

# Stochastic differential equations with Sobolev coefficients

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# Outline

Background

Main result

Sketch of the proof

Let  $A_0, A_1, \dots, A_m$  be Borel measurable vector fields on  $\mathbb{R}^d$ , and  $w_t = (w_t^1, \dots, w_t^m)^*$  an  $m$ -dimensional standard Brownian motion. Consider the Itô SDE

$$dX_t = \sum_{i=1}^m A_i(X_t) dw_t^i + A_0(X_t) dt, \quad X_0 = x. \quad (1)$$

Classical results:

- ▶  $A_0, A_1, \dots, A_m$  are continuous and have linear growth  $\Rightarrow$  SDE (1) has a weak solution;
- ▶ moreover, if the pathwise uniqueness holds for (1), then it has a unique strong solution.

## Two extreme cases

- (i) uniformly non-degenerate case: the matrix

$$a = (a_{ij})_{1 \leq i, j \leq d} \geq C \text{Id},$$

where  $a_{ij} = \sum_{k=1}^m A_k^i A_k^j$  and  $C > 0$ ;

- (ii) completely degenerate case:  $A_1 = \dots = A_m = 0$ , i.e. (1) reduces to an ODE.

## Case (i): uniformly non-degenerate

- ▶ Stroock-Varadhan (1979): if  $A_1, \dots, A_m$  are bounded continuous and  $A_0$  is bounded measurable, then SDE (1) has weak uniqueness;
- ▶ Veretennikov (1979): under stronger conditions, i.e.  $A_1, \dots, A_m$  are bounded Lipschitz, then SDE (1) has pathwise uniqueness;
- ▶ Krylov-Röckner (2005): unity diffusion ( $m = d$  and  $\sigma = (A_1, \dots, A_d) = Id$ ) and  $A_0 \in L^p_{loc}(\mathbb{R}^d)$  with  $p > d + 2 \Rightarrow$  (1) has a unique strong solution;
- ▶ X. Zhang (2005):  $A_1, \dots, A_m$  are continuous and belong to  $W^{1,2(d+1)}_{loc}(\mathbb{R}^d)$ ,  $A_0 \in L^{2(d+1)}_{loc}(\mathbb{R}^d) \Rightarrow$  (1) has a unique strong solution.

The coefficients may depend on time.

## Remark

*Safonov (1999): even in the non-degenerate case, if the diffusion coefficients  $A_1, \dots, A_m$  lose the continuity, there are counterexamples for which the weak uniqueness does not hold.*

Case (ii): completely degenerate

SDE (1) reduces to the ODE

$$dX_t = A_0(X_t) dt, \quad X_0 = x. \quad (2)$$

It is well known that if  $A_0$  is not Lipschitz continuous, then ODE (2) may have no uniqueness of solutions, or it may have no solution at all.

## Sobolev coefficient

- ▶ Cruzeiro (1983) first studied the existence of flow of measurable maps generated by Sobolev vector field  $b$ , provided that its gradient and divergence are exponentially integrable.
- ▶ Further developments: G. Peters (1996), Bogachev-Mayer Wolf (1999) and the book of Ustunel-Zakai (2000).

All these results require the exponential integrability of  $\nabla A_0$ .



- ▶ DiPerna-Lions (1989) obtained similar result without the exponential integrability of  $\nabla A_0$ . They proved this by studying the corresponding transport equation:

$$\frac{\partial u_t}{\partial t} = \langle A_0, \nabla u_t \rangle, \quad u|_{t=0} = u_0.$$

- ▶ Ambrosio (2004) generalized this result to BV vector fields.
- ▶ The extension to infinite dimensional case was done by Ambrosio-Figalli (2009) and Fang-Luo (2010).

Crippa-De Lellis (2008) obtained some new estimates on the perturbation of ODE (2), which allowed them to give a direct construction of the DiPerna-Lions flow.

More precisely, the absence of Lipschitz condition was filled by the following inequality: for  $f \in W_{\text{loc}}^{1,1}(\mathbb{R}^d)$ ,

$$|f(x) - f(y)| \leq C_d |x - y| (M_R |\nabla f|(x) + M_R |\nabla f|(y))$$

holds for  $x, y \in N^c$  and  $|x - y| \leq R$ , where  $M_R |\nabla f|$  is the local maximal function and  $N$  is a negligible set of  $\mathbb{R}^d$ ; the classical moment estimate is replaced by estimating the quantity

$$\int_{B(R)} \log \left( \frac{|X_t(x) - \tilde{X}_t(x)|}{\sigma} + 1 \right) dx,$$

where  $\sigma > 0$  is a small parameter.

X. Zhang (2010) developed this method for SDE and he obtained the stochastic flow of measurable maps generated by (1), mainly under the Sobolev regularity of the coefficients  $A_0, A_1, \dots, A_m$  and the boundedness of their divergences  $\operatorname{div}(A_i)$ .

