# **Regularity Structures**

Martin Hairer

University of Warwick

Seoul, August 14, 2014

#### What are regularity structures?

Algebraic structures providing "skeleton" for analytical "models" mimicking properties of Taylor polynomials: (T, G, A). Model:  $T \times \mathbf{R}^d \to \mathcal{D}'$ .

Polynomial model:  $(P, x_0) \mapsto P(\cdot - x_0)$ .

Algebraic properties: Group G acting by reexpansions on  $P \in T$ :

$$P(x - x_0) = P((x - x_1) + x_1 - x_0) = (\Gamma_{x_0, x_1} P)(x - x_1) .$$

For every  $\Gamma \in G$ ,  $\deg(\Gamma P - P) < \deg P$  and  $\Gamma PQ = (\Gamma P)(\Gamma Q)$ .

Analytical properties: Homogeneous monomials vanish at base point with order (speed) equal to their degree.

#### What are regularity structures?

Algebraic structures providing "skeleton" for analytical "models" mimicking properties of Taylor polynomials: (T, G, A). Model:  $T \times \mathbf{R}^d \to \mathcal{D}'$ .

Polynomial model:  $(P, x_0) \mapsto P(\cdot - x_0)$ .

Algebraic properties: Group G acting by reexpansions on  $P \in T$ :

$$P(x - x_0) = P((x - x_1) + x_1 - x_0) = (\Gamma_{x_0, x_1} P)(x - x_1) .$$

For every  $\Gamma \in \mathbf{G}$ ,  $\deg(\Gamma P - P) < \deg P$  and  $\Gamma PQ = (\Gamma P)(\Gamma Q)$ .

Analytical properties: Homogeneous monomials vanish at base point with order (speed) equal to their degree.

#### Another example

*T*: linear span of 1 (degree 0) and W (degree  $\frac{1}{2}$ ).

Model: For some fixed Hölder- $\frac{1}{2}$  function W, set

$$(a\mathbf{1}+b\mathbf{W},x_0)\mapsto a+b\big(W(\cdot)-W(x_0)\big)$$
.

Group G: 
$$\Gamma_{x_0,x_1} \mathbf{W} = \mathbf{W} + (W(x_0) - W(x_1))\mathbf{1}.$$
  
 $\Gamma_{x_0,x_1} \mathbf{1} = \mathbf{1}$ 

Important property: For a given regularity structure, one can have many different models. (Here: given by choice of W.)

#### Another example

*T*: linear span of 1 (degree 0) and W (degree  $\frac{1}{2}$ ).

Model: For some fixed Hölder- $\frac{1}{2}$  function W, set

$$(a\mathbf{1}+b\mathbf{W},x_0)\mapsto a+b\big(W(\cdot)-W(x_0)\big)$$
.

Group G: 
$$\Gamma_{x_0,x_1} \mathbf{W} = \mathbf{W} + (W(x_0) - W(x_1))\mathbf{1}.$$
  
 $\Gamma_{x_0,x_1} \mathbf{1} = \mathbf{1}$ 

Important property: For a given regularity structure, one can have many different models. (Here: given by choice of W.)

#### What are they good for?

Construct robust solution theories for very singular SPDEs. Examples:

$$\begin{aligned} \partial_t h &= \partial_x^2 h + (\partial_x h)^2 + \xi , & (d = 1) \\ \partial_t \Phi &= \Delta \Phi - \Phi^3 + \xi , & (d = 2, 3) \\ \partial_t u &= \Delta u + g_{ij}(u) \partial_i u \, \partial_j u + \sigma(u) \eta , & (d = 2, 3) \\ \partial_t v &= \partial_x^2 v + f(v) + \sigma(v) \xi . & (d = 1) \end{aligned}$$

#### Here $\xi$ is space-time white noise and $\eta$ is spatial white noise.

KPZ (h): universal model for interface propagation. Dynamical  $\Phi_3^4$ : universal model for dynamics of near mean-field phase transition models near critical temperature. PAM  $(u \text{ with } g = 0 \text{ and } \sigma(u) = u)$ : universal model for weakly killed diffusions.

