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SPDE and Applications - X, Levico Terme

May 30-June 4, 2016





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■ *p* = 1

Motivation: The linear case

Nonlinear, singular SPDE with gradient-type Gaussian noise driving infinitesimal vector field actions Motivation: The linear case

Zakaï equation

Zakaï equation

Consider the following linear SPDE in a separable Hilbert space $H, t \in [0, T]$:

$$dX_t = AX_t dt + \sum_{i=1}^N B_i X_t d\beta_t^i, \quad X_0 = x \in H,$$
(1)

for β^i , i = 1, ..., N, independent Brownian motions on $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}, \mathbb{P})$.

A, B_i , i = 1, ..., N are unbounded linear operators on H.

Sufficient conditions on A and the B_i such that there exist strong solutions to (1) are given e.g. in [Da Prato & Zabczyk, Cambridge Univ. Press (1992), Chapter 6.5].

Following [Da Prato, Iannelli & Tubaro, Univ. Padova (1982)], [Da Prato, Iannelli & Tubaro, Stochastics (1982)], one obtains strong solutions to (1), whenever

- A generates a C_0 -semigroup e^{tA} .
- The B_i , i = 1, ..., N generate mutually commuting C_0 -groups e^{tB_i} , $t \in \mathbb{R}$, i = 1, ..., N.
- For every i = 1, ..., N, $D(B_i^2) \supset D(A)$, and $\bigcap_{i=1}^N D((B_i^*)^2)$ is a dense subset of H.
- The operator $C := A \frac{1}{2} \sum_{i=1}^{N} B_i^2$, D(C) := D(A) generates a C_0 -semigroup e^{tC} , $t \ge 0$.

Compare also with [Tubaro, Stoch. Anal. Appl. (1988)], where the commutation assumption was removed using Kunita's method of *stochastic characteristics*.

Motivation: The linear case

– Zakaï equation

Idea

Set

$$U_t := \prod_{i=1}^N e^{\beta_t^i B_i}, \quad t \in [0, T],$$

and let Y_t be the solution to the time-dependent random PDE

$$\frac{d}{dt}Y_t = U_t^{-1}CU_tY_t, \quad Y_0 = x.$$
(2)

Then $X_t := U_t Y_t$, $t \ge 0$ is a solution to (1).

Note that:

• We can \mathbb{P} -a.s. find strong solutions Y to (2) such that $t \mapsto Y_t$ is predictable.

In this case, X (as above) takes values in D(C) ℙ ⊗ dt-a.s. and is a strong solution to (1).

Motivation: The linear case

-Zakaï equation

Example (strongly elliptic case)

Let $H = L^2(\mathbb{R}^d)$, $d \in \mathbb{N}$ and N = 1. Set

$$Ay := \sum_{i,j=1}^{d} a_{i,j} \partial_i \partial_j y + \sum_{i=1}^{d} q_i \partial_i y + ry, \quad y \in H^2(\mathbb{R}^d) =: D(A),$$

$$By = \sum_{i=1}^d b_i \partial_i y + cy, \quad y \in D(B) = \{ y \in L^2(\mathbb{R}^d) \mid By \in L^2(\mathbb{R}^d) \}.$$

Assume for simplicity that $a_{i,j}$, q_i , r, b_i , c are all C^3 and bounded with bounded derivatives up to order 3. Assume that there exists a constant $\gamma > 0$ such that

$$\sum_{i,j=1}^d (a_{i,j} - rac{1}{2}b_i b_j)\lambda_i\lambda_j \geqslant \gamma \sum_{i=1}^d \lambda_i^2,$$

for all $\lambda = (\lambda_1, \dots, \lambda_d) \in \mathbb{R}^d$. Then the above conditions are satisfied.

Motivation: The linear case

– Zakaï equation

Can we extend this method?

Both the commutation assumption and the ellipticity of $C := A - \frac{1}{2} \sum_{i=1}^{N} B_i^2$ seem very restrictive.

