

# Stochastic scalar conservation laws

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[G., Souganidis; CMS, 2014], [G., Souganidis; arXiv, 2015].

# Outline

- 1 Motivation
- 2 Long-time behavior
- 3 Regularization by noise

# Motivation

## Motivation

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- We will consider PDE driven by a 'rough' signal  $z$  of the type

$$du + \operatorname{div}(A(x, u) \circ dz) = 0.$$

If  $A$  is a diagonal matrix this becomes

$$du + \sum_{j=1}^N \partial_{x_j} A_j(x, u) \circ dz_j = 0$$

- In this talk, we mainly consider

$$du + \sum_{j=1}^N \partial_{x_j} A_j(u) \circ d\beta^j = 0.$$

- For example, stochastic Burgers' equation

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

# Motivation

- The motivation comes from two directions: Relation to Hamilton-Jacobi equations, mean-field games.
- Mean-field games going back to Lasry, Lions: Consider the SDE

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

for  $i = 1, \dots, L$ .

- Then the empirical law of  $X$  converges to a measure  $\pi_t$  with density  $m_t$  which evolves according to

$$dm + \operatorname{div}(\sigma^*(x, m) \circ d\beta) = 0.$$

In general  $\sigma^*$  is not a diagonal matrix.

# Well-posedness

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# Well-posedness

- Spatially inhomogeneous case:

$$du + \sum_{j=1}^N \partial_{x_j} A_j(x, u) \circ d\beta^j = 0.$$

- Pass to the kinetic formulation: Set

$$\chi(t, x, \xi) := 1_{[0, u(t, x)]}(\xi).$$

# Well-posedness

- Then (informally)

$$d\chi + \sum_{j=1}^N (\partial_u A_j)(x, \xi) \partial_{x_j} \chi \dot{\beta}^j + \left( \sum_{j=1}^N \operatorname{div}_{x_j} A(x, \xi) \dot{\beta}^j \right) \partial_\xi \chi = \partial_\xi m.$$

- Test against solutions to

$$d\varphi - \sum_{j=1}^N (\partial_u A_j)(x, \xi) \partial_{x_j} \varphi \dot{\beta}^j - \left( \sum_{j=1}^N \operatorname{div}_{x_j} A(x, \xi) \dot{\beta}^j \right) \partial_\xi \varphi = 0.$$

Then

$$d \int_{R^{N+1}} \chi \varphi dx d\xi = - \int_{R^{N+1}} \partial_\xi \varphi dm. \quad (*)$$

- Roughly speaking:  $u$  is a pathwise entropy solution if  $\chi$  satisfies (\*).

Theorem (Gess, Souganidis; CMS, 2015)

*Pathwise entropy solutions are well-posed.*



# Stochastic Burgers' equation

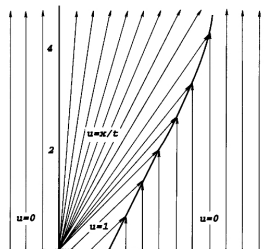
**Stochastic Burgers' equation - a simple example**

# Stochastic Burgers' equation

- e.g. Burgers' equation

$$\partial_t u + \frac{1}{2} \partial_x u^2 = 0$$

$$u(0) = 1_{[0,1]}$$



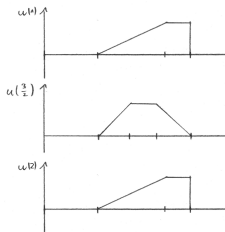
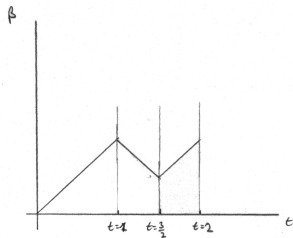
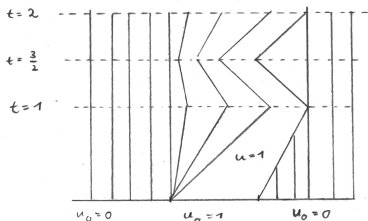
# Stochastic Burgers' equation

- Inhomogeneous case:

$$\partial_t u + \frac{1}{2} \partial_x u^2 \circ d\beta = 0$$

$$u(0) = 1_{[0,1]}$$

- Solution  $u$ :



# Long-time behavior

## Long-time behavior

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# Long-time behavior

- We aim to analyze the long-time behavior of

$$\partial_t u + \operatorname{div} A(u) = 0$$

and

$$du + \sum_{j=1}^N \partial_{x_j} A_j(u) \circ d\beta_j = 0$$

on the torus  $\mathbb{T}^N$ .

- We will show

$$u(t) \rightarrow \bar{u}_0 = \int_{\mathbb{T}^N} u_0(x) dx \quad \text{for } t \rightarrow \infty$$

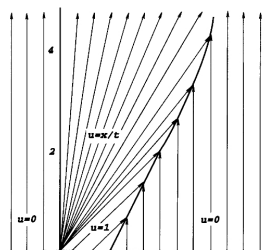
in  $L^1(\mathbb{T}^N)$ .

