

SPDEs, criticality, and renormalisation

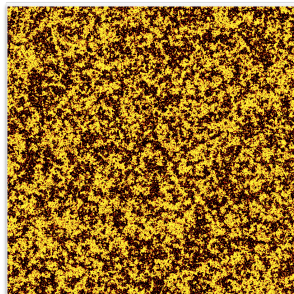
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An interesting model from Physics I

Ising model



Spin configurations:

$$\sigma(k) \in \{-1, 1\} \quad k \in \Lambda,$$

Energy:

$$H(\sigma) = -\frac{1}{2} \sum_{k \sim l} \sigma(k) \sigma(l),$$

Inverse temperature:

$$\beta,$$

Gibbs measure:

$$\mu_\beta(\sigma) \propto \exp(-\beta H(\sigma)).$$

An equation that should describe critical Ising I

Kac-model: (e.g. Presutti, Lebowitz, ... 90s): Spins interact with spins in a whole neighbourhood.

Interaction kernel: $\kappa_\gamma(k) = \gamma^n \kappa(\gamma k),$

Energy: $H_\gamma(\sigma) = -\frac{1}{2} \sum_{k,l} \kappa_\gamma(k-l) \sigma(k) \sigma(l),$

Coarse grained field: $h_\gamma(k) = \sum_l \kappa_\gamma(k-l) \sigma(l).$

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Evolution equation:

$$h_\gamma(t, k) = h_\gamma(0, k) + \int_0^t \mathcal{L}_\gamma h_\gamma(s, k) ds + m_\gamma(s, k), \quad (*)$$

where (for $\beta \approx 1$)

$$\mathcal{L}_\gamma h_\gamma(\sigma) \approx \left(\kappa_\gamma * h_\gamma(\sigma) - h_\gamma(\sigma) \right) - \frac{1}{3} \left(\kappa_\gamma * h_\gamma(\sigma, \cdot)^3 \right) + \dots$$

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Looks like an approximation to

$$h(t, x) \approx h(0, x) + \int_0^t \left(\Delta h(s, x) - h(s, x)^3 \right) ds + \int_0^t \xi(s, x) ds.$$

In one space dimension proved by [Bertini et al. 93].

SOS Surface growth model

Bertini-Giacomin '97 showed that a scaling limit of SOS-surface model is described by the **KPZ -equation**

$$\partial_t h = \partial_x^2 h + \lambda (\partial_x h)^2 + \xi.$$

- Introduced by Kardar-Parisi-Zhang ('86) to model fluctuation of a $1 + 1$ dimensional surface.
- Universal character: KPZ universality class.
- Recently, Spohn/Sasamoto and Amir/Corwin/Quastel '11 gave an explicit formula for one-point distribution.

Regularity of the stochastic heat equation

Stochastic heat equation:

$$\partial_t X = \Delta X - X + \xi \quad (\text{say on } [0, \infty] \times \mathbb{T}^n).$$

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$$\partial_t X_k(t) = -(|k|^2 + 1)X_k(t) + \dot{w}_k(t),$$

Ornstein-Uhlenbeck process. This implies

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In particular, **Sobolev norms**

$$\mathbb{E}\|X(t)\|_{H^s}^2 = \sum_{k \in \mathbb{Z}^n} |k|^{2s} \mathbb{E}|X_k(t)|^2 < \infty \Leftrightarrow s < \frac{2-n}{2}.$$

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In one dimension **Brownian regularity**. In $n \geq 2$ distribution valued.

KPZ:

$$\partial_t h = \partial_x^2 h + \lambda (\partial_x h)^2 + \xi.$$

- $h \in H^\alpha$ (actually \mathcal{C}^α) for $\alpha < \frac{1}{2} \Rightarrow$ cannot define $(\partial_x h)^2$.
- Bertini, Giacomini introduce **Hopf-Cole solution** to KPZ.

Write $h = \log(\lambda Z)$. Then formally

$$\partial_t Z = \partial_x^2 Z + Z\xi.$$

- Convergence result for SOS-model proved on level of Z .

Difficulty to define non-linearity II

$$\phi_n^4:$$

$$\partial_t \phi = \Delta \phi - \phi^3 + \phi + \xi.$$

- One spatial dimension $\phi \in \mathcal{C}^\alpha$, $\alpha < \frac{1}{2} \Rightarrow$ No problem.
- Two and three spatial dimensions ϕ not a function \Rightarrow classical methods do not work.
- Approximation for (Kac)-Ising only known in one spatial dimension.

