

Rough Mean Field Equations

PDE and Probability Methods

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Background

- General structure of a McKean Vlasov equation

$$dX_t = b(t, X_t, \mathcal{L}(X_t))dt + \sigma(t, X_t, \mathcal{L}(X_t))dB_t$$

- use the **probabilistic structure** of the noise to define the stochastic integration with respect to $(B_t)_t$

- standard example is **Wiener process**

- Understood as the **asymptotic version of a mean interacting particle system**

$$dX_t^i = b(t, X_t^i, \bar{\mu}_t^N) + \sigma(t, X_t^i, \bar{\mu}_t^N)dB_t^i$$

- i is an integer in $\{1, \dots, N\}$

- $(B_t^i)_t$ are **independent copies** of $(B_t)_t$

- $\bar{\mu}_t^N = \frac{1}{N} \sum_{i=1}^N \delta_{X_t^i}$ is the empirical distribution

- What about **rough signals**?

- \perp copies of a rough signal “à la Lyons” (Cass Lyons, 13)

Motivation

- Theory for general signals like **Gaussian** (non-Brownian) processes
- Have continuity of the **Itô-Lyons map** $input \mapsto output$
 - in the asymptotic regime

$$input = \mathcal{L}((B_t)_t), \quad output = \mathcal{L}((X_t)_t)$$

- in the particle system

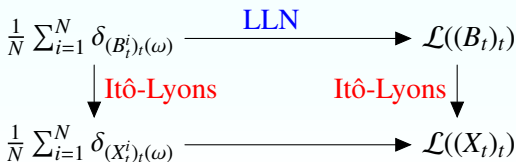
$$input = \frac{1}{N} \sum_{i=1}^N \delta_{(B_t^i)_t(\omega)}, \quad output = \frac{1}{N} \sum_{i=1}^N \delta_{(X_t^i)_t(\omega)}$$

but for a fixed ω !!

- Ask for a **diagram**

- **propagation chaos**

- LDP



Rough signal

- Rough trajectory $W(\omega)$ with same regularity as a Brownian path

$$|W_t(\omega) - W_s(\omega)| \leq C(\omega)|t - s|^\alpha, \quad \alpha \in (1/3, 1/2]$$

- assume that we can define an integral with respect to W and the “iterated integral” of W

$$\mathbb{W}_{s,t}(\omega) = \int_s^t (W_r(\omega) - W_s(\omega)) \otimes dW_r(\omega)$$

- if W is $1d \Rightarrow$ natural candidate is

$$\mathbb{W}_{s,t}(\omega) = \frac{1}{2}(W_t(\omega) - W_s(\omega))^2 \leq C|t - s|^{2\alpha}$$

- if dim greater than 2, “crossed iterated integrals” may not exist \Rightarrow probabilistic structure provides a construction (Stratonovich, Itô...)

- McKV involve infinitely many rough trajectories even if $d = 1!!$

$$\sigma(X_t^i, \bar{\mu}_t^N) dB_t^i = \text{”}\sigma\text{”}(X_t^1, \dots, X_t^N) dB_t^i$$

- X^j involves B^j for $j \neq i!$

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- if dim greater than 2, “crossed iterated integrals” may not exist
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- McKV involve infinitely many rough trajectories even if $d = 1!!$

- \Rightarrow requires a non-trivial rough structure (so far $\sigma = \sigma(x)$,
Cass-Lyons, Deuschel et al...)

