

Regularity structures for singular SPDEs

Joint work with M. Hoshino & S. Kusuoka

1 What is singular?

► **Theorem** – *The product map*

$$(a, b) \in C^\alpha \times C^\beta \mapsto ab \in C^{\alpha \wedge \beta}$$

is well-defined and continuous iff $\alpha + \beta > 0$.

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– Heat equation with white noise potential

$$(\partial_t - \Delta)u = \zeta u,$$

for a space white noise $\zeta = \zeta(x)$ on a 2-dimensional torus: ζ is $(-1)^-$ regular.

– The 2-dimensional Φ^4 **equation**

$$(\partial_t - \Delta)u = \xi - u^3,$$

for a spacetime white noise $\xi = \xi(t, x)$ on 2-dimensional space torus that is $(-2)^-$ (parabolic) regular.

– The 1-dimensional **(gKPZ) equation**

$$(\partial_t - \partial_x^2)u = f(u)\xi + g(u)(\partial_x u)^2,$$

for a spacetime white noise $\xi = \xi(t, x)$ on 1-dimensional space torus that is $(-3/2)^-$ (parabolic) regular.

2 The mantra of singular SPDEs

Adopt a language in which we disentangle the task of solving the PDE from the task of making sense of some ill-defined products.

Mantra – *For a given equation, if one can make sense of the a priori ill-defined products when u is replaced by some reference random objects then one can make sense of these products when u “looks like” the reference objects.*

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► **Theorem** – *There is a large class of equations where one can build some random reference objects in a meaningful way – this is what renormalization is about \rightsquigarrow Masato’s talk.*

3 RS & local consistent descriptions

We use **reference objects** to describe f locally

$$f(\cdot) \simeq \sum_{\tau \in \mathcal{B}} f_{\tau}(x) (\Pi_x \tau)(\cdot), \text{ near each } x,$$

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some other reference objects for the local behaviour of the 'derivatives' f_{τ}

$$f_{\tau}(y) \simeq \sum_{\sigma \in \mathcal{B}^+} f_{\tau\sigma}(x) g_x(\sigma)(y), \text{ near each } x,$$

and use the same reference objects $g_x(\sigma)(y)$ for all the f_{τ} and their own derivatives.

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A **model** = an a priori definition of a number of functionals of the “noise”.

4 Solving semilinear singular SPDEs

- Reformulate the PDE as an integral equation

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Is $u = R^M(\mathbf{u}^M)$ the solution of an equation?

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Renormalized equation – An automated approach in a semilinear setting: From Bruned, Chandra, Chevyrev & Hairer (Renormalising SPDEs in regularity structures) and Bailleul & Bruned (Locality for singular stochastic PDEs), for a large class of models M^ϵ built from a regularized noise ξ^ϵ , there is an explicit function $C^\epsilon(\cdot)$ such that the function $u^\epsilon = \mathcal{R}^{M^\epsilon}(u^{M^\epsilon})$ is a solution to the renormalized equation

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If M^ϵ converges then u^{M^ϵ} and u^ϵ converge.

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► **Theorem (Local in time well-posedness)** – Take $u_0 \in C^{0+}(\mathbf{T})$ and a function $a \in C_b^3$ with values in a compact subset of $(0, \infty)$. In the full subcritical regime, one can construct a *regularity structure*, containing infinitely many trees $\tau^{\mathbf{p}}$ of any fixed degree, within which, for any admissible model, the quasilinear equation (1) is well-posed locally in time.

Here τ runs over the finitely many trees of the BHZ regularity structure of the semilinear (gKPZ) equation and \mathbf{p} is an integer-valued decoration on the edges of τ .

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– Denote by $\ell^{\lambda, \epsilon}(\tau)$ the BPHZ counterterms of the equation

$$\partial_t u - \lambda \partial_x^2 u = f(u)\xi^\epsilon + g(u)(\partial_x u)^2$$

built from the regularized noise ξ^ϵ .

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► **Theorem (Renormalized equation)** – There is an explicit *polynomial function* $\chi_a(\tau)$ of a and its derivatives such that the solution u^ϵ to

$$\partial_t u^\epsilon - a(u^\epsilon) \partial_x^2 u^\epsilon = f(u^\epsilon) \xi^\epsilon + g(u^\epsilon) (\partial_x u^\epsilon)^2 + \sum_{\tau} \frac{\ell^{a(u^\epsilon(\cdot)), \epsilon}(\tau)}{S(\tau)} \chi_a(\tau) (u^\epsilon) \mathcal{F}(\tau)(u^\epsilon),$$

with initial condition $u_0 \in C^{0+}(\mathbf{T})$, converges (in law) on a random time interval, as $\epsilon \downarrow 0$.

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Local in time well-posedness

- Modelled distributions built with *series instead of finite sums*: Needed in Picard iteration

$$u \simeq K^M \left(F(u)\zeta + G(u)(Du)^2 + (A(u) - 1)D^2u \right).$$

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- In a semilinear setting small factor for *contraction* comes from the gain in time explosion at initial time in multilevel *Schauder estimates* for K^M .
- Here the map $\Phi : u \mapsto K^M((A(u) - 1)D^2u)$ is *not contracting* but it has a decomposition

$$\Phi = \Phi_1 + \Phi_2$$

where Φ_2 is a contraction for a small enough time horizon and an *iterate of Φ_1 is a contraction*. (Details are being written.)

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The solution of the coherence relation has here a particular structure

$$\mathcal{F}^a(\tau^{\mathbf{p}})(\mathbf{c}_0, \mathbf{c}'_0) = \chi_a(\tau)(\mathbf{c}_0) (a(\mathbf{c}_0) - 1)^{|\mathbf{p}|} \mathcal{F}(\tau)(\mathbf{c}_0)$$

for some functions $\chi_a(\tau)$ and $\mathcal{F}(\tau)$ defined inductively, with $\chi_a(\tau)$ a polynomial function of a and its derivatives.

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for some functions $\chi_a(\tau)$ and $\mathcal{F}(\tau)$ defined inductively, with $\chi_a(\tau)$ a polynomial function of a and its derivatives.

► **Lemma** – For any τ with null \mathbf{p} -decoration the function

$$\lambda \mapsto \ell^{\lambda,\epsilon}(\tau)$$

is *analytic* in any given bounded interval $(a, b) \subset (0, +\infty)$ with $a > 0$ and

$$\frac{1}{n!} \partial_\lambda^n \ell^{\lambda,\epsilon}(\tau) = \sum_{\mathbf{p} \in \mathbb{N}^{E_\tau}, |\mathbf{p}|=n} \ell^{\lambda,\epsilon}(\tau^{\mathbf{p}}).$$

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Then one has

$$\sum_{\tau^{\mathbf{p}}} \frac{\ell^{1,\epsilon}(\tau^{\mathbf{p}})}{S(\tau^{\mathbf{p}})} \chi_a(\tau)(\mathbf{c}_0) (a(\mathbf{c}_0) - 1)^{|\mathbf{p}|} \mathcal{F}(\tau)(\mathbf{c}_0)$$

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ご清聴ありがとうございました

Thank you for your attention!