(\mathbb{R}; \mathbb{R})$. For $b(\rho) := B'(\rho)$, there exist strictly positive λ and Λ such that

$$\sup_{\rho \in \mathbb{R}} b^2(\rho) \leq \Lambda \quad \liminf_{|\rho| \rightarrow \infty} b^2(\rho) = \lambda.$$

Theorem

Under the above assumptions: For $u_0 \in L^1 \cap L^\infty(\mathbb{R}^d)$ there exists a unique kinetic solution to starting from u_0 .

Regularization by noise for stochastic Hamilton-Jacobi equations

Regularization by noise for stochastic Hamilton-Jacobi equations

- Can we use nonlinear noise to regularize nonlinear PDE?
- Model example: Porous medium equation

$$\partial_t w = \frac{1}{6} \partial_{xx} w^3, \quad \text{on } \mathbb{R}$$

with smooth, compactly supported initial condition.

- Solutions are known to develop singularities:

$$\|\partial_x w(t)\|_{L^\infty} = \infty$$

for all $t > 0$ large enough.

- Linear multiplicative noise does not help:

$$dv = \frac{1}{6} \partial_{xx} v^3 dt + \sigma \partial_x v \circ d\beta_t, \quad \text{on } \mathbb{R}.$$

Then $w(t, x) = v(t, x - \sigma\beta_t)$.

- Instead, consider, for $\sigma > 0$,

$$dv = \frac{1}{6} \partial_{xx} v^3 dt + \sigma \frac{1}{2} \partial_x v^2 \circ d\beta_t, \quad \text{on } \mathbb{R}. \quad (\text{SPME})$$

- Note: If u is the viscosity solution to

$$du = \frac{1}{6} \partial_x (\partial_x u)^3 dt + \sigma \frac{1}{2} (\partial_x u)^2 \circ d\beta_t, \quad \text{on } \mathbb{R},$$

then, $v = \partial_x u$ solves (SPME).

- General framework: Consider

$$du = F(t, x, u, Du, D^2u) dt + \frac{1}{2} |Du|^2 \circ d\beta_t, \quad \text{on } \mathbb{R}^N,$$

and compare to the regularity of non-perturbed problem

$$dw = F(t, x, w, Dw, D^2w), \quad \text{on } \mathbb{R}^N.$$

- F satisfies the usual assumptions from the theory of stochastic viscosity solutions

- Control on the rate of loss of semiconcavity: There is a $V_F \in Lip_{loc}(\mathbb{R}_+ \setminus \{0\})$ such that, for $\ell_0 > 0$,

$$D^2 g \leq \frac{Id}{\ell_0} \Rightarrow D^2(S_F(t, g)) \leq \frac{Id}{\ell(t)},$$

where $t \mapsto S_F(t, g)$ denotes the solution to

$$\begin{aligned} \partial_t w + F(t, x, w, Dw, D^2 w) &= 0 \\ w(0) &= g \end{aligned}$$

and ℓ the solution to

$$\begin{aligned} \dot{\ell}(t) &= V_F(\ell(t)) \\ \ell(0) &= \ell_0. \end{aligned}$$

Theorem

Let u be the solution to

$$\partial_t u = F(t, x, u, Du, D^2 u) + \frac{1}{2} |Du|^2 \circ d\beta_t,$$

with

$$D^2 u_0 \leq \frac{Id}{\ell_0} \quad \text{some } \ell_0 \in [0, \infty).$$

Then for each $t \geq 0$, one has

$$D^2 u(t, \cdot) \leq \frac{Id}{L_t},$$

where L is the maximal solution to

$$dL_t = V_F(L_t) dt + d\beta_t, \quad \text{with } L_t \geq 0, \\ L_0 = \ell_0.$$

- Return to the model example

$$dv = \frac{1}{6} \partial_{xx} v^3 dt + \sigma \frac{1}{2} \partial_x v^2 \circ d\beta_t, \quad \text{on } \mathbb{R}.$$

- Deterministic case: $\|\partial_x w(t)\|_\infty = \infty$ for all $t > 0$ large enough.
- We have the sharp bound

$$\|\partial_x v(t)\|_\infty \leq \frac{1}{L^+(t) \wedge L^-(t)},$$

where L^\pm solve

$$dL^\pm = -\frac{1}{L^\pm(t)} dt \pm \sigma d\beta_t, \quad \text{with } L_t^\pm \geq 0,$$

$$L^\pm(0) = \frac{1}{\|(\partial_x v_0)_\pm\|_\infty}.$$

- In conclusion,
 - If $\sigma^2 > 2$: For all $t \geq 0$, \mathbb{P} -a.s.

$$v(t) \in W^{1,\infty}$$

- If $\sigma^2 \leq 2$: \mathbb{P} -a.s. for all $t > 0$ large enough

$$v(t) \notin W^{1,\infty}$$

- Observation: As long as the solution is Lipschitz continuous, it is time reversible.

Speed of propagation for stochastic Hamilton-Jacobi equations

Speed of propagation for stochastic Hamilton-Jacobi equations

- Consider

$$du = H(Du, x) \cdot d\xi \quad \text{in } \mathbb{R}^d \times (0, T] \quad u(\cdot, 0) = u_0 \quad \text{in } \mathbb{R}^d,$$

with $H: \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$ convex and Lipschitz continuous in the first argument, $\xi \in C_0([0, T])$.

- By Lions-Souganidis there exists a unique viscosity solution.
- Given $T > 0$ let

$$\rho_H(\xi, T) := \sup \left\{ R \geq 0 : \text{there exist solutions } u^1, u^2 \text{ and } x \in \mathbb{R}^d, \right. \\ \left. \text{such that } u^1(\cdot, 0) = u^2(\cdot, 0) \text{ in } B_R(x) \text{ and } u^1(x, T) \neq u^2(x, T) \right\},$$

where $B_R(x)$ is the ball in \mathbb{R}^d centered at x with radius R .

