

Stochastic regularization effects of semi-martingales on random functions

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Stochastic regularization

Itô-Wentzell-Tanaka trick

Stochastic regularization in a nutshell

The following slides are based on the lecture notes of Franco Flandoli (2015) and on his St. Flour lecture Notes "Random Perturbation of PDEs and Fluid Dynamic Models" (2010).

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The map u is smooth and solves the Heat equation:

$$\frac{\partial u}{\partial t} = \frac{1}{2} \Delta u, \quad u(0, \cdot) = \varphi(\cdot),$$

and

$$u(t, x) = \int_{\mathbb{R}^d} P_t^{\text{heat}}(x - y) \varphi(y) dy.$$

A second example

- Consider the following ODE:

$$dX_t = b(t, X_t)dt, \quad X_0 = x_0,$$

for some $b : [0, T] \times \mathbb{R}^d \rightarrow \mathbb{R}^d$.

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- When b is not smooth, uniqueness may fail...
- Take for instance $d = 1$ and $b(t, x) := b(x) := 2\operatorname{sgn}(x)\sqrt{|x|}$ and $x_0 := 0$, then every function of the form

$$X_t := \pm(t - t_0)^2 \mathbf{1}_{t \geq t_0}, \quad t_0 \geq 0$$

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What is then a good solution?

A second example

- Add some noise:

$$dX_t = b(t, X_t)dt + \sigma dB_t, \quad X_0 = x_0,$$

where $(B_t)_t$ is a Brownian motion and $\sigma > 0$.

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- Why is it useful?
- **Selection of solutions:** Assume that for any σ there exists a unique solution, then let \mathbb{P}_σ denotes its law. Then prove that $(\mathbb{P}_\sigma)_{\sigma>0}$ is tight and converges in law (as σ tends to 0) to some measure supported on the set of solutions to the ODE.

For instance, Bafico and Baldi (81') proved that for $b(x) = 2\text{sgn}(x)\sqrt{|x|}$ and $x_0 = 0$ it converges to:

$$\frac{1}{2}\delta_{+t^2} + \frac{1}{2}\delta_{-t^2}.$$

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- (Veretennikov 81') If b is bounded then the equation admits pathwise uniqueness.
- (Krylov-Röckner 05') If b belongs to $L^q([0, T]; L^p(\mathbb{R}^d))$ with $\frac{d}{p} + \frac{2}{q} < 1$ ($p, q \geq 2$) then the equation admits pathwise uniqueness.

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- How does it work?

A second example

- Recall that:

$$X_t = x_0 + \int_0^t b(s, X_s) ds + \sigma B_t$$

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We try to get regularity of the blue term using the Itô-Tanaka Trick.

- Example: use the celebrated Itô-Tanaka formula for $b = \delta_a$ and for B :

$$\int_0^t \delta_a(B_s) ds = |B_t - a| - |a| - \int_0^t \operatorname{sgn}(B_s - a) dB_s.$$

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- Idea: to express $\int_0^t b(s, X_s) ds$ by means of more regular objects

The Itô-Tanaka trick

- Apply Itô's formula with a smooth mapping U :

$$U(t, X_t) = U(T, X_T) - \int_t^T \left(\frac{\partial U}{\partial t} + b \cdot \nabla U + \frac{1}{2} \sigma^2 \Delta U \right) (s, X_s) ds \\ - \sigma \int_t^T \nabla U(s, X_s) dB_s$$

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- So if U is solution to the Fokker-Planck (Backward) PDE

$$\frac{\partial U}{\partial t} + b \cdot \nabla U + \frac{\sigma^2}{2} \Delta U = -b, \quad U(T, x) = 0,$$

then

$$\int_t^T b(s, X_s) ds = -U(t, X_t) + \sigma \int_t^T \nabla U(s, X_s) dB_s$$

and so

$$X_t = x_0 + U(0, x_0) - U(t, X_t) + \sigma \int_0^t (\nabla U(s, X_s) + Id.) dB_s.$$

Applications of the Itô-Tanaka trick to SPDEs

- The Itô-Tanaka Trick can be used to obtain new results in linear transport equations by introducing a stochastic perturbation (see *Flandoli, Gubinelli, Priola; 10'; Invent. Math.*).

