Isotropic Ornstein-Uhlenbeck Flows

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IRTG Summer School Disentis





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- SDEs And Spatial Semimartingales
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Stochastic Flows: A First Example SDEs And Spatial Semimartingales Conclusion

Motivating Example

• Consider the following stochastic differential equation

$$d\begin{pmatrix} X_t\\ Y_t \end{pmatrix} = A\begin{pmatrix} X_t\\ Y_t \end{pmatrix} dt + \begin{pmatrix} 17 & 0\\ 0 & 42 \end{pmatrix} d\begin{pmatrix} W_t^{(1)}\\ W_t^{(2)} \end{pmatrix}$$

$$X_s = x$$
, $Y_s = y$, A is a real matrix.

• The Solution to this equation is of course:

$$\begin{pmatrix} X_t \\ Y_t \end{pmatrix} = e^{(t-s)A} \begin{pmatrix} x \\ y \end{pmatrix} + e^{(t-s)A} \int_0^t e^{-(u-s)A} \begin{pmatrix} 17 \\ 42 \end{pmatrix} dW_u$$

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The flow property

• Consider the solution as a function of the initial value.

$$\Phi_{s,t}: \begin{pmatrix} x \\ y \end{pmatrix} \mapsto e^{(t-s)A} \begin{pmatrix} x \\ y \end{pmatrix} + e^{(t-s)A} \int_0^t e^{-(u-s)A} \begin{pmatrix} 17 \\ 42 \end{pmatrix} dW_u$$

- The function $\Phi = \Phi_{s,t}(\cdot, \omega)$ satisfies:
- it is a diffeomorphism for any ω, s, t
- $\Phi_{t,t}(\cdot,\omega)$ is the identity for all ω and t

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$$\Phi_{s,t}(\cdot,\omega) = \Phi_{u,t}(\cdot,\omega) \circ \Phi_{s,u}(\cdot,\omega)$$

 These properties state that Φ is a stochstic flow of diffeomorphisms.

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Stochastic Flows: A First Example SDEs And Spatial Semimartingales Conclusion

Kunita-Type SDEs

Let is write

$$M(t, \begin{pmatrix} x \\ y \end{pmatrix}) = A \begin{pmatrix} x \\ y \end{pmatrix} t + \begin{pmatrix} 17 & 0 \\ 0 & 42 \end{pmatrix} W_t$$

• Then the SDE becomes

$$d\begin{pmatrix}X_t\\Y_t\end{pmatrix} = M(dt, \begin{pmatrix}X_t\\Y_t\end{pmatrix}), X_s = x, Y_s = y$$

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Semimartingale Fields

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 satisfies:

- it is a semimartingale for fixed x, y
- it has smooth covariations in x, y for fixed t
- the part of finite variaton is smooth
- This states that is a semimartingale field

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Stochstic Flows And SDEs

$$\phi_{s,t}(x) = x + \int_s^t M(du, \phi_{s,u}(x))$$

- The solution of an SDE driven by a sufficiently smooth semimartingale field generates a stochastic flow.
- For a sufficiently smooth stochastic flow there is a semimartingale field that generates it via an SDE.

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Isotropic Covariance Tensors, Isotropic Brownian Fields

Definition

A function $b: \mathbb{R}^d \to \mathbb{R}^{d \times d}$ is an isotropic covariance tensor if:

- **1** $x \mapsto b(x)$ is smooth enough and the derivatives are bounded.
 - 2 $b(0) = E_d$ (the *d*-dimensional identity)
 - 3 $x \mapsto b(x)$ is not constant.

Definition

Let b be as above. $\{M(t,x): t \ge 0, x \in \mathbb{R}^d\}$ is an isotropic Brownian field if:

• (1, x) $\mapsto M(t, x)$ is a centered Gaussian process.

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$$cov(M(s,x), M(t,y)) = (s \wedge t)b(x - y)$$

3 $(t,x) \mapsto M(t,x)$ is continuous for almost all ω .