#### What are they good for?

Construct robust solution theories for very singular SPDEs. Examples:

$$\begin{aligned} \partial_t h &= \partial_x^2 h + (\partial_x h)^2 + \xi , & (d = 1) \\ \partial_t \Phi &= \Delta \Phi - \Phi^3 + \xi , & (d = 2, 3) \\ \partial_t u &= \Delta u + g_{ij}(u) \partial_i u \, \partial_j u + \sigma(u) \eta , & (d = 2, 3) \\ \partial_t v &= \partial_x^2 v + f(v) + \sigma(v) \xi . & (d = 1) \end{aligned}$$

Here  $\xi$  is space-time white noise and  $\eta$  is spatial white noise.

KPZ (*h*): universal model for interface propagation. Dynamical  $\Phi_3^4$ : universal model for dynamics of near mean-field phase transition models near critical temperature. PAM (*u* with g = 0 and  $\sigma(u) = u$ ): universal model for weakly killed diffusions.

#### What are they good for?



Try to define distribution " $\eta(x) = \frac{1}{|x|} - C\delta(x)$ ".

**Problem:** Integral of 1/|x| diverges, so we need to set " $C = \infty$ " to compensate!

Formal definition:

$$\eta_{\chi}(\phi) = \int_{\mathbf{R}} \frac{\phi(x) - \chi(x)\phi(0)}{|x|} dx ,$$

for some smooth compactly supported cut-off  $\chi$  with  $\chi(0) = 1$ . Yields one-parameter family  $c \mapsto \eta_c$  of models, but no canonical "choice of origin" for c.

Try to define distribution " $\eta(x) = \frac{1}{|x|} - C\delta(x)$ ".

**Problem:** Integral of 1/|x| diverges, so we need to set " $C = \infty$ " to compensate!

Formal definition:

$$\eta_{\chi}(\phi) = \int_{\mathbf{R}} \frac{\phi(x) - \chi(x)\phi(0)}{|x|} dx ,$$

for some smooth compactly supported cut-off  $\chi$  with  $\chi(0) = 1$ . Yields one-parameter family  $c \mapsto \eta_c$  of models, but no canonical "choice of origin" for c.

Try to define distribution " $\eta(x) = \frac{1}{|x|} - C\delta(x)$ ".

**Problem:** Integral of 1/|x| diverges, so we need to set " $C = \infty$ " to compensate!

Formal definition:

$$\eta_{\chi}(\phi) = \int_{\mathbf{R}} \frac{\phi(x) - \chi(x)\phi(0)}{|x|} \, dx \; ,$$

for some smooth compactly supported cut-off  $\chi$  with  $\chi(0) = 1$ . Yields one-parameter family  $c \mapsto \eta_c$  of models, but no canonical "choice of origin" for c.

Try to define distribution " $\eta(x) = \frac{1}{|x|} - C\delta(x)$ ".

**Problem:** Integral of 1/|x| diverges, so we need to set " $C = \infty$ " to compensate!

Formal definition:

$$\eta_{\chi}(\phi) = \int_{\mathbf{R}} \frac{\phi(x) - \chi(x)\phi(0)}{|x|} \, dx \; ,$$

for some smooth compactly supported cut-off  $\chi$  with  $\chi(0) = 1$ . Yields one-parameter family  $c \mapsto \eta_c$  of models, but no canonical "choice of origin" for c.

- 1. If nonlinear term is  $\sigma(u) \xi$ , Itô calculus can be used. Relies crucially on martingale property, broken by regularisation.
- 2. KPZ and 1D stochastic Burgers can be treated using controlled rough paths by Lyons / Gubinelli (H. '11 / H. '13).
- 3. Solve  $\partial_t Z = \partial_x^2 Z + Z \xi$  (SHE) and interpret  $h = \log Z$  as KPZ (Hopf '50 / Cole '51 / Bertini-Giacomin '97).
- 4. Dynamical  $\Phi_2^4$  model: write  $\Phi = \Psi + \tilde{\Phi}$  with  $\Psi$  solution to linear equation and derive well-posed equation for  $\tilde{\Phi}$  (Albeverio-Röckner '91 / Da Prato-Debussche '03).
- 5. Alternative theory using paraproducts can in principle treat KPZ and  $\Phi_3^4$  (Gubinelli-Imkeller-Perkowski '14).

