

McKean - Vlasov Equations

- Fix con. $\mathcal{F}, \text{IP}, (\mathbb{F})$. $\xi^i \in \mathcal{Z}_0$, i.i.d. \mathbb{R}^d -valued r.v.'s. (W^i) is indept. \mathbb{F} -m-lim-BM's
- Consider n -interacting particle system $(X_t^{n,k})_{k=1}^n$
 $\dot{X}_t^{n,i} = b(X_t^{n,:}, m_t^n) dt + \sigma(X_t^{n,:}, m_t^n) dW_t^i \quad (*)$
 where $X_0^{n,:} = \xi^i$. $m_t^n = \frac{1}{n} \sum_i^n \delta_{X_t^{n,i}}$
- ⇒ Next consider its n -limit of (X^k)

Assumptions:

- i) $\xi^i \in L^2$.
 - ii) b, σ are bld.
 - iii) $b: \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}^n$. $\sigma: \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}^{n \times n}$
 satisfy Lipschitz condition: $\exists L > 0$.
- $$|b(x, m) - b(x', m')|_{\mathbb{R}^n} + |\sigma(x, m) - \sigma(x', m')|_{\mathbb{R}^{n \times n}} \\ \leq L(|x - x'|_{\mathbb{R}^n} + |m - m'|)$$

(1) McKean-Vlasov Limit:

Lemmas. $(*)$ has unique strong solution. $\forall n$.

Pf: Set $\vec{X}_t = (X_t^{(1)}, \dots, X_t^{(n)}) \in \mathbb{R}^n$

$$L_n(x_1 \dots x_n) = \sum_{k=1}^n \delta_{x_k/n}, \text{ and}$$

$$\mathcal{B}(x) = (b(x_1, L_n(x)), \dots, b(x_n, \dots))^T$$

$$\mathcal{I}(x) = \lim_{\epsilon \rightarrow 0} (\sigma(x_1, L_n(x)) - \sigma(x_n - \epsilon)).$$

$$\Rightarrow \vec{x}_t = \mathcal{B}(\vec{x}_t) \lambda t + \sum \delta_i \lambda \vec{w}_t.$$

Only need to check Lip.

Condition:

$$\text{Note } W_2(L_n(x), L_n(\eta)) \leq \bar{\pi} = \frac{1}{n} \sum_{i=1}^n \delta_{x_i, \eta_i}$$

$$\int_{\mathbb{R}^d} \min_{i=1}^n |u - v|^2 d\bar{\pi} = \frac{1}{n} |\vec{x} - \vec{\eta}|^2.$$

Motivation:

Note as $n \rightarrow \infty$, the interaction of m^* .

become weaker since the contribution
of particle i^{th} has order n^{-1} .

\Rightarrow expert = $M_t^i \rightarrow M^+$. (deter. measure flow)

It looks like: $dY_t^i = b(Y_t^i, M_t) dt + \sigma(Y_t^i, M_t) dW_t$

Def: i) $C^L := C(E_0, T), \mu^L, \|x\|_T := \sup_{t \in [0, T]} |x_t|_{\frac{1}{2}}$

ii) McKean - Vlasov equation is defined by:

$dY_t = b(Y_t, M_t) dt + \sigma(Y_t, M_t) dW_t$ with

$Y_t = \int \cdot M_s \in \mathcal{P}^2(C^L)$.

Remk: i) M is seen as C^L -valued r.v.

There's a canonical map:

$\mathcal{P}^2(C^L) \ni M \mapsto (M_t)_t \in C([0, T], \mathcal{P}^2(\mathbb{R}^d))$

ii) $N = e$ we also have:

$\sup_{t \in [0, T]} W_{1, 2}(M_t, V_t) \leq W_{1, 2}(M, V)$.

(use $\|\cdot\|_T$ on right side)

iii) We refer to M rather (M, Y)
as solution since $M \mapsto Y$ uniquely

Thm. There's a unique strong solution of MV-equation (Def ii)). Besides, set

$$\hat{M} = \overline{\sum} \delta x^{n,k} / n. \text{ We have :}$$

i) $\lim_{n \rightarrow \infty} \overline{E} (W_{C,2}^2 (M^n - M)) = 0$

ii) $(X^{n,1}, \dots X^{n,k}) \xrightarrow{w} (Y_1, \dots Y_k)$. ("copy of

the solutions of MV-equation")

Rmk: i) $(X^{n,1}, \dots X^{n,i})$ is asymptotically i.i.d.

ii) The kinds of limits refer to propagation of chaos :

$$\text{For } \frac{1}{n} \sum \delta x^{n,i} \xrightarrow{w} \mu_0 \in P(X),$$

$$\text{We have } \frac{1}{n} \sum \delta x^{n,i} \xrightarrow{pr} \mu_t \sim Y_t$$

where (Y, μ) is solution with

$\mathcal{L} \sim \mu_0$. i.e. Chaoticity of

initial dist. propagate to $t > 0$.