## Some notations

- $\gamma_d$  the standard Gaussian measure on  $\mathbb{R}^d$
- $\mathbb{D}_1^p(\gamma_d)$  the first order Sobolev space
- $\mathbb{D}_1^\infty(\gamma_d) \cap_{p>1} \mathbb{D}_1^p(\gamma_d)$
- $\delta(A)$  the divergence with respect to  $\gamma_d$

## Theorem (Fang-Luo-Thalmaier)

Assume that

- (i) the coefficients  $A_0, A_1, \dots, A_m$  have linear growth;
- (ii) the diffusion coefficients  $A_1, \dots, A_m$  belong to  $\mathbb{D}_1^\infty(\gamma_d)$  and the drift  $A_0 \in \mathbb{D}_1^q(\gamma_d)$  for some  $q > 1$ ;
- (iii) there is  $\lambda_0 > 0$  such that

$$\int_{\mathbb{R}^d} \exp \left[ \lambda_0 \left( |\delta(A_0)| + \sum_{j=1}^m (|\delta(A_j)|^2 + |\nabla A_j|^2) \right) \right] d\gamma_d < +\infty. \quad (3)$$

Then there is a unique stochastic flow of measurable maps  $X: [0, T] \times \Omega \times \mathbb{R}^d \rightarrow \mathbb{R}^d$ , which solves (1) for almost all initial  $x \in \mathbb{R}^d$  and the push-forward  $(X_t(w, \cdot))_{\#} \gamma_d$  admits a density with respect to  $\gamma_d$ , which is in  $L^1 \log L^1$ .

# Ingredients of the proof

- ▶ a priori estimate of the density of the flow;
- ▶ a priori estimate of the perturbation of SDE (1).

Assume that  $A_0, A_1, \dots, A_m \in C_c^\infty(\mathbb{R}^d, \mathbb{R}^d)$ , then  $X_t$  is a stochastic flow of diffeomorphisms on  $\mathbb{R}^d$ . Kunita (1990) proved that  $(X_t)_\# \gamma_d = K_t \gamma_d$ ,  $(X_t^{-1})_\# \gamma_d = \tilde{K}_t \gamma_d$ , and

$$\tilde{K}_t(x) = \exp \left( - \sum_{i=1}^m \int_0^t \delta(A_i)(X_s(x)) \circ dw_s^i - \int_0^t \delta(\tilde{A}_0)(X_s(x)) ds \right), \quad (4)$$

where  $\tilde{A}_0 = A_0 - \frac{1}{2} \sum_{i=1}^m \langle A_i, \nabla A_i \rangle$ .

We do not have such a simple expression for  $K_t$ , but the following equality holds:

$$K_t(x) = [\tilde{K}_t(X_t^{-1}(x))]^{-1}. \quad (5)$$

# Estimate of the density

Using (4) and (5), we can get the a priori estimate below.

## Theorem (Fang-Luo-Thalmaier)

For any  $p > 1$ ,

$$\|K_t\|_{L^p(\mathbb{P} \times \gamma_d)} \leq \left[ \int_{\mathbb{R}^d} \exp \left( pt \left[ 2|\delta(A_0)| + \sum_{j=1}^m (|A_j|^2 + |\nabla A_j|^2 + 2(p-1)|\delta(A_j)|^2) \right] \right) d\gamma_d \right]^{\frac{p-1}{p(2p-1)}}.$$

This is an extension of Cruzeiro's estimate for the density of the flow associated to ODE (2).



# Regularizing the coefficients

Let  $P_t$  be the Ornstein-Uhlenbeck semigroup on  $\mathbb{R}^d$ :

$$P_t A(x) = \int_{\mathbb{R}^d} A(e^{-t}x + \sqrt{1 - e^{-2t}}y) d\gamma_d(y).$$

If  $A$  has linear growth, then there is  $C > 0$  such that

$$\sup_{0 < t \leq 1} |P_t A(x)| \leq C(1 + |x|), \quad \text{for all } x \in \mathbb{R}^d.$$

Take a sequence of cut-off functions  $\varphi_n \in C_c^\infty(\mathbb{R}^d, [0, 1])$  satisfying

$$\varphi_n|_{B(n)} \equiv 1, \quad \text{supp}(\varphi_n) \subset B(n+2) \quad \text{and} \quad \sup_{n \geq 1} \|\nabla \varphi_n\|_{L^\infty} \leq 1.$$

Now we define  $A_i^n = \varphi_n P_{1/n} A_i$ , then  $A_i^n \in C_c^\infty(\mathbb{R}^d, \mathbb{R}^d)$ ,  
 $0 \leq i \leq m$ .

