

Taming infinities

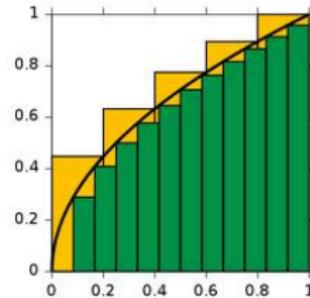
M. Hairer

University of Warwick

29.05.2014

Infinities and infinitesimals

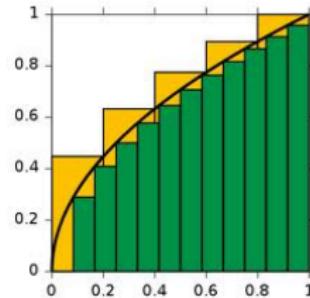
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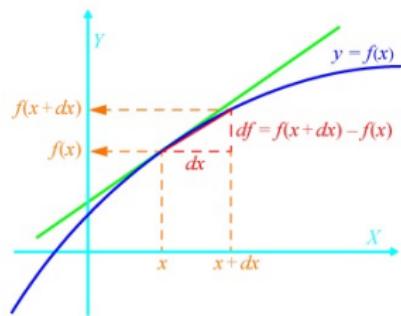
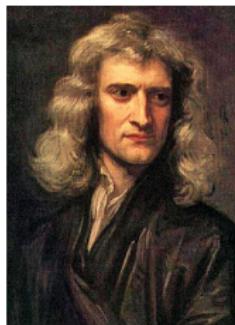
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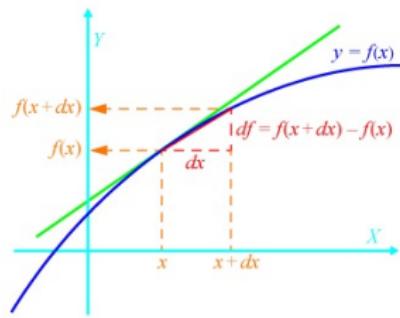
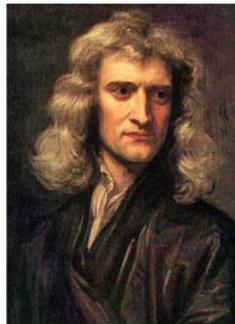
Leibniz and Newton (1680's): Compute derivatives by dividing infinitesimals.



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Resolution

Cauchy, Bolzano, Weierstrass, etc (1820's onward): Rigorous formulation of limits without need for infinitesimals.

Dramatic consequences: Development of a consistent calculus allowing to model many physical processes and underpinning much of modern technology.

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Quantum field theory

General methodology: “Guess” form of Lagrangian, predict outcomes of experiments as function of free parameters, perform experiments to determine them.

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Cure: Discard infinities in a systematic way to extract finite parts.

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Some reactions

Not everybody liked these techniques...



“This is just not sensible mathematics. Sensible mathematics involves neglecting a quantity when it is small - not neglecting it just because it is infinitely great and you do not want it.” – Paul Dirac

More reactions

... not even those who developed them!



"The shell game that we play [...] is technically called 'renormalization'. But no matter how clever the word, it is still what I would call a dippy process!" – Richard Feynman

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Renormalisability

Some models are perturbatively renormalisable: at every order, parameters can be adjusted (in a diverging way!) to provide consistent answers.

Outcome: Theory with as many parameters as the naïve model.

Moral: “Form” of a model matters, not finiteness of constants.



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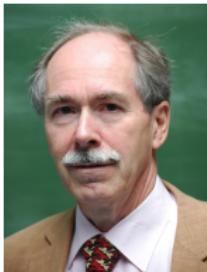
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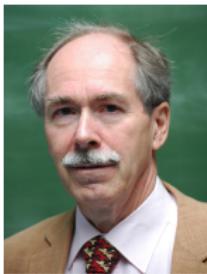
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Example

Try to define distribution “ $\eta(x) = \frac{1}{|x|}$ ”.

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

Formal definition:

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

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

Approximation: $1/(\varepsilon + |x|) - 2|\log \varepsilon| \delta(x)$ converges to η_c for some c .