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$${f 0}~~b(x)=O^*b(Ox)O$$
 for any $x\in {\mathbb R}^d$ and $O\in O(d)$

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Isotropic Brownian Flows (IBFs)

- translation invariance
- rotation invariance
- one-point motion is a *d*-dimensional standard Brownian Motion
- SDEs for two-point-distance,...
- Lyapunov-Exponents are known, deterministic and constant
- Lebesgue measure is invariant
- No straightforward entropy definition

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Introduce a drift into the SDE

$$\phi_{s,t}(x) = x + \int_s^t M(du, \phi_{s,u}(x)) - c \int_s^t \phi_{s,u}(x) du \qquad (1)$$

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Statement Of The Result Sketch Of Proof

Spatial Regularity Lemma

Lemma

2

Let $\phi = \phi_{0,1} : \mathbb{R}^d \to \mathbb{R}^d$ be as in (1) (with s = 0 and t = 1). Then we have a.s.

$\lim_{R \to \infty} \sup_{||x|| \ge R} \frac{||\phi(x) - e^{-c}x||}{||x||} = 0$ (2)

$$\lim_{R \to \infty} \sup_{||x|| \ge R} \frac{||\phi(x)||}{||x||} = e^{-c}$$
(3)

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Reformulation To Compact State Space

• It is sufficient to show

$$\lim_{R \to \infty} \sup_{R \le ||x|| \le R+1} \frac{||\phi(x) - e^{-c}x||}{||x||} = 0$$

•
$$x \mapsto \frac{||\phi(x) - e^{-c}x||}{||x||}$$
 is continuous
• $\mathbb{X} := \{x \in \mathbb{R}^d : R \le ||x|| \le R + 1\}$ is compact

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The Chaining Lemma

$$(\mathbb{X}, d)$$
 compact metric space, $\phi \colon \mathbb{X} \to \mathbb{R}_+$ be a.s. cont. func.,
 $(\delta_i)_{i\geq 0}$ positive real numbers with $\sum_{i=0}^{\infty} \delta_i < \infty$,
 $(\chi_i)_{i=0}^{\infty} \delta_i$ -dense in \mathbb{X} with $\chi_0 = \{x_0\}$, with $d(x, x_0) \leq \delta_0 \forall x \in \mathbb{X}$.

Lemma (Cranston, Scheutzow and Steinsaltz '00)

For arbitrary positive $\epsilon, z \ge 0$ and an arbitrary sequence of positive reals $(\epsilon_i)_{i\ge 0}$ such that $\epsilon + \sum_{i=0}^{\infty} \epsilon_i = 1$ we have

$$\mathbb{P}ig(\sup_{x\in\mathbb{X}}\phi(x)>zig) \le \mathbb{P}ig(\phi(x_0)>\epsilon zig) + \sum_{i=0}^{\infty}|\chi_{i+1}|\sup_{d(x,y)\leq\delta_i}\mathbb{P}ig(|\phi(x)-\phi(y)|>\epsilon_i zig).$$

Statement Of The Result Sketch Of Proof

Estimates

$$\mathbb{P}\left[||\phi(x_0)-e^{-c}x_0||>\frac{\tilde{\epsilon}R}{2}\right]\leq c_4e^{-\frac{\tilde{\epsilon}^2}{8d^2}R^2}$$

$$\mathbb{P} \left[|||\phi(x) - e^{-c}x|| - ||\phi(y) - e^{-c}y||| > 2^{-j-2}\tilde{\epsilon}R \right]$$

$$\leq \mathbb{P} \left[|||\phi(x) - e^{-c}x|| - ||\phi(y) - e^{-c}y||| > 2^{-j-2}\tilde{\epsilon}R3^{j}|x - y| \right]$$

$$\leq \mathbb{P} \left[|||\phi(x) - \phi(y)||| > 2^{-j-3}\tilde{\epsilon}R3^{j}|x - y| \right]$$

$$\leq \mathbb{P} \left[B_{1}^{*} \geq \frac{\log(2^{-3-j}\tilde{\epsilon}R3^{j}) - \lambda}{\sigma} \right]$$

$$\leq c_{5}(2^{-j-3}\tilde{\epsilon}R3^{j})^{-\frac{\log(2^{-3-j}\tilde{\epsilon}R3^{j}) - 2\lambda}{2\sigma^{2}}}$$

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Open Problems - Directions Of Research

• Pesin Formula for IOUFs

- Meaningful definition of the entropy for IBFs
- Pesin Theory for IBFs

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References

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