

# REACTION-DIFFUSION EQUATIONS

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## WORKSHOP IN STOCHASTIC ANALYSIS

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Joint work with

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# Acknowledgements



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- **Well posed when  $b$  and  $\sigma$  are Lipschitz continuous (Donati-Martin and Pardoux, 1993)**

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  - 5 **The maximum principle produces many more solutions.**

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**Theorem (Bonder+Groisman, 2009)**

*If  $\sigma > 0$  and  $b = \text{convex}$ , then a.s.  $\exists$  finite-time blowup.*

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- Whence so does  $\sup_{x \in [0,1]} u(t, x) \geq X_t$ . QED

# Optimality of the Bonder–Groisman Theorem

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- If  $\sigma : \mathbb{R} \rightarrow \mathbb{R}$  is locally Lipschitz then let

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## Theorem (Dalang+K+Zhang, 2018)

Suppose  $u(0) \in \mathbb{C}_0^\alpha$  for some  $\alpha > 0$ ,  $\sigma$  and  $b$  are locally Lipschitz, and

$$|b(z)| = O(|z| \log |z|) \quad \text{and} \quad K_N^\sigma = O(|\log N|^{1/4}) \quad \text{as } N, |z| \rightarrow \infty.$$

Then, the SPDE has a unique continuous “random field solution.”

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- We have a second set of conditions for optimality of  $b \in L \log L$ . Leads to a conditional result that uses the sharp form of Gross’ log-Sobolev inequality for Lebesgue measure

# Optimality of the Bonder–Groisman Theorem (Proof outline)

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- Suffice it to say that this condition ensures *a priori* “optimal regularity”:

$$u(0) \in \mathbb{C}_0^\alpha \quad \Rightarrow \quad \mathbb{P} \{u(t) \in \mathbb{C}_0^\alpha \quad \forall t > 0\} = 1.$$

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- **In fact, we develop delicate moment bounds:  $\exists \varepsilon, \delta \ll 1$ :**

$$\mathbb{E} \left( \sup_{t \in [0, \varepsilon]} \sup_{x \in [0, 1]} |U_N(t, x)|^k \right) = O(N^{k\delta}) \quad \forall k \geq 2.$$

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- The same if  $b(z) = c_1 \pm c_2 |z| \log_+ |z|$ . Now use **comparison** and **SMP** to finish. QED

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- 3 For all  $\phi \in C_0^1([0, 1])$ ,

$$\begin{aligned} \int_0^1 u(t, x) \phi(x) dx &= \int_0^1 u_0(x) \phi(x) dx + \frac{1}{2} \int_0^1 u(s, x) \phi''(x) dx \\ &+ \int_{(0, t) \times (0, 1)} b(u(s, x)) \phi(x) ds dx + \int_{(0, t) \times (0, 1)} \sigma(u(s, x)) \phi(x) \xi(ds dx), \end{aligned}$$

a.s. on  $\{\tau > t\}$ .

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## Theorem (Dalang+K+Zhang, 2018)

Suppose  $u(0) \in L^2[0, 1]$  is nonrandom and  $\sigma$  is bounded. Then,  $\mathbb{P}\{\tau = \infty\} = 1$  for every  $\mathbb{L}_{loc}^2$  solution  $u$ . Moreover,

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- We don't know if any such solution exists.

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Among other things, this theorem uses the following form of the log-Sobolev inequality for the law  $\text{Unif}[0, 1]$ :

## Theorem (Gross, 1993)

For every  $h \in C_0^1([0, 1])$  and  $\varepsilon \in (0, 1)$ ,

$$\int_0^1 |h(x)|^2 \log |h(x)| dx \leq \varepsilon \|h'\|_{\mathbb{L}^2}^2 + \frac{1}{4} \log(1/\varepsilon) \|h\|_{\mathbb{L}^2}^2 + \|h\|_{\mathbb{L}^2}^2 \log(\|h\|_{\mathbb{L}^2}^2),$$

where  $0 \log 0 := 0$ .