Regularization by oscillations

Massimiliano Gubinelli

CEREMADE Université Paris Dauphine

Ascona, May 26th, 2011

Outline

I would like to discuss three different regularization phenomena due to the presence of noise which however share similar structural properties.

- \triangleright Davie's theorem for SDEs
- \triangleright Stochastic Burgers equation
- \triangleright Schrödinger equation with random dispersion

Davie's phenomenon

Consider this integral equation in R*^d* :

$$
x_t = \int_0^t b(s, x_s) \mathrm{d} s + w_t, \qquad 0 \leqslant t \leqslant 1
$$

where $w \in C([0,1]; \mathbb{R}^d)$ is a path picked according Wiener measure and *b* is a generic **bounded** vectorfield.

A. M. Davie showed that there exist a full measure set Γ ⊂ *C*([0, 1]; R*^d*) such that every $w \in \Gamma$ admits only one solution $x \in C([0,1]; \mathbb{R}^d)$ to the integral equation.

Easy consequence: any discrete scheme will converge to this solution (in particular non-adapted schemes).

Related work: [Veretennikov, Krylov-Röckner, Flandoli-Priola-G.]

Smoothing effect of typical brownian paths

Let $x_t = u_t + w_t$:

$$
u_t = \int_0^t b(s, w_s + u_s) \mathrm{d} s, \qquad 0 \leqslant t \leqslant 1
$$

Key fact

$$
\mathbb{E}\left[\left(\int_s^t (b(r,W_r+x)-b(r,W_r+y))\mathrm{d}r\right)^p\right]\leqslant C_p|x-y|^p|t-s|^{p/2}
$$

 \Rightarrow The random field $(t, s, x) \mapsto \int_s^t b(r, W_r + x) dr$ is almost surely almost **Lipshitz** in the space variable.

 \Rightarrow Via an approximation argument this results in uniqueness for the ODE.

A simple argument, or the Itô trick

Consider the Fourier transform of local time of a Brownian motion. By Itô formula

$$
Y(t,\xi)=\int_0^te^{i\xi\cdot W_s}\text{d} s=2\frac{e^{i\xi\cdot W_t}-e^{i\xi\cdot W_0}}{|\xi|^2}-\frac{2i\xi}{|\xi|^2}\cdot\int_0^te^{i\xi\cdot W_s}\text{d} W_s
$$

And for large ξ ∈ R*^d*

$$
\mathbb{E}\left[|Y(t,\xi)|^p\right] \leqslant C_p|\xi|^{-p}|t|^{p/2}
$$

so there is a gain of one power of ξ with respect to a trivial estimate.

Alternative approach using Gaussian computations

$$
\mathbb{E}\left[|Y(t,\xi)|^2\right] = \int_0^t \int_0^t \mathbb{E}[e^{i\xi \cdot (B_r - B_s)}] ds dr
$$

$$
= 2 \int_0^t \int_0^r e^{-|\xi|^2 (r-s)/2} ds dr \lesssim \frac{t}{|\xi|}
$$

⇒ Extensions to fBM [ongoing work with R. Catellier].

Stochastic Burgers equation

[joint work with M. Jara]

Here the stochastic Burgers equation on $\mathbb{T} = [-\pi, \pi]$

$$
du_t = \frac{1}{2} \partial_{\xi}^2 u_t(\xi) dt + \frac{1}{2} \partial_{\xi} (u_t(\xi))^2 dt + \partial_{\xi} dW_t
$$

where $W_t(\xi) = \sum_{k \in \mathbb{Z}_0} e_k(\xi) \beta_t^k$ with $\mathbb{Z}_0 = \mathbb{Z} \setminus \{0\}$, $e_k(\xi) = e^{ik\xi} / \sqrt{2\pi}$ and $(\beta_t^k)_{t\geq0, k\in\mathbb{Z}_0}$ are complex Brownian motions with $(\beta_t^k)^* = \beta_t^{-k}$ and covariance $\mathbb{E}[\beta_t^k \beta_t^q] = \delta_{q+k=0} t.$

The solution *u* would like to be the derivative of the solution of the Kardar–Parisi–Zhang equation

$$
dh_t = \frac{1}{2}\partial_{\xi}^2 h_t(\xi)dt + \frac{1}{2}(\partial_{\xi}h_t(\xi))^2dt + dW_t.
$$
 (1)

which is believed to capture the macroscopic behavior of a large class of surface growth phenomena.