Some observations:

- In view of the Trotter product formula, the commutation seems natural.
- $\frac{1}{2} \sum_{i=1}^{N} B_i^2 X \, dt$ is exactly the Itô-Stratonovich correction term of $\sum_{i=1}^{N} B_i X \, d\beta^i.$
- One could also define the B_i weakly in a Gelfand triple $H^1 \subset L^2 \subset (H^1)^*$.
- The transformation in (2) also makes sense, when the operator A (C resp.) is nonlinear, see [Barbu, Brzeźniak, Hausenblas & Tubaro, Stoch. Processes Appl. (2013)], [Barbu & Röckner, J. Eur. Math. Soc. (2015)]!
- One has to ensure that $(U_t)_{t \ge 0}$ is in some sense compatible with A.

Framework and the SPDE

Framework and the SPDE

Stochastic singular *p*-Laplace equations

Stochastic singular *p*-Laplace equations

Consider the nonlinear Stratonovich SPDE in $L^{2}(\mathcal{O})$,

$$dX_t = \operatorname{div}[\Psi(\nabla X_t)] dt + \sum_{i=1}^N \sum_{j=1}^d b_j^i \partial_j X_t \circ d\beta_t^i, \quad X_0 = x \in L^2(\mathcal{O}),$$
(3)

where $\mathcal{O} \subset \mathbb{R}^d$, $d \ge 2$ is an open bounded convex C^3 -smooth domain and $b: \overline{\mathcal{O}} \to \mathbb{R}^{N \times d}$ is a C^2 -smooth "coefficient field". Assume Neumann boundary conditions.

 β^{i} , i = 1, ..., N, are independent Brownian motions on $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}, \mathbb{P})$.

For reasons made clear later, we shall also assume that

- $\Psi = \partial(\frac{1}{p}|\cdot|^p)$, where $p \in [1, 2]$. However, let us first assume for simplicity that $p \in (1, 2)$.
- The "row operators" $\langle b_i, \nabla \cdot \rangle$ commute mutually (or N = 1).
- The "row operators" $\langle b_i, \nabla \cdot \rangle$ commute (weakly) with the Neumann Laplace.

- Framework and the SPDE
 - The noise coefficient operators

Commutation

Let b_i be some row of b. Define the "row operators"

$$B_i u := \sum_{j=1}^d b_i^j \partial_j u, \quad u \in H^1(\mathcal{O}).$$

By [Sumitomo, Hokkaido Math. J. (1972)], it is necessary and sufficient for B_i to commute with the Laplace-Beltrami operator on smooth functions that

 b_i is a Killing vector field,

meaning that, the Jakobian of b_i is skew-symmetric, i.e.

$$\partial_j b_i^k + \partial_k b_i^l = 0 \quad \forall 1 \leqslant j, k \leqslant d.$$
(4)

This automatically implies that div $b_i = 0$.

Framework and the SPDE

The noise coefficient operators

Theorem

Assume that $b_i \in C^2$, $1 \leq i \leq N$, with

1 $\langle b_i, \nu \rangle = 0$ on $\partial \mathcal{O}$ for all *i*, where ν denotes the outer normal on $\partial \mathcal{O}$,

2
$$N = 1$$
 or $\langle b_i, \nabla b_i^l \rangle = \langle b_j, \nabla b_i^l \rangle$ for $i \neq j$ and all $1 \leq l \leq d$ on $\overline{\mathcal{O}}$,

3
$$Db_i$$
 is skew-symmetric for all i on $\overline{\mathcal{O}}$.

Then B_i leaves Neumann boundary conditions invariant (on a core) for every *i*.

Also, for every $u \in H^1(\mathcal{O})$

$$B_i J_{\delta} u = J_{\delta} B_i u \quad \forall 1 \leqslant i \leqslant N \quad \forall \delta > 0.$$
 (Comm)

Here, $J_{\delta} = (\operatorname{Id} - \delta \Delta)^{-1}$ denotes the resolvent of the Neumann Laplace $-\Delta$.