Rough integral

• Once $W(\omega) = (W(\omega), \mathbb{W}(\omega))$ is given, one may define an **integral for curves that behave like $W(\omega)$**

◦ **controlled trajectory X**

$$X_t(\omega) - X_s(\omega) = \delta_x X_s(\omega)(W_t(\omega) - W_s(\omega)) + R_{s,t}(\omega)$$

$$|\delta_x X_t(\omega) - \delta_x X_s(\omega)| \leq C^X(\omega)|t - s|^\alpha, \quad |R_{s,t}(\omega)| \leq C^X(\omega)|t - s|^{2\alpha}$$

◦ **rough integral** ((t_i) mesh of $[s, t]$)

$$\int_s^t X_r(\omega) dW_r(\omega) \approx \sum_i X_{t_i}(\omega)(W_{t_{i+1}} - W_{t_i})(\omega) + \sum_i \delta_x X_{t_i}(\omega) \mathbb{W}_{t_i, t_{i+1}}(\omega)$$

• Back to our case \leadsto if $\sigma = \sigma(x)$ **smooth**, define

$$\int_s^t \sigma(X_r(\omega)) dW_r(\omega)$$

by expanding

$$\begin{aligned} \sigma(X_t(\omega)) &= \sigma(X_s(\omega)) + \sigma'(X_s(\omega))(X_t - X_s)(\omega) + R_{s,t}^X(\omega) \\ &= \sigma(X_s(\omega)) + \sigma'(X_s(\omega))\delta_x X_s(\omega)(W_t - W_s)(\omega) + R_{s,t}^X(\omega) \end{aligned}$$

Rough integral

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• Back to our case \rightsquigarrow first step is to define

$$\int_s^t \sigma(X_r(\omega), \mathcal{L}(X_r)) dW_r(\omega)$$

◦ requires to expand $\sigma(X_r(\omega), \mathcal{L}(X_r))$ **including the measure**

Wasserstein derivative

- Lions' approach for differentiating on $\mathcal{P}_2(\mathbb{R})$
- Given $\mathcal{U} : \mathcal{P}_2(\mathbb{R}) \rightarrow \mathbb{R}$
- Lift of \mathcal{U}

$$\hat{\mathcal{U}} : L^2(\Omega, \mathbb{Q}) \ni X \mapsto \mathcal{U}(\mathcal{L}(X))$$

- \mathcal{U} differentiable if $\hat{\mathcal{U}}$ **Fréchet differentiable**
- Derivative of \mathcal{U}
 - Fréchet of $\hat{\mathcal{U}}$

$$D\hat{\mathcal{U}}(X) = \partial_\mu \mathcal{U}(\mu)(X), \quad \partial_\mu \mathcal{U}(\mu) : \mathbb{R} \ni x \mapsto \partial_\mu \mathcal{U}(\mu)(x) \quad \mu = \mathcal{L}(X)$$

- derivative of \mathcal{U} in $\mu \rightsquigarrow \partial_\mu \mathcal{U}(\mu) \in L^2(\mathbb{R}, \mu; \mathbb{R})$
- Finite-dimensional projection

$$\partial_{x_i} \left[\mathcal{U} \left(\frac{1}{N} \sum_{j=1}^N \delta_{x_j} \right) \right] = \frac{1}{N} \partial_\mu \mathcal{U} \left(\frac{1}{N} \sum_{j=1}^N \delta_{x_j} \right) (x_i), \quad x_1, \dots, x_N \in \mathbb{R}$$

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- derivative of \mathcal{U} in $\mu \rightsquigarrow \partial_\mu \mathcal{U}(\mu) \in L^2(\mathbb{R}, \mu; \mathbb{R})$
- X and X' two random variables

$$\begin{aligned} \mathcal{U}(\mathcal{L}(X')) - \mathcal{U}(\mathcal{L}(X)) &= \mathbb{E}^{\mathbb{Q}}[\partial_\mu \mathcal{U}(\mathcal{L}(X))(X(\cdot))(X' - X)(\cdot)] + \dots \\ &= \int \partial_\mu \mathcal{U}(\mathcal{L}(X))(x) \times (x' - x) d\mathcal{L}(X, X')(x, x') + \dots \end{aligned}$$