Problems with the weak formulation

For sufficiently smooth test functions $\varphi : \mathbb{T} \to \mathbb{R}$ look for solutions of

$$
u_t(\varphi) = u_0(\varphi) + \int_0^t u_s(\partial_{\xi}^2 \varphi) ds + \int_0^t \langle \partial_{\xi} \varphi, B(u_s) \rangle ds + W_t(\partial_{\xi} \varphi)
$$

 $where B(u_s)(\xi) = (u_s(\xi))^2.$

- \blacktriangleright We would like to start the equation from initial condition u_0 which is space white noise, this is expected to be an invariant measure.
- \blacktriangleright The linearized equation

$$
X_t(\varphi) = u_0(\varphi) + \int_0^t X_s(\partial_{\xi}^2 \varphi) ds + W_t(\partial_{\xi} \varphi)
$$

has trajectories which look like white noise in space.

 \Rightarrow The nonlinear term $B(u_s)$ is not defined.

Smoothing

Write $u_t = X_t + v_t$, then

$$
v_t(\varphi) = \int_0^t v_s(\partial_{\xi}^2 \varphi) ds + \int_0^t \langle \partial_{\xi} \varphi, B(X_s + v_s) \rangle ds
$$

The covariance of the OU process is

$$
\mathbb{E}[X_t(e_k)X_s(e_m)] = \delta_{m+k=0}e^{-m^2|t-s|/2}
$$

Key fact

The quantity

$$
\int_0^t \langle \partial_{\xi} \varphi, B(X_s) \rangle ds
$$

is well defined due to the rapid space-time decorrelation of the OU process.

Lazy smoothing estimation

Call "good" a process *y* such that

$$
y_t(\phi) = y_0(\phi) + \int_0^t v_s(\partial_{\xi}^2 \phi) ds + \mathcal{A}_t(\phi) + W_t(\partial_{\xi} \phi)
$$

where

- \blacktriangleright $\mathcal{A}_t(\varphi)$ is a zero-quadratic variation process
- \blacktriangleright y_t is space-time white noise at all times
- **►** The reversed process $\hat{y}_t = y_{T-t}$ has the same properties with drift $\widehat{A} = -A$.

Forward/backward Itô trick

Adding Itô formula for the finite quadratic variation process *y*

$$
h(y_t) = h(y_0) + \int_0^t L^0 h(y_s) ds + \int_0^t Dh(y_s) dA_s + M_t^+
$$

(here L^0 is the OU generator) with Itô formula for the backward process

$$
h(y_{T-t}) = h(y_T) + \int_T^{T-t} L^0 h(y_{T-s}) ds - \int_T^{T-t} Dh(y_{T-s}) dA_{T-s} + M_t^{-}
$$

gives

$$
M_t^+ - M_{T-t}^- + M_T^- = \int_0^t 2L^0 h(y_s) ds
$$

Easy to find an H such that 2L $^{0}H = \mathfrak{d}_{\xi}B$ which allows to replace the Burgers drift

 \int_0^t 0 ∂ξ*B*(*ys*)d*s*

with a sum of forward and backward martingales such that

$$
\langle M^{\pm}(\varphi) \rangle_T = \int_0^T \mathcal{E}(\langle \varphi, H(\cdot) \rangle)(y_s) ds
$$

where

$$
\mathcal{E}(h)(x) = \sum_{q \in \mathbb{Z}_0} q^2 (D_q h)(x)^2.
$$

The function $\mathcal{E}(\langle \varphi, H(\cdot) \rangle)(\psi_s)$ is now well defined for ψ_s sampled according white noise and we can estimate it.