- Framework and the SPDE
 - The noise coefficient operators

Shigekawa's result

Theorem ([Shigekawa, Acta Appl. Math. (2000)])

Fix $1 \leq i \leq N$. (Comm) is implied by the following: Suppose there exists a linear subspace $\mathcal{D} \subset \text{dom}(-\Delta)$ such that the following conditions hold:

- $\square \Delta(\mathcal{D}) \subseteq \operatorname{dom}(B_i),$
- $2 \quad B_i(\mathcal{D}) \subseteq \operatorname{dom}(-\Delta),$
- **3** \mathcal{D} is a core for $(-\Delta, \operatorname{dom}(-\Delta))$,
- 4 dom $(-\Delta) \subseteq$ dom (B_i) and dom $(-\Delta) \subseteq$ dom (B_i^*) ,
- 5 for any $u \in \mathcal{D}$, it holds that

$$B_i \Delta u = \Delta B_i u.$$

Framework and the SPDE

- The *p*-Laplace operator

The *p*-Laplace operator

With $\Psi = \partial(\frac{1}{p}|\cdot|^p)$, in particular, with " ∂ ", we denote the Gâteaux differential for p > 1 (the subdifferential, resp., for p = 1), that is, $\partial(\frac{1}{p}|\cdot|^p)(\xi) = |\xi|^{p-2}\xi, \xi \in \mathbb{R}^d$. We will discuss the case of p = 1 later.

The quasi-linear partial differential operator " $u \mapsto div[\Psi(\nabla u)]$ " is called *p-Laplace* and, in particular, *singular p-Laplace*, if p < 2.

Its negative is an extension (in the sense of monotone graphs) of $F: H^1(\mathcal{O}) \to (H^1(\mathcal{O}))^*$

$$F(u)(v) := \int_{\mathcal{O}} \langle \Psi(\nabla u(\xi)), \nabla v(\xi) \rangle \, d\xi$$

In fact, F is the Gâteaux differential of the convex functional

$$\Phi(u) := \frac{1}{p} \int_{\mathcal{O}} |\nabla u|^p \, d\xi, \quad u \in H^1(\mathcal{O}).$$

We see that the 2-Laplace is just the Neumann Laplace operator.

Framework and the SPDE

Examples

Example

Let
$$d = 3$$
. Let $\mathcal{O} = B_1(0)$ be the 3D unit ball. Let $N = 1$. Let

$$b(\xi) = (\xi_3 - \xi_2, \xi_1 - \xi_3, \xi_2 - \xi_1)$$
 and denote $1 := (1, 1, 1)$ (clearly, $b(\xi) = \xi \times 1$).

Then b is a Killing vector field and (3) becomes

$$\begin{cases} dX_t \in \operatorname{div} \left[\Psi \left(\nabla X_t \right) \right] dt + \langle \xi \times \nabla X_t, 1 \rangle \circ d\beta_t, & \text{in } (0, T) \times \mathcal{O}, \\ X_0 = x, & \text{in } \mathcal{O}, \\ \frac{\partial X_t}{\partial \nu} = 0, & \text{on } (0, T) \times \partial \mathcal{O}. \end{cases}$$

Example

Above, we can take $b : \xi \mapsto \xi \times \zeta_0$ for any $\zeta_0 \in \mathbb{R}^3 \setminus \{0\}$. This is the infinitesimal generator of SO(3), where ζ_0 spans the axis of rotation.

In 2D, we can take the unit disk and $b: (\xi_1, \xi_2) \mapsto (\xi_2, -\xi_1)$, generating SO(2).

Framework and the SPDE

Examples

Example

Let N = d and $b_i^j = \delta_{i,j}$, $1 \le i, j \le d$. Then the above conditions are satisfied on \mathbb{T}^d and (3) reduces to

$$\begin{cases} dX_t \in \operatorname{div} \left[\Psi \left(\nabla X_t \right) \right] dt + \langle \nabla X_t, \circ d\beta_t \rangle, & \text{in } (0, T) \times \mathcal{O}, \\ X_0 = x, & \text{in } \mathcal{O}, \\ \frac{\partial X_t}{\partial \nu} = 0, & \text{on } (0, T) \times \partial \mathcal{O}. \end{cases}$$

This example is relevant in mathematical image processing of binary tomography.