- 1. If nonlinear term is  $\sigma(u) \xi$ , Itô calculus can be used. Relies crucially on martingale property, broken by regularisation.
- 2. KPZ and 1D stochastic Burgers can be treated using controlled rough paths by Lyons / Gubinelli (H. '11 / H. '13).
- 3. Solve  $\partial_t Z = \partial_x^2 Z + Z \xi$  (SHE) and interpret  $h = \log Z$  as KPZ (Hopf '50 / Cole '51 / Bertini-Giacomin '97).
- 4. Dynamical  $\Phi_2^4$  model: write  $\Phi = \Psi + \tilde{\Phi}$  with  $\Psi$  solution to linear equation and derive well-posed equation for  $\tilde{\Phi}$  (Albeverio-Röckner '91 / Da Prato-Debussche '03).
- 5. Alternative theory using paraproducts can in principle treat KPZ and  $\Phi_3^4$  (Gubinelli-Imkeller-Perkowski '14).

- 1. If nonlinear term is  $\sigma(u) \xi$ , Itô calculus can be used. Relies crucially on martingale property, broken by regularisation.
- 2. KPZ and 1D stochastic Burgers can be treated using controlled rough paths by Lyons / Gubinelli (H. '11 / H. '13).
- 3. Solve  $\partial_t Z = \partial_x^2 Z + Z \xi$  (SHE) and interpret  $h = \log Z$  as KPZ (Hopf '50 / Cole '51 / Bertini-Giacomin '97).
- 4. Dynamical  $\Phi_2^4$  model: write  $\Phi = \Psi + \tilde{\Phi}$  with  $\Psi$  solution to linear equation and derive well-posed equation for  $\tilde{\Phi}$  (Albeverio-Röckner '91 / Da Prato-Debussche '03).
- 5. Alternative theory using paraproducts can in principle treat KPZ and  $\Phi_3^4$  (Gubinelli-Imkeller-Perkowski '14).

- 1. If nonlinear term is  $\sigma(u) \xi$ , Itô calculus can be used. Relies crucially on martingale property, broken by regularisation.
- 2. KPZ and 1D stochastic Burgers can be treated using controlled rough paths by Lyons / Gubinelli (H. '11 / H. '13).
- 3. Solve  $\partial_t Z = \partial_x^2 Z + Z \xi$  (SHE) and interpret  $h = \log Z$  as KPZ (Hopf '50 / Cole '51 / Bertini-Giacomin '97).
- 4. Dynamical  $\Phi_2^4$  model: write  $\Phi = \Psi + \tilde{\Phi}$  with  $\Psi$  solution to linear equation and derive well-posed equation for  $\tilde{\Phi}$  (Albeverio-Röckner '91 / Da Prato-Debussche '03).
- 5. Alternative theory using paraproducts can in principle treat KPZ and  $\Phi_3^4$  (Gubinelli-Imkeller-Perkowski '14).

- 1. If nonlinear term is  $\sigma(u) \xi$ , Itô calculus can be used. Relies crucially on martingale property, broken by regularisation.
- 2. KPZ and 1D stochastic Burgers can be treated using controlled rough paths by Lyons / Gubinelli (H. '11 / H. '13).
- 3. Solve  $\partial_t Z = \partial_x^2 Z + Z \xi$  (SHE) and interpret  $h = \log Z$  as KPZ (Hopf '50 / Cole '51 / Bertini-Giacomin '97).
- 4. Dynamical  $\Phi_2^4$  model: write  $\Phi = \Psi + \tilde{\Phi}$  with  $\Psi$  solution to linear equation and derive well-posed equation for  $\tilde{\Phi}$  (Albeverio-Röckner '91 / Da Prato-Debussche '03).
- 5. Alternative theory using paraproducts can in principle treat KPZ and  $\Phi_3^4$  (Gubinelli-Imkeller-Perkowski '14).