Metatheorem (Hairer 13)

*Stable existence and uniqueness theory **locally in time** on **compact domains** for equations that are **locally subcritical**.*

Locally subcritical: *On small scales the non-linear term is lower order.*

Example ϕ_n^4 : scaling $x \mapsto \varepsilon x$, $t \mapsto \varepsilon^2 t$ and $\phi \mapsto \varepsilon^{\frac{n-2}{2}} \phi$, leaves stochastic heat equation invariant. Under this scaling ϕ_n^4 equation becomes

$$\partial_t \hat{\phi} = \Delta \hat{\phi} - \varepsilon^{4-n} \hat{\phi}^3 + \hat{\xi}$$

locally subcritical in for $n \leq 3$.

The need for renormalisation I

Approximation by regularised noise

$$d\phi_\delta = \left[\Delta\phi_\delta - (\phi_\delta^3 - \phi_\delta) \right] dt + dW_\delta.$$

- From now on $n = 2$, $\Omega = \mathbb{T}^2$ torus.
- $W_\delta(t, x) := \sum_{|k| \leq \frac{1}{\delta}} e^{ik \cdot x} w^k(t)$. Noise white in time, spatial correlations $\sim \delta$.

For $\delta > 0$ equation is well-posed. What happens if $\delta \rightarrow 0$?

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Related results: **Small noise**, pass to this limit on the level of large deviations (Kohn & Otto & Westdickenberg (Reznikoff) & Vanden-Eijnden – Cerrai & Freidlin – Barret & Bovier & Meleard).

Constructive field theory ('70s)

X_δ = stationary solution of (approximated) stochastic heat equation

$$dX_\delta = \Delta X_\delta - X_\delta + dW_\delta.$$

$$\mathbb{E}[X_\delta(x)X_\delta(y)] \lesssim |\log|x-y|| \wedge \log(\delta).$$

Question:

- Does X_δ^3 converge to a random distribution?
- Does $\langle X_\delta^3, \varphi \rangle$ converge to a random variable for smooth φ ?
- Does $\mathbb{E}[\langle X_\delta^3, \varphi \rangle^2]$ remain bounded as $\delta \rightarrow 0$?

Renormalised powers II

$$\mathbb{E}[\langle X_\delta^3, \varphi \rangle^2] = \int_{\mathbb{T}^2} \int_{\mathbb{T}^2} \varphi(x) \varphi(y) \mathbb{E}[X_\delta^3(x) X_\delta^3(y)] dx dy$$

Gaussian Moments

$$\begin{aligned} \mathbb{E}[X_\delta^3(x) X_\delta^3(y)] &= 6 \mathbb{E}[X_\delta(x) X_\delta(y)]^3 + 9 \mathbb{E}[X_\delta(x) X_\delta(y)] \mathbb{E}[X_\delta(x) X_\delta(x)]^2 \\ &\lesssim |\log(x - y)|^3 + |\log(\delta)|^2 |\log(x - y)| \end{aligned}$$

- $|\log(x - y)|$ term is integrable. $|\log(\delta)|$ term diverges.
- $\mathbb{E}[\langle X_\delta^3, \varphi \rangle^2]$ diverges as $\delta \rightarrow 0$.

Renormalised powers III

$$: X_\delta^3(x) := X_\delta^3(x) - 3C_\delta X_\delta(x) \quad \text{where } C_\delta = \mathbb{E}[X_\delta(x)^2] \sim |\log(\delta)|.$$

$$\Rightarrow \mathbb{E} \left[: X_\delta^3(x) : : X_\delta^3(y) : \right] = 6 \mathbb{E}[X_\delta(x) X_\delta(y)]^3.$$

$$\Rightarrow \mathbb{E}[\langle : X_\delta^3 : , \varphi \rangle^2] \text{ remains bounded.}$$

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Theorem (Glimm, Jaffe, Nelson, Gross,... 70s)

$: X_\delta^3 :$ converges to a random distribution $: X^3 :$ in every negative Sobolev space.

■ $: X^3 :$ called third **Wick power**.

$$d\phi_\delta = \left[\Delta\phi_\delta - (\phi_\delta^3 - \phi_\delta) \right] dt + dW_\delta$$

Theorem (Hairer, Ryser, W' 12)

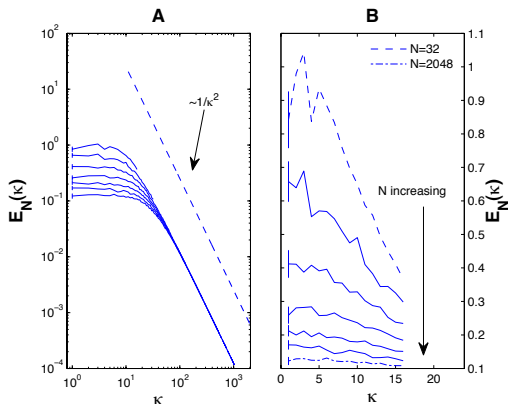
ϕ_δ^0 bounded in $C(\mathbb{T}^2)$. Then for $t_* > 0$, $s > 0$ almost surely

$$\|\phi_\delta\|_{C([t_*, T], H^{-s})} \rightarrow 0 \quad \text{for } \delta \rightarrow 0!$$

- Convergence is slow (logarithmic).