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- Limitation to other problems: ([Flandoli et al.](#))

"The generalization to nonlinear transport equations, where b depends on u itself, would be a major next step for applications to fluid dynamics but it turns out to be a difficult problem. Specifically there are already some difficulties in dealing with a vector field b which depends itself on the random perturbation W . There is no obvious extension of the Itô-Tanaka trick to integrals of the form $\int_0^T f(\omega, s, X_s^x(\omega)) ds$ with random f ."

Stochastic regularization

Itô-Wentzell-Tanaka trick

Generalizations to random mappings

The problem pointed out previously is to provide an expression for:

$$\int_0^T f(s, \omega, X_s) ds,$$

where f is now random (previously we had $f = b$ where b was deterministic) in a predictable way.

- If we reproduce the ideas before we need to consider the Fokker-Planck SPDE:

$$U(t, x) = - \int_t^T \left(\frac{1}{2} \Delta + b(s, \omega, x) \cdot \nabla \right) U(s, x) ds - \int_t^T f(s, \omega, x) ds.$$

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- **But:** in that case $U(t, x)$ is not adapted (even if the data b, f are adapted) so you can not use classical Itô calculus and the previous approach fails.

Generalizations to random mappings

- **Idea:** make it adapted, and consider rather the following Fokker-Planck **BSPDE**:

$$U^a(t, x) = - \int_t^T \mathcal{L}_s U^a(s, x) ds - \int_t^T f(s, \omega, x) ds - \int_t^T Z(s, x) dB_s,$$

$$\text{with } \mathcal{L}_s := \frac{1}{2} \Delta + b(s, \omega, x) \cdot \nabla.$$

If solvable, U^a and Z are two predictable processes.

Itô-Wentzell-Tanaka trick

Theorem (Duboscq, R.)

Assume that U^a and Z exist and are regular enough, then

$$\int_0^T f(s, \omega, X_s) ds = -U^a(0, X_0) - \int_0^T (\nabla U^a(s, X_s) + Z(s, X_s)) dB_s \\ - \int_0^T \nabla Z(s, X_s) ds, \mathbb{P} - a.s..$$

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Now we need to study the BSPDE and the regularity of (U^a, Z) .

Analysis of the BSPDE

Theorem (Duboscq, R.)

Let $p, q \geq 2$. Assume that b, f are adapted and that f belongs to a " $L^p - L^q$ space" and is Malliavin differentiable. There exists a unique strong (predictable) solution to the Fokker-Planck BSPDE

$$(U^a, Z) \in ("L^p - L^q \text{ space}")^2.$$

Futhermore, we have the following representation of U^a

$$U^a(t, x) = \mathbb{E} \left[- \int_t^T P_{t,r}^X f(r, x) dr \middle| \mathcal{F}_t \right]. \quad (1)$$

In addition, for a.e. (t, x) , $U^a(t, x)$ is Malliavin differentiable, and for a.e. $x \in \mathbb{R}^d$, a version of the process $(Z(t, x))_{t \in [0, T]}$ is given by

$$Z(t, x) = D_t U^a(t, x) = \mathbb{E} \left[- \int_t^T D_t P_{t,r}^X f(r, x) dr \middle| \mathcal{F}_t \right]. \quad (2)$$

Analysis of the BSPDE

Theorem (Duboscq, R.)

... Finally, U^a admits the following mild (a.k.a. Duhamel's formula) representation

$$U^a(t, x) = - \int_t^T P_{t,r}^X f(r, x) dr - \int_t^T P_{t,r}^X Z(r, x) dB_r, \quad (3)$$

where $P_{s,t}^X \phi$ is the unique solution to:

$$P_{s,t}^X \phi(x) = \phi(x) - \int_s^t \mathcal{L}_r P_{r,t}^X \phi(x) dr, \quad 0 \leq s \leq t.$$

Analysis of the BSPDE

Remarks

- We are not working in L^2
- We provide an explicit representation which is a counterpart of the one for linear BSDEs (**no reversibility of the semigroup**)
- Malliavin differentiability in $L^p - L^q$ spaces is not completely trivial...there are catches
- Duhamel's formula in that context is new