#### Central limit theorem: Gaussian universality

KPZ strong Universality conjecture: At large scales, the space-time fluctuations of a large class of 1 + 1-dimensional interface propagation model are described by a universal Markov process H, self-similar with exponents 1 - 2 - 3:

$$\lambda^{-1}H(\lambda^2 x, \lambda^3 t) \stackrel{\text{law}}{=} H(x, t) \ .$$

**Exactly solvable models:** Borodin, Corwin, Quastel, Sasamoto, Spohn, etc. Yields partial characterisation of limiting "KPZ fixed point" (*H*): agrees with experimental evidence (Takeuchi & AI).

Central limit theorem: Gaussian universality

KPZ strong Universality conjecture: At large scales, the space-time fluctuations of a large class of 1 + 1-dimensional interface propagation model are described by a universal Markov process H, self-similar with exponents 1 - 2 - 3:

$$\lambda^{-1}H(\lambda^2 x, \lambda^3 t) \stackrel{\text{\tiny law}}{=} H(x, t) \; .$$

**Exactly solvable models:** Borodin, Corwin, Quastel, Sasamoto, Spohn, etc. Yields partial characterisation of limiting "KPZ fixed point" (*H*): agrees with experimental evidence (Takeuchi & AI).

Central limit theorem: Gaussian universality

KPZ strong Universality conjecture: At large scales, the space-time fluctuations of a large class of 1 + 1-dimensional interface propagation model are described by a universal Markov process H, self-similar with exponents 1 - 2 - 3:

$$\lambda^{-1}H(\lambda^2 x, \lambda^3 t) \stackrel{\text{\tiny law}}{=} H(x, t) \ .$$

**Exactly solvable models:** Borodin, Corwin, Quastel, Sasamoto, Spohn, etc. Yields partial characterisation of limiting "KPZ fixed point" (H): agrees with experimental evidence (Takeuchi & AI).

#### Central limit theorem: Gaussian universality



Spohn, etc. Yields partial characterisation of limiting "KPZ fixed point" (H): agrees with experimental evidence (Takeuchi & AI).

### Heuristic picture

Universality for symmetric interface fluctuation models: scaling exponents 1-2-4, Gaussian limit. Heuristic picture of the evolution of interface models under "zooming out":



KPZ equation: red line.

### Heuristic picture

Universality for symmetric interface fluctuation models: scaling exponents 1-2-4, Gaussian limit. Heuristic picture of the evolution of interface models under "zooming out":



#### Conjecture: the KPZ equation is the only model on the "red line".

**Conjecture:** Let  $h_{\varepsilon}$  be any "natural" one-parameter family of asymmetric interface models with  $\varepsilon$  denoting the strength of the asymmetry such that propagation speed  $\approx \sqrt{\varepsilon}$ .

As  $\varepsilon \to 0$ , there is a choice of  $C_{\varepsilon} \sim \varepsilon^{-1}$  such that  $\sqrt{\varepsilon}h_{\varepsilon}(\varepsilon^{-1}x,\varepsilon^{-2}t) - C_{\varepsilon}t$  converges to solutions h to the KPZ equation.

Height function of WASEP (Bertini-Giacomin '97). Accumulation points satisfy weak version of KPZ for certain generalisations of WASEP (Jara-Gonçalves '10).

Conjecture: the KPZ equation is the only model on the "red line".

**Conjecture:** Let  $h_{\varepsilon}$  be any "natural" one-parameter family of asymmetric interface models with  $\varepsilon$  denoting the strength of the asymmetry such that propagation speed  $\approx \sqrt{\varepsilon}$ .

As  $\varepsilon \to 0$ , there is a choice of  $C_{\varepsilon} \sim \varepsilon^{-1}$  such that  $\sqrt{\varepsilon}h_{\varepsilon}(\varepsilon^{-1}x,\varepsilon^{-2}t) - C_{\varepsilon}t$  converges to solutions h to the KPZ equation.

Height function of WASEP (Bertini-Giacomin '97). Accumulation points satisfy weak version of KPZ for certain generalisations of WASEP (Jara-Gonçalves '10).

Conjecture: the KPZ equation is the only model on the "red line".

**Conjecture:** Let  $h_{\varepsilon}$  be any "natural" one-parameter family of asymmetric interface models with  $\varepsilon$  denoting the strength of the asymmetry such that propagation speed  $\approx \sqrt{\varepsilon}$ .