# Approximating SDEs

Consider the sequence of SDEs

$$dX_t^n = \sum_{i=1}^m A_i^n(X_t^n) dw_t^i + A_0^n(X_t^n) dt, \quad X_0^n = x.$$

Let  $K_t^n$  be the density of  $(X_t^n)_\# \gamma_d$  with respect to  $\gamma_d$ , then by Theorem 3, there is  $T_0$  small enough such that

$$\sup_{0 \leq t \leq T_0} \sup_{n \geq 1} \|K_t^n\|_{L^p(\mathbb{P} \times \gamma_d)} \leq \Lambda_{p, T_0} < +\infty. \quad (6)$$

Using the estimate (6), we can prove

### Theorem

For  $R > 0$  and  $n, k \geq 1$ , define

$$G_R^{n,k}(w) = \left\{ x \in \mathbb{R}^d : \sup_{0 \leq t \leq T_0} (|X_t^n(w, x)| \vee |X_t^k(w, x)|) \leq R \right\}.$$

Then there are  $C_{T_0} > 0$  and  $C_{d,q,R} > 0$  s.t. for any  $\sigma > 0$ ,

$$\begin{aligned} \mathbb{E} \left[ \int_{G_R^{n,k}} \log \left( \frac{\sup_{0 \leq t \leq T_0} |X_t^n - X_t^k|^2}{\sigma^2} + 1 \right) d\gamma_d \right] \\ \leq C_{T_0} \Lambda_{p,T_0} \left( C_{d,q,R} + \frac{\Delta_{n,k}}{\sigma} + \frac{\Delta_{n,k}^2}{\sigma^2} \right), \end{aligned}$$

where  $\Delta_{n,k} = \|A_0^n - A_0^k\|_{L^q} + \left( \sum_{i=1}^m \|A_i^n - A_i^k\|_{L^{2q}}^2 \right)^{1/2}$ .

Taking  $\sigma = \Delta_{n,k}$ , then the above theorem implies the family

$$I_{n,k} := \mathbb{E} \left[ \int_{G_R^{n,k}} \log \left( \frac{\sup_{0 \leq t \leq T_0} |X_t^n - X_t^k|^2}{\Delta_{n,k}^2} + 1 \right) d\gamma_d \right]$$

is bounded. Note that  $\Delta_{n,k} \rightarrow 0$  as  $n, k \rightarrow \infty$ . By the linear growth of the coefficients, we can prove that for any  $\alpha > 0$ ,

$$\lim_{n,k \rightarrow \infty} \mathbb{E} \int_{\mathbb{R}^d} \left( \sup_{0 \leq t \leq T_0} |X_t^n - X_t^k|^\alpha \right) d\gamma_d = 0.$$

That is,  $\{X^n : n \geq 1\}$  is a Cauchy sequence in

$$L^\alpha(\Omega \times \mathbb{R}^d, C([0, T_0], \mathbb{R}^d)),$$

hence there exists  $X : \Omega \times \mathbb{R}^d \rightarrow C([0, T_0], \mathbb{R}^d)$  such that

$$\lim_{n \rightarrow \infty} \mathbb{E} \int_{\mathbb{R}^d} \left( \sup_{0 \leq t \leq T_0} |X_t^n - X_t|^\alpha \right) d\gamma_d = 0. \quad (7)$$

Now we can prove the main result on the small interval  $[0, T_0]$ .  
Using the estimate (6) and letting  $n \rightarrow \infty$  in the equation below:

$$X_t^n = x + \sum_{i=1}^m \int_0^t A_i^n(X_s^n) dw_s^i + \int_0^t A_0^n(X_s^n) ds,$$

we obtain that for a.e.  $x \in \mathbb{R}^d$ , the following equality holds  $\mathbb{P}$ -almost surely:

$$X_t = x + \sum_{i=1}^m \int_0^t A_i(X_s) dw_s^i + \int_0^t A_0(X_s) ds, \quad \text{for all } t \in [0, T_0].$$

The absolute continuity can be proved as follows:  $\forall \xi \in L^\infty(\Omega)$  and  $\psi \in C_c^\infty(\mathbb{R}^d)$ , we have

$$\begin{aligned} \mathbb{E} \int_{\mathbb{R}^d} \xi \cdot \psi(X_t^n(x)) d\gamma_d(x) &= \mathbb{E} \int_{\mathbb{R}^d} \xi \cdot \psi(y) K_t^n(y) d\gamma_d(y) \\ &\quad \begin{array}{ccc} (7) \downarrow & & \downarrow (6) \end{array} \\ \mathbb{E} \int_{\mathbb{R}^d} \xi \cdot \psi(X_t(x)) d\gamma_d(x) &= \mathbb{E} \int_{\mathbb{R}^d} \xi \cdot \psi(y) K_t(y) d\gamma_d(y) \end{aligned}$$

Therefore by the arbitrariness of  $\xi$  and  $\psi$ , we have  $(X_t)_\# \gamma_d = K_t \gamma_d$ .

Using the flow property of  $X_t^n$  and  $X_t$ , together with the following estimate

$$\sup_{0 \leq t \leq T} \sup_{n \geq 1} \mathbb{E} \int_{\mathbb{R}^d} K_t^n |\log K_t^n| d\gamma_d < +\infty,$$

we can prove the general case.

Thanks for your attention!