- Classical (e.g. Crandall, Lions): If ξ is a BV-path, then

$$\rho_H(\xi, T) \leq L \|\xi\|_{TV([0, T])},$$

where $\|\xi\|_{TV([0, T])}$ is the total variation semi-norm of ξ and L is the Lipschitz constant of H . It is easy to see that this is sharp when $\xi \equiv 1$.

- General continuous signal ξ : If $H(p, x) = H_1(p) - H_2(p)$, where H_1, H_2 convex, Lipschitz with Lipschitz constant L and $H_1(0) = H_2(0) = 0$, then, for any constant A , if

$$u(0, \cdot) \equiv A \text{ on } B_R(0),$$

then

$$u(t, \cdot) \equiv A \text{ on } B_{R(t)}(0), \quad \text{for } R(t) := R - L \left(\max_{s \in [0, t]} \xi(s) - \min_{s \in [0, t]} \xi(s) \right).$$

- This does not imply a finite domain of dependence.
- In fact: For $H(p) = |p_1| - |p_2|$ equality is attained in

$$\rho_H(\xi, T) \leq L \|\xi\|_{TV([0, T])},$$

for all continuous ξ .

- No finite domain of dependence if $\xi \notin BV([0, T])$

Given $\xi \in C_0([0, T])$, the sequence $(\tau_i)_{i \in \mathbb{Z}}$ of successive extrema of ξ :

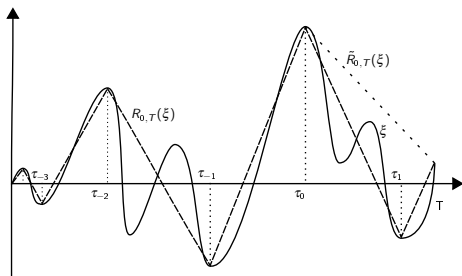


Figure: The (fully) reduced path

Definition

- (i) The reduced path $R_{0,T}(\xi)$ is a piecewise linear function which agrees with ξ on $(\tau_i)_{i \in \mathbb{Z}}$.
- (ii) The fully reduced path $\tilde{R}_{0,T}(\xi)$ is a piecewise linear function agreeing with ξ on $(\tau_{-i})_{i \in \mathbb{N}} \cup \{T\}$.

Recall:

$$du = H(Du, x) \cdot d\xi \quad \text{in } \mathbb{R}^d \times (0, T] \quad u(\cdot, 0) = u_0 \quad \text{in } \mathbb{R}^d.$$

Let $S^\xi(s, t)$ be the nonlinear semigroup of solutions.

Theorem

Let $\xi \in C_0([0, T])$. Then

$$S^\xi(0, T) = S^{R_{0,T}(\xi)}(0, T).$$

Hence

$$\rho_H(\xi, T) \leq L \|R_{0,T}(\xi)\|_{TV([0,T])}.$$

Theorem

Let B be a Brownian motion, $T > 0$. Then, a.s.

$$\|R_{0,T}(B)\|_{TV([0,T])} < \infty.$$

Sharpness of the upper bound:

Theorem

Let $H(p) = |p|$ on \mathbb{R}^d with $d \geq 1$. Then, for all $T > 0$ and $\xi \in C_0([0, T])$,

$$\rho_H(\xi, T) \geq \|\tilde{R}_{0, T}(\xi)\|_{TV([0, T])}.$$

When $d = 1$, then

$$\rho_H(\xi, T) = \|\tilde{R}_{0, T}(\xi)\|_{TV([0, T])}.$$

Proof.

By keeping track of explicit cancellations. □

Sharp speed of propagation in higher dimension?

Consider $H(p) = |p|$ and

$$\dot{\xi}(t) = \begin{cases} 4 & \text{for } (0,1) \\ -2 & \text{on } (1,2) \\ 1 & \text{on } (2,3) \end{cases} \quad \text{and} \quad \ddot{R}_{0,3}(\xi) = \begin{cases} 4 & \text{on } (0,1) \\ -\frac{1}{2} & \text{on } (1,3); \end{cases}$$

Which gives

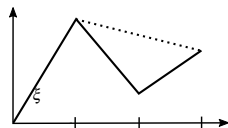


Figure: Evolution of $S_{|\cdot|}^{\xi}(0, t)$ at $t=0, t=1, t=2, t=3$



Figure: Evolution of $S_{|\cdot|}^{\ddot{R}(\xi)}(0, t)$ at $t=0, t=1, t=3$

Let $\delta_1 > \delta_2 > \delta_3 > 0$ and ξ continuous on $[0, 3]$ with



$$\xi = \begin{cases} \delta_1 & \text{on } (0, 1), \\ -\delta_2 & \text{on } (1, 2), \\ +\delta_3 & \text{on } (2, 3). \end{cases}$$

Then

$$\rho_{|\cdot|}(\xi, 3) = \delta_1 + \delta_2 + \delta_3 = \|\xi\|_{TV([0,3])} = \|R_{0,T}(\xi)\|_{TV([0,T])} > \|\tilde{R}_{0,T}(\xi)\|_{TV([0,T])}.$$

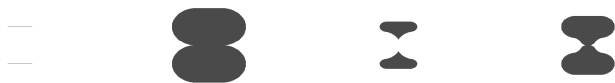


Figure: Evolution of $S_{|\cdot|}^{\xi}(0, \cdot)P_1$



Figure: Evolution of $S_{|\cdot|}^{\xi}(0, \cdot)P_2$