Extended rough structure

- Expand $\sigma_t = \sigma(X_t(\omega), \mathcal{L}(X_t))$

$$\begin{aligned}\sigma_t - \sigma_s &= \partial_x \sigma(X_s(\omega), \mathcal{L}(X_s)) \partial_x X_s(\omega) (W_t(\omega) - W_s(\omega)) \\ &\quad + \mathbb{E}[\partial_\mu \sigma(X_s(\omega), \mathcal{L}(X_s))(X_s(\cdot)) \partial_x X_s(\cdot) (W_s - W_t)(\cdot)] + R_{s,t}^\sigma(\omega) \\ &= [\delta_x \sigma]_s(\omega) (W_t(\omega) - W_s(\omega)) \\ &\quad + \mathbb{E}[[\delta_\mu \sigma]_s(\omega, \cdot) (W_t - W_s)(\cdot)] + R_{s,t}^\sigma(\omega) \\ &= [\delta_x \sigma]_s(\omega) (W_t(\omega) - W_s(\omega)) \\ &\quad + \int_{\Omega} [\delta_\mu \sigma]_s(\omega, \omega') (W_t - W_s)(\omega') d\mathbb{P}(\omega') + R_{s,t}^\sigma(\omega)\end{aligned}$$

- regularity on the derivatives of σ (need second order derivatives, but Fréchet is too demanding)

Extended rough structure

- Expand $\sigma_t = \sigma(X_t(\omega), \mathcal{L}(X_t))$

$$\begin{aligned}\sigma_t - \sigma_s &= [\delta_x \sigma]_s(\omega)(W_t(\omega) - W_s(\omega)) \\ &\quad + \int [\delta_\mu \sigma]_s(\omega, \omega')(W_t - W_s)(\omega') d\mathbb{P}(\omega') + R_{s,t}^\sigma(\omega)\end{aligned}$$

- By analogy with above, need to define another cross-integral

$$\mathbb{W}_{s,t}^\perp(\omega, \omega') = \int_s^t (W'_r - W'_s)(\omega') dW_r(\omega)$$

- W' independent copy of W on a copy Ω' of Ω
- **pay attention!!!** \mathbb{P} refers to the law in the mean-field interaction
- **asymptotic setting** \rightsquigarrow equip Ω' with $\mathcal{L}((B_t)_t)$ (works for Gaussian processes of dimension 2)
- **particle system** \rightsquigarrow equip Ω' with $\frac{1}{N} \sum_{j=1}^N \delta_{B_t^j(\omega)} \Rightarrow$ need to define

$$\left(\int_s^t (B_r^j - B_s^j) dB_r^i \right)(\omega)$$

Extended rough structure

- Expand $\sigma_t = \sigma(X_t(\omega), \mathcal{L}(X_t))$

$$\begin{aligned}\sigma_t - \sigma_s &= [\delta_x \sigma]_s(\omega)(W_t(\omega) - W_s(\omega)) \\ &\quad + \int [\delta_\mu \sigma]_s(\omega, \omega')(W_t - W_s)(\omega') d\mathbb{P}(\omega') + R_{s,t}^\sigma(\omega)\end{aligned}$$

- By analogy with above, need to define another cross-integral

$$\mathbb{W}_{s,t}^\perp(\omega, \omega') = \int_s^t (W'_r - W_s)(\omega') dW_r(\omega)$$

- W' independent copy of W on a copy Ω' of Ω

- Rough integral should be

$$\begin{aligned}\int_s^t \sigma_r(\omega) d\mathbf{W}_s &= \sum_i \sigma_{t_i}(\omega)(W_{t_{i+1}}(\omega) - W_{t_i}(\omega)) \\ &\quad + \sum_i \delta_x \sigma_{t_i}(\omega) \mathbb{W}_{t_i, t_{i+1}}(\omega) + \sum_i \mathbb{E}[\delta_\mu \sigma_{t_i}(\omega, \cdot) \mathbb{W}_{t_i, t_{i+1}}^\perp(\omega, \cdot)]\end{aligned}$$