# Some results from the deterministic case

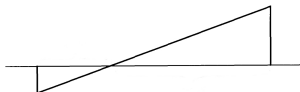
- Recall: Burgers' equation

$$du + \frac{1}{2} \partial_x u^2 = 0$$

$$u(0) = 1_{[0,1]}$$



- For large times get asymptotic shape: “N-wave”



## Some results from the deterministic case

Some existing results:

- $N = 1$ : [Lax; *CPAM*, 1957]. Rate for strictly convex flux:

$$\|u(t) - \bar{u}_0\|_\infty \lesssim t^{-1}.$$

- $N = 2$ : [Engquist, E; *CPAM*; 1993]. Rate for strictly convex flux:

$$\|u(t) - \bar{u}_0\|_\infty \lesssim t^{-1}.$$

- $N \geq 1$ : [Chen, Frid; *ARMA*; 1999], [Chen, Perthame; *Proc. AMS*; 2009]. If  $A$  is 'genuinely nonlinear' then

$$\|u(t) - \bar{u}_0\|_1 \rightarrow 0 \quad \text{for } t \rightarrow \infty.$$

- Open problem: Rate of convergence for  $N \geq 3$ ?

## (New) rates for the deterministic case

- Assume that the flux  $A$  is *genuinely nonlinear*: there exist  $\theta \in (0, 1]$  and  $C > 0$  such that, for all  $\sigma \in S^{N-1}$ ,  $z \in \mathbb{R}$  and  $\varepsilon > 0$ ,

$$|\{\xi \in \mathbb{R} : |A'(\xi) \cdot \sigma - z| \leq \varepsilon\}| \leq C\varepsilon^\theta.$$

For example: For  $A$  strictly convex,  $N = 1$  we have  $\theta = 1$ .

- Let  $u$  be the unique entropy solution to

$$\partial_t u + \sum_{j=1}^N \partial_{x_j} A_j(u) = 0.$$

Theorem (G., Souganidis; 2015)

For  $t \geq 1$  and  $u_0 \in L^2(\mathbb{T}^N)$ ,

$$\|u(\cdot, t; \cdot, u_0) - \bar{u}_0\|_1 \leq t^{-\frac{\theta}{2+\theta}} (\|u_0\|_2^2 + 1).$$



## (New) rates for the stochastic case

- Let us return to

$$du + \sum_{j=1}^N \partial_{x_j} A_j(u) \circ d\beta_j = 0.$$

- Again assume that  $A$  is genuinely nonlinear.

Theorem (G., Souganidis; 2015)

For  $t \geq 1$  and  $u_0 \in L^2(\mathbb{T}^N)$ ,

$$\mathbb{E} \|u(\cdot, t; \cdot, u_0) - \bar{u}_0\|_1 \leq (\|u_0\|_2^2 + 1) t^{-\frac{\theta}{3+\theta}}.$$

- E.g.  $\theta = 1$ : deterministic rate  $t^{-\frac{1}{3}}$ , stochastic rate  $t^{-\frac{1}{4}}$ . But: No claim of optimality.
- Note: Brownian motion scales like  $\sqrt{t}$ , which “slows down” characteristics.

# Regularization by noise

## Regularization by noise

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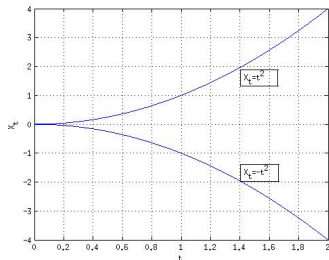
# Regularization by noise - linear case

- We recall: Consider

$$du + b(x) \cdot \nabla u = 0 \quad (\text{TE})$$

for non-Lipschitz  $b$  (but, say, Hölder continuous). E.g.  $b(x) = \text{sgn}(x)\sqrt{|x|}$ .

- In general, characteristics collide causing shocks (i.e. discontinuities). Solution is not better than  $u(t) \in BV$  even if  $u_0$  is smooth.
- Weak solutions are non-unique: e.g.  $b(x) = \text{sgn}(x)\sqrt{|x|}$



- Question: Can noise restore uniqueness or increase regularity?

# Regularization by noise - linear case

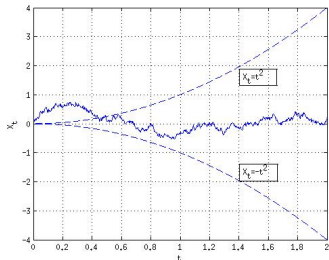
- Consider

$$du + b(x) \cdot \nabla u = -\nabla u \circ dW_t. \quad (\text{STE})$$

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

$$du + b(x) \cdot \nabla u = -\nabla u \circ dW_t$$

$$b(x) = \text{sgn}(x) \sqrt{|x|}$$



- Entirely open: What about the nonlinear case, e.g. Burgers?

## Quasi-solutions and averaging

- Consider the Burgers' equation

$$\partial_t u + \partial_x u^2 = 0 \quad \text{on } \mathbb{T} \quad (\text{B})$$

- Weak solutions to (B) are not unique.
- We consider quasi-solutions [De Lellis, Otto, Westdickenberg; *ARMA*, 2003]: A weak solution  $u$  to (B) is a quasi-solution, if for some Radon measure  $m$

$$\partial_t \chi + \xi \partial_x \chi = \partial_\xi m.$$

Quasi-solutions to (B) are not unique.

- [De Lellis, Westdickenberg; *AHP*, 2003]: There is a quasi-solution to (B) such that

$$u(t) \notin W^{1,\lambda} \quad \text{for all } \lambda > \frac{1}{3}.$$

- Question: Does noise improve the situation?

## (New) results for the stochastic case

- Consider the stochastic Burgers' equation

$$du + \partial_x u^2 \circ d\beta_t = 0 \quad \text{on } \mathbb{T} \quad (\text{SB})$$

Theorem (G., Souganidis; 2015)

Let  $u$  be a pathwise quasi-solution to (SB). Then,  $t > 0$ ,

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

- Thus: quasi-solutions to (SB) are more regular than to (B), i.e. regularization by noise.

# Thanks

**Thanks!**