Formulation of the equation

Let $B_{\varepsilon}(x) = B(\rho_{\varepsilon} * x)$ a regularization of the non-linearity.

By previous arguments we have that for good processes *y* this limit exists

$$
\lim_{\varepsilon\to 0}\int_0^t\langle\varphi,\partial_{\xi}B_{\varepsilon}(y_s)\rangle ds=\mathcal{B}_t(\varphi)
$$

and we can use it to define the drift in the Burgers equation.

A solution *u* of the Burgers equation is a good process such that

$$
u_t(\varphi) = u_0(\varphi) + \int_0^t u_s(\partial_{\xi}^2 \varphi) ds + \mathcal{B}_t(\varphi) + W_t(\partial_{\xi} \varphi)
$$

The Itô trick provides compactness estimates for Galerkin approximation. Uniqueness is open (in this approach).

The process $\mathcal{B}_t(\varphi)$ is only 3/2– Hölder in time.

Schrödinger equation with random dispersion

Consider the (Stratonovich-) stochastic Schrödinger equation

 $d\Phi_t = i\Delta \Phi_t \circ dB_t + |\Phi_t|^2 \Phi_t dt$

for ϕ : $[0, T] \times \mathbb{T} \to \mathbb{C}$.

[Debussche–De Bouard]

Let

$$
U_t=e^{i\Delta B_t}
$$

so that distributionally

$$
dU_t = i\Delta U_t \circ dB_t
$$

and set $\psi_t = U_t^{-1} \phi_t$. Then

$$
\psi_t = \psi_0 + \int_0^t U_s^{-1} (|U_s \psi_s|^2 U_s \psi_s) ds.
$$

Regularization

Define

$$
X_t(\theta) = \int_0^t U_s^{-1}(|U_s\theta|^2 U_s\theta)ds
$$

It turns out that this random map has the following pathwise regularity

$$
||X_t(\theta) - X_s(\theta)||_{L^2(\mathbb{T})} \lesssim |t - s|^{\gamma} ||\theta||_{L^2(\mathbb{T})}^3
$$

for some $\gamma > 1/2$.

Then take any $\theta \in C^{\gamma}([0, T], L^2(\mathbb{T}))$ and consider

$$
\lim_{\Delta t \to 0} \sum_i \left[X_{t_{i+1}}(\theta_{t_i}) - X_{t_i}(\theta_{t_i}) \right] = \int_0^t X_{ds}(\theta_s).
$$

By Young theory we have the estimate

$$
\left\|\int_0^{\cdot} X_{\rm ds}(\theta_s)\right\|_{C^{\gamma}([0,T],L^2(\mathbb{T}))} \lesssim_X \|\theta\|_{C^{\gamma}([0,T],L^2(\mathbb{T}))}^3.
$$

Existence and uniqueness of global solutions

The Schrödinger equation can be reformulated as a Young equation

$$
\psi_t = \psi_0 + \int_0^t X_{ds}(\psi_s)
$$

giving rise to local solutions

$$
\psi \in C^{\gamma}([0,T_*],L^2(\mathbb{T})).
$$

The *L* 2 conservation law allows to obtain global solutions.

Standard arguments for Young equations allows to prove convergence of the Euler scheme

$$
\psi_{t_{i+1}} = \psi_{t_i} + [X_{t_{i+1}}(\psi_{t_i}) - X_{t_i}(\psi_{t_i})]
$$

Similar phenomena for the deterministic KdV equation

$$
\partial_t u_t = \partial_{\xi}^3 u_t + \partial_{\xi} u_t^2.
$$

with initial conditions in $H^{\alpha}(\mathbb{T})$ for $\alpha > -1/2$. [see paper on CPAA]

Thanks