The vector fields b_i are infinitesimal generators of translation groups in the coordinate directions of \mathbb{T}^d .

Framework and the SPDE

- Stochastic variational inequalities (SVI)

Stochastic variational inequalities (SVI)

The Statonovich SPDE

$$dX_t = \operatorname{div}[\Psi(\nabla X_t)] dt + \sum_{i=1}^N \sum_{j=1}^d b_i^j \partial_j X_t \circ d\beta_t^j, \quad X_0 = x \in L^2(\mathcal{O}),$$

is formally equivalent to the Itô SPDE

$$dX_{t} = \operatorname{div}[\Psi(\nabla X_{t})] dt + \frac{1}{2} \sum_{i=1}^{N} B_{i}^{2} X_{t} dt + \sum_{i=1}^{N} B_{i} X_{t} d\beta_{t}^{i}, \quad X_{0} = x \in L^{2}(\mathcal{O}),$$
(SPDE)

where $B_i^2 := -B_i^* B_i$.

Let Z be any process ("test-process") of the form

$$dZ_t = G_t \, dt + \frac{1}{2} \sum_{i=1}^N B_i^2 Z_t \, dt + \sum_{i=1}^N B_i Z_t \, d\beta_t^i,$$

where G is some progressively process taking values in $L^2(\mathcal{O})$ and such that Z takes values in H^1 .

Nonlinear, singular SPDE with gradient-type Gaussian noise driving infinitesimal vector field actions Framework and the SPDE Heuristics for SVI

Heuristics for SVI

Consider d(X - Z) and apply Itô's formula "formally" for $u \mapsto \frac{1}{2} ||u||_{L^2(\mathcal{O})}^2$. We obtain, $\frac{1}{2} \|X_t - Z_t\|_{L^2}^2 = \frac{1}{2} \|x - Z_0\|_{L^2}^2$ + $\int_{-\infty}^{1} (\operatorname{div}[\Psi(\nabla X_s)] - G_s, X_s - Z_s)_{L^2} ds$ $+\frac{1}{2}\sum_{i=1}^{N}\int_{0}^{t}(B_{i}^{2}(X_{s}-Z_{s}),X_{s}-Z_{s})_{L^{2}}ds$ + $\sum_{i=1}^{N} \int_{0}^{t} (X_{s} - Z_{s}, B_{i}(X_{s} - Z_{s}) d\beta_{s}^{i})_{L^{2}}$ $+\frac{1}{2}\sum_{i=1}^{N}\int_{0}^{t}\|B_{i}(X_{s}-Z_{s})\|_{L^{2}}^{2}\,ds.$

However, on a formal level, we have that

$$(B_i^2(X-Z), X-Z)_{L^2} = -\|B_i(X-Z)\|_{L^2}^2.$$

Nonlinear, singular SPDE with gradient-type Gaussian noise driving infinitesimal vector field actions L Framework and the SPDE

Definition of SVI solutions

This motivates the following:

Definition

Let $x \in L^2(\Omega, \mathcal{F}_0, \mathbb{P}; L^2(\mathcal{O})), T > 0$. An $\{\mathcal{F}_t\}$ -progressively measurable process $X \in L^2([0, T] \times \Omega; L^2(\mathcal{O}))$ is called an *SVI-solution* to (SPDE) if $\Phi(X) \in L^1([0, T] \times \Omega)$ and for every $Z \in L^2([0, T] \times \Omega; H^1(\mathcal{O}))$ such that there exist $Z_0 \in L^2(\Omega, \mathcal{F}_0, \mathbb{P}; H^1(\mathcal{O})), G \in L^2([0, T] \times \Omega; L^2(\mathcal{O})), \{\mathcal{F}_t\}$ -progressively measurable, such that the following equality holds $L^2(\mathcal{O})$, that is,

$$Z_t = Z_0 + \int_0^t G_s \, ds + \frac{1}{2} \sum_{i=1}^N \int_0^t B_i^2 Z_s \, ds + \sum_{i=1}^N \int_0^t B_i Z_s \, d\beta_s^i,$$

 \mathbb{P} -a.s. for all $t \in [0, T]$, we have that the following variational inequality holds true:

Framework and the SPDE

Heuristics for SVI

Definition (cont'd)

$$\frac{1}{2}\mathbb{E}\|X_t - Z_t\|_{L^2(\mathcal{O})}^2 + \mathbb{E}\int_0^t \Phi(X_s) \, ds$$

$$\leqslant \frac{1}{2}\mathbb{E}\|x - Z_0\|_{L^2(\mathcal{O})}^2 + \mathbb{E}\int_0^t \Phi(Z_s) \, ds$$

$$- \mathbb{E}\int_0^t (G_s, X_s - Z_s)_{L^2(\mathcal{O})} \, ds,$$

for almost all $t \in [0, T]$.

Moreover, if $X \in L^2(\Omega; C([0, T]; L^2(\mathcal{O})))$, we say that X is a *(time-) continuous SVI solution* to (SPDE).

Main result

The main result

Theorem ([Ciotir & T, to appear in J. Funct. Anal. (2016)])

Let $x \in L^2(\Omega, \mathcal{F}_0.\mathbb{P}; L^2(\mathcal{O}))$. Then there is a unique time-continuous SVI solution $X \in L^2(\Omega; C([0, T]; L^2(\mathcal{O})))$ to (SPDE) in the sense of the previous definition. For two SVI solutions X, Y with initial conditions $x, y \in L^2(\Omega; L^2(\mathcal{O}))$, resp., we have that

$$\operatorname{ess\,sup}_{t\in[0,T]} \mathbb{E} \|X_t - Y_t\|_{L^2(\mathcal{O})}^2 \leq \mathbb{E} \|x - y\|_{L^2(\mathcal{O})}^2.$$

Some Remarks

The existence of time-continuous SVI-solutions is proved via several approximation steps.

The Stratonovich type of the equation, the a priori estimate from the previous slide and the monotone structure of the drift help us to pass to the limit. We also use a Wong-Zakai type convergence result by [Barbu, Brzeźniak, Hausenblas & Tubaro, Stoch. Processes Appl. (2013)].

It is worth mentioning this paper, where similar equations in a setting with more regularity were studied using the semigroup method presented in the beginning of this talk.

Also, in [Barbu & Röckner, J. Eur. Math. Soc. (2015)], the semigroup-transformation approach is used in an infinite dimensional-noise setting. This paper also introduces a product Itô formula for the "stochastic multiplier" $U_t = e^{\sum_{i=1}^N \beta_t^i B_i}$. Compare also with [Munteanu & Röckner, Preprint (2016)].

p = 1

The above methods also work for the case of p = 1, that is, $\Psi = \text{Sgn}$, so that we get the stochastic Stratonovich total variation flow

$$dX_t \in \operatorname{div}[\operatorname{Sgn}(\nabla X_t)] dt + \sum_{i=1}^N \sum_{j=1}^d b_j^i \partial_j X_t \circ d\beta_t^i, \quad X_0 = x \in L^2(\mathcal{O}),$$

which is multi-valued, since Sgn is maximal monotone only if one defines

$$\operatorname{Sgn}(0) := \overline{B}_1(0),$$

and Sgn(ξ) := $\xi/|\xi|$ for $\xi \neq 0$. Hence zeros of the gradient $\nabla X = 0$ create a special situation. We have that Sgn = $\partial |\cdot|$ (the subdifferential of the Euclidean norm).

— Main result

-p = 1

p = 1

However, note that

- The convex energy Φ of "u → − div[Sgn(∇u)]", is the total variation ||Du||(O) of the distributional gradient measure Du, where u ∈ BV(O) is a function of bounded variation in L¹(O) (rather than ∫_O |∇u| dξ, u ∈ W^{1,1}(O), which is not lower semi-continuous in L¹).
- The SVI-framework is taylored for an access to multi-valued equations via the energies (SVI-solutions are even more robust under Mosco-convergence of the energies, see [Gess & T, J. Differential Equations (2016)]).

— Main result

— This talk

This talk was based on [Ciotir & T, to appear in J. Funct. Anal. (2016)]. ¹

Thank you for your attention!²

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