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Stochastics / Finance

Random walk: $W_{t+\varepsilon}^\varepsilon = W_t^\varepsilon + \sqrt{\varepsilon} \xi_t$ with $\{\xi_t\}$ independent identically distributed random variables, zero mean, unit variance.
Donsker's invariance principle: $W_t^\varepsilon \rightarrow W_t$ with W_t a Brownian motion.

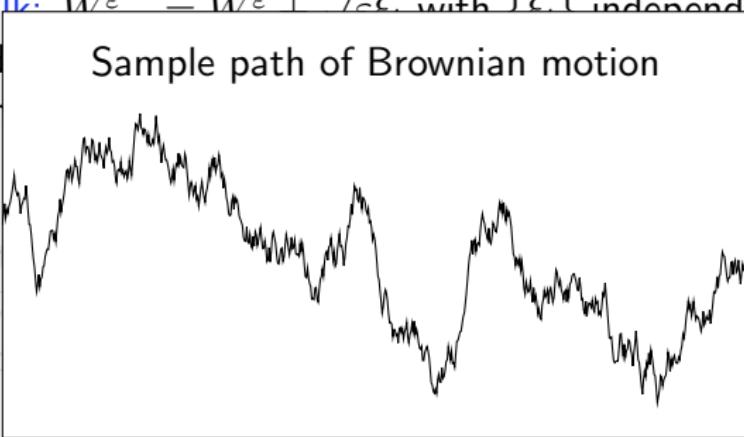
Simple asset price model: $S_{t+\varepsilon}^\varepsilon = S_t^\varepsilon (1 + \sqrt{\varepsilon} \xi_t)$. (So $\delta S_t^\varepsilon = S_t^\varepsilon \delta W_t^\varepsilon$.) Formally, one expects in the limit $\varepsilon \rightarrow 0$ to have $dS/dt = S dW/dt$, so that $S_t = S_0 \exp(W_t)$.

Wrong: The limit satisfies $S_t = S_0 \exp(W_t - t/2)$. (Can be “guessed” from $\mathbf{E} S_t = \mathbf{E} S_0$.)

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Random walk. $W_t^\varepsilon = W_{t-}^\varepsilon + \sqrt{\varepsilon} \cdot \text{f.c. with f.c. independent}$
identically distributed increments. Donsker's invariance principle. Brownian motion.

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What went wrong??

Problem: While S^ε converges to a limit and dW^ε/dt converges to a limit, these are **too rough** for their product to be well-posed.

In general $(f, \xi) \mapsto f \cdot \xi$ well-posed on $\mathcal{C}^\alpha \times \mathcal{C}^\beta$ if and **only if** $\alpha + \beta > 0$. Here: just below borderline.

Consequence: Limit depends on details of discretisation. For example, if we set instead $S_{t+\varepsilon} = \frac{S_t + S_{t+\varepsilon}}{2}(1 + \sqrt{\varepsilon}\xi_t)$, then $S_t^\varepsilon \rightarrow S_0 \exp(W_t)$ as expected. In general: one-parameter family $S_0 \exp(W_t - ct)$ with $c \in \mathbf{R}$.

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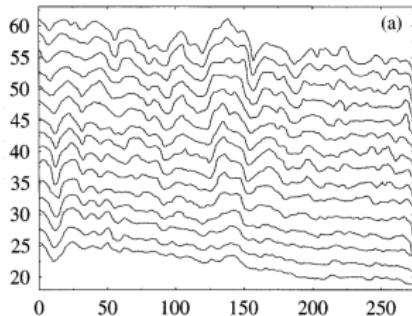
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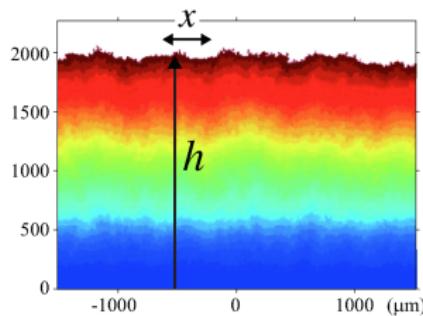
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Universality

Universality: Many random systems “look the same” and are “scale invariant” when viewed at scales much larger than that of the mechanism producing them, provided that they share some basic features:



Maunuksela & Al, PRL

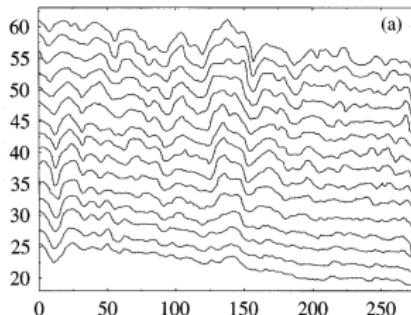


Takeuchi & Al, Sci. Rep.