As  $\varepsilon \to 0$ , there is a choice of  $C_{\varepsilon} \sim \varepsilon^{-1}$  such that  $\sqrt{\varepsilon}h_{\varepsilon}(\varepsilon^{-1}x,\varepsilon^{-2}t) - C_{\varepsilon}t$  converges to solutions h to the KPZ equation.

#### Height function of WASEP (Bertini-Giacomin '97).

Accumulation points satisfy weak version of KPZ for certain generalisations of WASEP (Jara-Gonçalves '10).

Conjecture: the KPZ equation is the only model on the "red line".

**Conjecture:** Let  $h_{\varepsilon}$  be any "natural" one-parameter family of asymmetric interface models with  $\varepsilon$  denoting the strength of the asymmetry such that propagation speed  $\approx \sqrt{\varepsilon}$ .

As  $\varepsilon \to 0$ , there is a choice of  $C_{\varepsilon} \sim \varepsilon^{-1}$  such that  $\sqrt{\varepsilon}h_{\varepsilon}(\varepsilon^{-1}x,\varepsilon^{-2}t) - C_{\varepsilon}t$  converges to solutions h to the KPZ equation.

Height function of WASEP (Bertini-Giacomin '97). Accumulation points satisfy weak version of KPZ for certain generalisations of WASEP (Jara-Gonçalves '10).

Class of models:

$$\partial_t h_{\varepsilon} = \partial_x^2 h_{\varepsilon} + \sqrt{\varepsilon} P(\partial_x h_{\varepsilon}) + F$$
,

with P an even polynomial, F a Gaussian field with compactly supported correlations  $\rho(t,x)$  s.t.  $\int \rho = 1.$ 

**Theorem (H., Quastel '14, in progress)** As  $\varepsilon \to 0$ , there is a choice of  $C_{\varepsilon} \sim \varepsilon^{-1}$  such that  $\sqrt{\varepsilon}h(\varepsilon^{-1}x,\varepsilon^{-2}t) - C_{\varepsilon}t$  converges to solutions to (KPZ)<sub> $\lambda$ </sub> with  $\lambda$  depending in a non-trivial way on all coefficients of P.

Class of models:

$$\partial_t h_{\varepsilon} = \partial_x^2 h_{\varepsilon} + \sqrt{\varepsilon} P(\partial_x h_{\varepsilon}) + F$$
,

with P an even polynomial, F a Gaussian field with compactly supported correlations  $\rho(t, x)$  s.t.  $\int \rho = 1$ .

**Theorem (H., Quastel '14, in progress)** As  $\varepsilon \to 0$ , there is a choice of  $C_{\varepsilon} \sim \varepsilon^{-1}$  such that  $\sqrt{\varepsilon}h(\varepsilon^{-1}x,\varepsilon^{-2}t) - C_{\varepsilon}t$  converges to solutions to  $(\text{KPZ})_{\lambda}$  with  $\lambda$  depending in a non-trivial way on all coefficients of P.

Class of models:

$$\partial_t h_{\varepsilon} = \partial_x^2 h_{\varepsilon} + \sqrt{\varepsilon} P(\partial_x h_{\varepsilon}) + F$$
,

with P an even polynomial, F a Gaussian field with compactly supported correlation Nonlinearity  $\lambda(\partial_x h)^2$ 

**Theorem (H., Quastel '14, in progress)** As  $\varepsilon \to 0$ , there is a choice of  $C_{\varepsilon} \sim \varepsilon^{-1}$  such that  $\sqrt{\varepsilon}h(\varepsilon^{-1}x,\varepsilon^{-2}t) - C_{\varepsilon}t$  converges to solutions to  $(\text{KPZ})_{\lambda}$  with  $\lambda$  depending in a non-trivial way on all coefficients of P.

Class of models:

$$\partial_t h_{\varepsilon} = \partial_x^2 h_{\varepsilon} + \sqrt{\varepsilon} P(\partial_x h_{\varepsilon}) + F$$
,

with P an even polynomial, F a Gaussian field with compactly supported correlations  $\rho(t, x)$  s.t.  $\int \rho = 1$ .