Numerics

Ryser & Nigam & Tupper *preprint '11*



N number of grid points $\sim \frac{1}{\delta^2}$.

$E_N(\kappa) \sim$ strength of κ Fourier mode.

Small noise

Small noise $\sigma(\delta) \leq 1$

$$d\bar{u}_\delta = \left[\Delta \bar{u}_\delta - (\bar{u}_\delta^3 - \bar{u}_\delta) \right] dt + \sigma(\delta) dW_\delta$$

Theorem

$\phi_\delta^0 \rightarrow u^0$ in $C(\mathbb{T}^2)$. Then for $t_* > 0$, $s > 0$ almost surely in H^{-s}

$$\|\bar{u}_\delta\|_{C([t_*, T], H^{-s})} \rightarrow \begin{cases} 0 & \text{if } |\log(\delta)|^{-\frac{1}{2}} \ll \sigma(\delta) \ll 1 \\ \bar{u}^* & \text{if } |\log(\delta)|^{-\frac{1}{2}} = \sigma(\delta) \\ \bar{u} & \text{if } 0 \ll \sigma(\delta) \ll |\log(\delta)|^{-\frac{1}{2}} \end{cases}.$$

- \bar{u} solution to deterministic Allen-Cahn equation.
- \bar{u}^* solution to $\dot{u} = \Delta \bar{u}_\delta - (\bar{u}_\delta^3 - \bar{u}_\delta) - C_* u$.

The scheme needs to be modified

Approximation by regularised noise

$$d\phi_\delta = \left[\Delta\phi_\delta - (\phi_\delta^3 - C_\delta\phi_\delta - \phi_\delta) \right] dt + dW_\delta.$$

$$C_\delta \sim |\log \delta|.$$

Theorem (da Prato/Debussche '03)

ϕ_δ converge to a limit. This limit is called solution to

$$du = \left[\Delta u - (: u^3 : - u) \right] dt + dW,$$

or dynamic ϕ_2^4 model.

Strategy I:

1.) Lift Gaussian process

Lemma

For every $t \in [0, T], p \geq 1, s > 0$

- $X_\delta \rightarrow X,$
- $X_\delta^2 - C_\delta \rightarrow: X^2 :,$
- $X_\delta^3 - C_\delta X_\delta \rightarrow: X^3 :,$

in $L^p(\mathcal{B}_{\infty,\infty}^{-s})$.

- Gaussian moment calculation.
- **equivalence of moments** in fixed Wiener chaos - Nelson estimate.
- Regularity measured in Besov spaces.

2.) Non-linear evolution as **continuous** function of lifted Gaussian process

Standard regularisation trick: $v_\delta = \phi_\delta - X_\delta$.

$$\begin{aligned}\frac{dv_\delta}{dt} &= \Delta v_\delta - ((X_\delta + v_\delta)^3 - 3 C_\delta(X_\delta + v_\delta)) \\ &= \Delta v_\delta - (: X_\delta^3 : + 3 : X_\delta^2 : v_\delta + 3 X_\delta v_\delta^2 + v_\delta^3).\end{aligned}$$

Multiplicative inequality: If $s < 0 < \alpha$ and $s + \alpha > 0$. Then

$$\|u v\|_{B_{\infty,\infty}^s} \lesssim \|u\|_{B_{\infty,\infty}^s} \|v\|_{B_{\infty,\infty}^\alpha}.$$

Used to deal with nonlinearity.

Comments:

- Only works on bounded domains.
- Extra argument needed for **global in time** solutions. Extra argument required (invariant measures à la Bourgain).
- In 3-d the normalisation is more tricky. Careful with word "Wick".
- More term in expansion necessary, one extra term diverges. Approximations of type

$$d\phi_\delta = \left[\Delta\phi_\delta - (\phi_\delta^3 - C_\delta\phi_\delta - \phi_\delta) \right] dt + dW_\delta,$$

for $C_\delta = \frac{C_1}{\delta} + C_2|\log(\delta)|$.

Summary/Outlook:

- Non-linear white noise driven SPDEs arise as scaling limits for particle systems in interesting regimes.
- Solutions to these equations have very poor regularity properties and it is not always clear how to treat non-linear terms.
- Infinite constants have to be dealt with.

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From friday: How to implement renormalisation for more complicated equations.