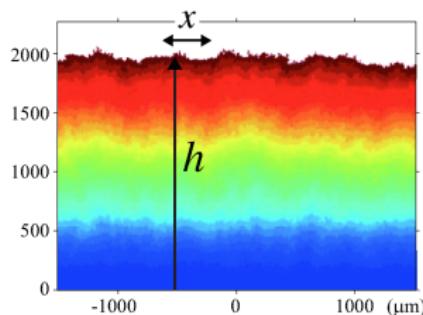
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Crossover regimes

Described by simple “normal form” equations:

$$\partial_t h = \partial_x^2 h + (\partial_x h)^2 + \xi - C , \quad (\text{KPZ}; d = 1)$$

$$\partial_t \Phi = -\Delta(\Delta \Phi + C \Phi - \Phi^3) + \nabla \xi . \quad (\Phi^4; d = 2, 3)$$

Here ξ is space-time white noise (think of independent random variables at every space-time point).

KPZ: universal model for weakly asymmetric interface growth.

Φ^4 : universal model for phase coexistence near mean-field.

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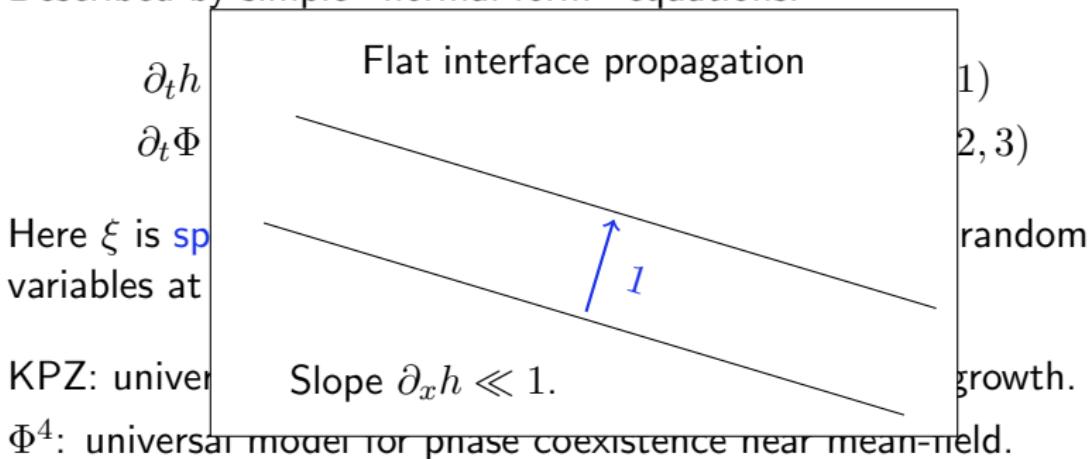
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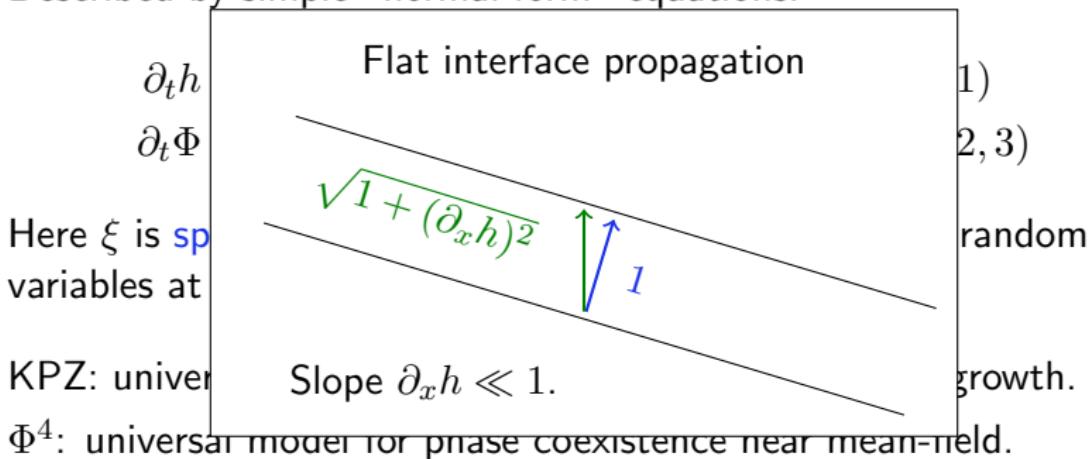
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Well-posedness results