**Theorem (H., Quastel '14, in progress)** As  $\varepsilon \to 0$ , there is a choice of  $C_{\varepsilon} \sim \varepsilon^{-1}$  such that  $\sqrt{\varepsilon}h(\varepsilon^{-1}x,\varepsilon^{-2}t) - C_{\varepsilon}t$  converges to solutions to  $(\text{KPZ})_{\lambda}$  with  $\lambda$  depending in a non-trivial way on all coefficients of P.

Class of models:

$$\partial_t h_{\varepsilon} = \partial_x^2 h_{\varepsilon} + \sqrt{\varepsilon} P(\partial_x h_{\varepsilon}) + F$$
,

with P an even polynomial, F a Gaussian field with compactly supported correlations  $\rho(t, x)$  s.t.  $\int \rho = 1$ .

**Theorem (H., Quastel '14, in progress)** As  $\varepsilon \to 0$ , there is a choice of  $C_{\varepsilon} \sim \varepsilon^{-1}$  such that  $\sqrt{\varepsilon}h(\varepsilon^{-1}x,\varepsilon^{-2}t) - C_{\varepsilon}t$  converges to solutions to  $(\text{KPZ})_{\lambda}$  with  $\lambda$  depending in a non-trivial way on all coefficients of P.

# Case $P(u) = u^4$

Write 
$$\tilde{h}_{\varepsilon}(x,t) = \sqrt{\varepsilon}h(\varepsilon^{-1}x,\varepsilon^{-2}t) - C_{\varepsilon}t$$
. Satisfies  
 $\partial_t \tilde{h}_{\varepsilon} = \partial_x^2 h_{\varepsilon} + \varepsilon (\partial_x \tilde{h}_{\varepsilon})^4 + \xi_{\varepsilon} - C_{\varepsilon}$ ,

#### with $\xi_{\varepsilon}$ an $\varepsilon\text{-approximation}$ to white noise.

Fact: Derivatives of microscopic model do not converge to 0 as  $\varepsilon \rightarrow 0$ : no small gradients! Heuristic: gradients have  $\mathcal{O}(1)$  fluctuations but are small on average over large scales... General formula:

$$\lambda = \frac{1}{2} \int P''(u) \,\mu(du) \,, \qquad C_{\varepsilon} = \frac{1}{\varepsilon} \int P(u) \,\mu(du) + \mathcal{O}(1) \,,$$

with  $\mu$  a Gaussian measure, explicitly computable variance.

# Case $P(u) = u^4$

Write 
$$\tilde{h}_{\varepsilon}(x,t) = \sqrt{\varepsilon}h(\varepsilon^{-1}x,\varepsilon^{-2}t) - C_{\varepsilon}t$$
. Satisfies  
 $\partial_t \tilde{h}_{\varepsilon} = \partial_x^2 h_{\varepsilon} + \varepsilon (\partial_x \tilde{h}_{\varepsilon})^4 + \xi_{\varepsilon} - C_{\varepsilon}$ .

with  $\xi_{\varepsilon}$  an  $\varepsilon\text{-approximation}$  to white noise.

Fact: Derivatives of microscopic model do not converge to 0 as  $\varepsilon \rightarrow 0$ : no small gradients! Heuristic: gradients have  $\mathcal{O}(1)$  fluctuations but are small on average over large scales... General formula:

$$\lambda = \frac{1}{2} \int P''(u) \,\mu(du) \,, \qquad C_{\varepsilon} = \frac{1}{\varepsilon} \int P(u) \,\mu(du) + \mathcal{O}(1) \,,$$

with  $\mu$  a Gaussian measure, explicitly computable variance.

# Case $P(u) = u^4$

Write 
$$\tilde{h}_{\varepsilon}(x,t) = \sqrt{\varepsilon}h(\varepsilon^{-1}x,\varepsilon^{-2}t) - C_{\varepsilon}t$$
. Satisfies  
 $\partial_t \tilde{h}_{\varepsilon} = \partial_x^2 h_{\varepsilon} + \varepsilon (\partial_x \tilde{h}_{\varepsilon})^4 + \xi_{\varepsilon} - C_{\varepsilon}$ .

with  $\xi_{\varepsilon}$  an  $\varepsilon$ -approximation to white noise.