Solving the equation

- Search for a **fixed point**

$$\Gamma : \mathbf{X} = (X_\cdot(\omega), \delta_x X_\cdot(\omega), R^X)_\omega \mapsto \left(X_0 + \int_0^\cdot \sigma_r(\omega, \cdot) dW_r, \sigma_\cdot(\omega, \cdot), \dots \right)_\omega$$

◦ suffices to work with T small

- When $\sigma = \sigma(x)$ strategy is to **localize on the variation of $W(\omega)$**

$$\begin{aligned} w(0, T, \omega) &= \frac{1}{\alpha} \text{-var}_{[0, T]}([W(\omega)]) \\ &\quad + \frac{1}{2\alpha} \text{-var}_{[0, T]}(\overline{W}(\omega)) \end{aligned}$$

where

$$\frac{1}{\alpha} \text{-var}_{[0, T]} = \sup_{(t_i)} \sum_i |W_{t_{i+1}} - W_{t_i}|^{\frac{1}{\alpha}}$$

- here \leadsto **no more possible** to do that because of McKV!
- **need a variant of Gronwall** (Cass Litterer Lyons)

Solving the equation

- Search for a **fixed point**

$$\Gamma : \mathbf{X} = (X(\cdot), \delta_x X(\cdot), R^X)_{\omega} \mapsto \left(X_0 + \int_0^{\cdot} \sigma_r(\omega, \cdot) d\mathbf{W}_r, \sigma(\cdot, \cdot), \dots \right)_{\omega}$$

- Find a **norm** $\|\cdot\|_{\omega}$ on $(X(\cdot), \delta_x X(\cdot), R^X(\omega))$

$$\|\Gamma(\mathbf{X})(\omega) - \Gamma(\mathbf{X}')(\omega)\|_{\omega} \leq \rho C^{N(\omega)} \left(\int_{\Omega} \|\mathbf{X}(\omega) - \mathbf{X}'(\omega)\|_{\omega}^p d\mathbb{P}(\omega) \right)^{1/p}$$

- where $\rho < 1$ and $\int_{\Omega} C^{pN(\omega)} d\mathbb{P}(\omega) \rightarrow 1$ as T tends to 0

$$\|\mathbf{X}(\omega)\|_{\omega} = |(X_0, \delta_x X_0)(\omega)| + \sup_{[s,t] \subset [0,T]} \left(\frac{|\delta_x X_t(\omega) - \partial_x X_s(\omega)|}{w(s,t,\omega)^{\alpha}} + \frac{|R_{s,t}(\omega)|}{w(s,t,\omega)^{2\alpha}} \right)$$

- $w(s,t,\omega) \sim \frac{1}{\alpha} \text{-var}_{[s,t]} W(\omega) + \frac{1}{2\alpha} \text{-var}_{[s,t]}(\mathbb{W}(\omega), \mathbb{W}^{\perp}(\omega, \cdot))$

- $N(\omega)$ s.t. $(t_i)_{1 \leq i \leq N(\omega)}$ with $w(t_i, t_{i+1}, \omega) = \epsilon < 1$ and $t_{N(\omega)} \geq T$

Continuity and propagation of chaos

- Kind of statement: If we can **control accordingly the tails** of the variations of W , \mathbb{W} and \mathbb{W}^\perp , then existence and uniqueness
- Continuity of the **law of the output** with respect to the law of $(W, \mathbb{W}, \mathbb{W}^\perp)$.

◦ example: Gaussian processes with Hölder covariance of Hölder exponent $> 2/3$

- Revisit **propagation of chaos**

◦ for N particle system, the law of the triplet takes the form

$$\frac{1}{N^2} \sum_{i,j=1}^N \delta_{B^i(\omega), \mathbb{B}^{i,i}(\omega), \mathbb{B}^{i,j}(\omega)}$$

where

$$\mathbb{B}^{i,j} = \int B^j dB^i$$

◦ converges to the law of $(B, \mathbb{B}, \mathbb{B}^\perp)$