Write ξ_ε for mollified version of space-time white noise. Consider

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(Periodic boundary conditions on torus / circle.)

Theorem (H. 2013): There are $C_\varepsilon \rightarrow \infty$ so that solutions converge to a limit independent of the regularisation. (The constants themselves do depend on that choice.) For KPZ, limit coincides with Cole-Hopf solution (if C_ε is chosen appropriately).

Corollary of proof: Rates of convergence, precise local description of limit, suitable continuity, etc.

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Universality result for KPZ

(In progress; joint with J. Quastel.) Consider

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

with P even polynomial, ξ smooth space-time Gaussian field with compactly supported correlations.

Theorem: As $\varepsilon \rightarrow 0$, there is a choice of $C_\varepsilon \sim \varepsilon^{-1}$ such that $\varepsilon^{1/2} h(\varepsilon^{-1}x, \varepsilon^{-2}t) - C_\varepsilon t$ converges to solutions to $(\text{KPZ})_\lambda$ with λ depending in a non-trivial way on all coefficients of P .

Remark: Convergence to KPZ with $\lambda \neq 0$ even if $P(u) = u^4$!!

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Main Idea

Problem: Solutions are not smooth.

Insight: What is “smoothness”? Proximity to polynomials; we know how to multiply these...

Idea: Replace polynomials by a (finite / countable) collection of tailor-made space-time functions / distributions with similar algebraic / analytic properties. Depends on the realisation of the noise, but not on “details” of the equation. (Values of constants, initial condition, boundary conditions, etc.)

Amazing fact: If we chose the objects replacing polynomials in a smart way, these very singular solutions are “smooth”!

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Insight: What is “smoothness”? Proximity to polynomials; we know how to multiply these...

Idea: Replace polynomials by a (finite / countable) collection of **tailor-made** space-time functions / distributions with similar algebraic / analytic properties. Depends on the realisation of the noise, but not on “details” of the equation. (Values of constants, initial condition, boundary conditions, etc.)

Amazing fact: If we chose the objects replacing polynomials in a smart way, these very singular solutions are “smooth”!

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General picture

Method of proof: Build objects for the following diagram:

$$\begin{array}{ccccc} \mathcal{F} & \times & \mathcal{M} & \times & \mathcal{C}^\alpha(\mathbf{R}^d) \\ \downarrow & & \uparrow \Psi & & \downarrow \\ \mathcal{F} & \times & ? & \times & \mathcal{C}^\alpha(\mathbf{R}^d) \\ \text{?} & & \uparrow \Psi & & \uparrow u_0 \\ & & \xi_\varepsilon & & \end{array} \xrightarrow{\quad S_A \quad} \mathcal{D}^\gamma \quad \quad \quad \xrightarrow{\quad S_C \quad} \mathcal{S}'(\mathbf{R}^{d+1})$$

$\downarrow \mathcal{R}$

\mathcal{F} : Formal right-hand side of the equation.

S_C : Classical solution to the PDE with smooth input.

S_A : Abstract fixed point: locally jointly continuous!

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Strategy: find $M_\varepsilon \in \mathfrak{R}$ such that $M_\varepsilon \Psi(\xi_\varepsilon)$ converges.

Some concluding remarks

1. Diverging terms can (sometimes) be **cured** by suitable **counterterms**, in probabilistic models, not just in QFT.
2. Forces one to deal with families of models parametrised by constants defined “up to an infinite part”. Not a problem as soon as “observables” are finite and answers are consistent...
3. Goal: obtain “universality” results for models from statistical mechanics with tuneable parameters. (Current theory works well for continuous rather than discrete models.)

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