Fact: Derivatives of microscopic model do not converge to 0 as  $\varepsilon \to 0$ : no small gradients! Heuristic: gradients have  $\mathcal{O}(1)$  fluctuations but are small on average over large scales... General formula:

$$\lambda = \frac{1}{2} \int P''(u) \,\mu(du) \,, \qquad C_{\varepsilon} = \frac{1}{\varepsilon} \int P(u) \,\mu(du) + \mathcal{O}(1) \,,$$

with  $\mu$  a Gaussian measure, explicitly computable variance.

Rewrite general equation in integral form as

$$H = \mathcal{P}\big(\mathcal{E}(\mathscr{D}H)^4 + a(\mathscr{D}H)^2 + \Xi\big) ,$$

with  ${\cal E}$  an abstract "integration operator" of order 1,  ${\cal P}$  convolution with heat kernel.

Find two-parameter lift of noise  $\xi_{\varepsilon} \mapsto \Psi_{\alpha,c}(\xi_{\varepsilon})$  so that  $h = \mathcal{R}H$  solves

$$\partial_t h = \partial_x^2 h + \alpha H_4(\partial_x h, c) + a H_2(\partial_x h, c) + \xi_{\varepsilon}$$
  
=  $\partial_x^2 h + \alpha (\partial_x h)^4 + (a - 6\alpha c)(\partial_x h)^2 + (3\alpha c^2 - ac) + \xi_{\varepsilon}.$ 

Rewrite general equation in integral form as

$$H = \mathcal{P}\big(\mathcal{E}(\mathscr{D}H)^4 + a(\mathscr{D}H)^2 + \Xi\big) ,$$

with  ${\cal E}$  an abstract "integration operator" of order 1,  ${\cal P}$  convolution with heat kernel.

Find two-parameter lift of noise  $\xi_{\varepsilon} \mapsto \Psi_{\alpha,c}(\xi_{\varepsilon})$  so that  $h = \mathcal{R}H$  solves

$$\partial_t h = \partial_x^2 h + \alpha H_4(\partial_x h, c) + a H_2(\partial_x h, c) + \xi_{\varepsilon}$$
  
=  $\partial_x^2 h + \alpha (\partial_x h)^4 + (a - 6\alpha c)(\partial_x h)^2 + (3\alpha c^2 - ac) + \xi_{\varepsilon}.$ 

Rewrite general equation in integral form as



Rewrite general equation in integral form as

$$H = \mathcal{P}\big(\mathcal{E}(\mathscr{D}H)^4 + a(\mathscr{D}H)^2 + \Xi\big) ,$$

with  ${\cal E}$  an abstract "integration operator" of order 1,  ${\cal P}$  convolution with heat kernel.

Find two-parameter lift of noise  $\xi_{\varepsilon} \mapsto \Psi_{\alpha,c}(\xi_{\varepsilon})$  so that  $h = \mathcal{R}H$  solves

$$\partial_t h = \partial_x^2 h + \alpha H_4(\partial_x h, c) + a H_2(\partial_x h, c) + \xi_{\varepsilon} = \partial_x^2 h + \alpha (\partial_x h)^4 + (a - 6\alpha c)(\partial_x h)^2 + (3\alpha c^2 - ac) + \xi_{\varepsilon} .$$

Rewrite general equation in integral form as

$$H = \mathcal{P}\big(\mathcal{E}(\mathscr{D}H)^4 + a(\mathscr{D}H)^2 + \Xi\big) ,$$

with  ${\cal E}$  an abstract "integration operator" of order 1,  ${\cal P}$  convolution with heat kernel.

Find two-parameter lift of noise  $\xi_{\varepsilon} \mapsto \Psi_{\alpha,c}(\xi_{\varepsilon})$  so that  $h = \mathcal{R}H$  solves

$$\partial_t h = \partial_x^2 h + \alpha H_4(\partial_x h, c) + a H_2(\partial_x h, c) + \xi_{\varepsilon} = \partial_x^2 h + \alpha (\partial_x h)^4 + (a - 6\alpha c)(\partial_x h)^2 + (3\alpha c^2 - ac) + \xi_{\varepsilon} .$$

- 1. Strong Universality without exact solvability???
- 2. Hyperbolic / dispersive problems??
- Obtain convergence results for discrete models (H.-Maas-Weber '12; Mourrat-Weber, in progress).
- 4. Non-Gaussian noise / fully nonlinear continuum models.
- 5. Control over larger scales  $\Rightarrow$  KPZ fixed point.
- 6. Characterisation of possible renormalisation maps. When does it yield a modified equation in closed form?
- Systematic way of choosing renormalisation procedure and proving convergence (H.-Quastel; Bruned-H.-Zambotti, in progress).

- 1. Strong Universality without exact solvability???
- 2. Hyperbolic / dispersive problems??
- 3. Obtain convergence results for discrete models (H.-Maas-Weber '12; Mourrat-Weber, in progress).
- 4. Non-Gaussian noise / fully nonlinear continuum models.
- 5. Control over larger scales  $\Rightarrow$  KPZ fixed point.
- 6. Characterisation of possible renormalisation maps. When does it yield a modified equation in closed form?
- Systematic way of choosing renormalisation procedure and proving convergence (H.-Quastel; Bruned-H.-Zambotti, in progress).

- 1. Strong Universality without exact solvability???
- 2. Hyperbolic / dispersive problems??
- 3. Obtain convergence results for discrete models (H.-Maas-Weber '12; Mourrat-Weber, in progress).
- 4. Non-Gaussian noise / fully nonlinear continuum models.
- 5. Control over larger scales  $\Rightarrow$  KPZ fixed point.
- 6. Characterisation of possible renormalisation maps. When does it yield a modified equation in closed form?
- Systematic way of choosing renormalisation procedure and proving convergence (H.-Quastel; Bruned-H.-Zambotti, in progress).

- 1. Strong Universality without exact solvability???
- 2. Hyperbolic / dispersive problems??
- 3. Obtain convergence results for discrete models (H.-Maas-Weber '12; Mourrat-Weber, in progress).
- 4. Non-Gaussian noise / fully nonlinear continuum models.
- 5. Control over larger scales  $\Rightarrow$  KPZ fixed point.
- 6. Characterisation of possible renormalisation maps. When does it yield a modified equation in closed form?
- Systematic way of choosing renormalisation procedure and proving convergence (H.-Quastel; Bruned-H.-Zambotti, in progress).

- 1. Strong Universality without exact solvability???
- 2. Hyperbolic / dispersive problems??
- 3. Obtain convergence results for discrete models (H.-Maas-Weber '12; Mourrat-Weber, in progress).
- 4. Non-Gaussian noise / fully nonlinear continuum models.
- 5. Control over larger scales  $\Rightarrow$  KPZ fixed point.
- 6. Characterisation of possible renormalisation maps. When does it yield a modified equation in closed form?
- Systematic way of choosing renormalisation procedure and proving convergence (H.-Quastel; Bruned-H.-Zambotti, in progress).

- 1. Strong Universality without exact solvability???
- 2. Hyperbolic / dispersive problems??
- 3. Obtain convergence results for discrete models (H.-Maas-Weber '12; Mourrat-Weber, in progress).
- 4. Non-Gaussian noise / fully nonlinear continuum models.
- 5. Control over larger scales  $\Rightarrow$  KPZ fixed point.
- 6. Characterisation of possible renormalisation maps. When does it yield a modified equation in closed form?
- Systematic way of choosing renormalisation procedure and proving convergence (H.-Quastel; Bruned-H.-Zambotti, in progress).

- 1. Strong Universality without exact solvability???
- 2. Hyperbolic / dispersive problems??
- 3. Obtain convergence results for discrete models (H.-Maas-Weber '12; Mourrat-Weber, in progress).
- 4. Non-Gaussian noise / fully nonlinear continuum models.
- 5. Control over larger scales  $\Rightarrow$  KPZ fixed point.
- 6. Characterisation of possible renormalisation maps. When does it yield a modified equation in closed form?
- 7. Systematic way of choosing renormalisation procedure and proving convergence (H.-Quastel; Bruned-H.-Zambotti, in progress).