Pf.: i) \mathcal{U} & $\bar{\mathcal{E}}$:

$$\text{Denote: } \lambda_t^2(m, v) = \inf_{\substack{x \sim m, Y \sim v}} \bar{\mathbb{E}} \left(\|X - Y\|_t^2 \right)$$

$$dY_t^m = b(Y_t^m, m_t)dt + \sigma(Y_t^m, m_t)dW_t.$$

has unique solution Y^m .

Set map: $\phi: \mathbb{P}^{CC^1} \rightarrow \mathbb{P}^{CC^1}$.

$$m \mapsto L(Y^m)$$

Next we find $\phi(m) = m$. So that

(Y^m, m) is solution of MV-equation

By continuity and AM-GM:

$$|Y_t^m - Y_t^v|^2 \leq 2t \int_0^t |b(Y_r^m, m_r) - b(Y_r^v, m_r)| dr$$

$$+ 2 \left| \int_0^t (\sigma(Y_r^m) - \sigma(Y_r^v)) dW_r \right|^2.$$

$$\Rightarrow \bar{\mathbb{E}} \|Y^m - Y^v\|_t^2 \stackrel{\text{Prob.}}{\leq} 2t \bar{\mathbb{E}} \left(\int_0^t |b(\dots) - b(\dots)|^2 dr \right)$$

$$+ 8 \bar{\mathbb{E}} \left(\int_0^t |\sigma(\dots) - \sigma(\dots)|^2 dr \right)$$

$$\stackrel{\text{Lip.}}{\leq} (8 + 2) 2L \left(\bar{\mathbb{E}} \left(\int_0^t \|Y^m - Y^v\|_r^2 dr \right) + W_2^2(m_r, v_r) \right)$$

Apply Brownian's inequality : we have,

$$\mathbb{E} \|Y^m - Y^v\|_t^2 \leq C_{L,T} \int_0^t W_s^2 (\lambda r \cdot M \cdot V_r) dr.$$

$$\leq C \int_0^t \lambda_r^2 (M \cdot V_r) dr$$

$$\Rightarrow \mu_t \in \Phi(M), \Phi(V) = LHS \leq C + \lambda_t^2 (M \cdot V).$$

Choose t small enough and use fixed point theorem. Set $[0, T] = \bigcup [t_k, t_{k+1})$. Σ .

Δt is small enough.

i) Limit:

$$\text{Set } V^n = \frac{1}{n} \sum_i \delta_{Y^i}, M^n = \frac{1}{n} \sum_i \delta_{X^{n,i}}.$$

$$Y^i \in L^2 \Rightarrow V^n \xrightarrow{w^2} M. (n \rightarrow \infty)$$

$$\lambda_b^2 (V^n, M^n) \leq \frac{1}{n} \sum_i \delta_{(X^{n,i}, Y^i)} \cdot n \cdot \sum_i \|X^{n,i} - Y^i\|_t^2. \text{ a.s.}$$

Next, we estimate $\|X^{n,i} - Y^i\|_t^2$:

As above of i) : we have,

$$\mathbb{E} e^{\epsilon \|X^{n,i} - Y^i\|_t^2} \leq C \mathbb{E} e^{\epsilon \int_0^t \lambda_r^2 (M \cdot V_r) dr}$$

$$\Rightarrow \lambda^2 t^2 (\mu^n, V) \leq C \bar{E}^C \int_1^t \lambda r^2 (\mu^n, M) dr$$

$$S_1 = \bar{E}^C \lambda^2 (\mu^n, M) \leq 2 C \bar{E}^C \int_1^t \lambda r^2 (\mu^n, M) dr +$$

triangle

$$2 \bar{E}^C \lambda^2 (\nu^n, M)$$

Apply Gronwall's. Set $t = T$.

$$\bar{E}^C W_2^2 (\mu^n, M) \leq C \bar{E}^C W_2^2 (\nu^n, M) \xrightarrow{n \rightarrow \infty} 0$$

Besides. Note that

$$\bar{E}^C \sup_{1 \leq i \leq k} \| X^{n,i} - Y^i \|_{\mathbb{H}^k} \lesssim \bar{E}^C W_2^2 (\mu^n, M)$$

c.k $\rightarrow 0$ ($n \rightarrow \infty$)

(2) PDEs from MV equation:

The MV-equation admits a similar Kolmogorov forward equation. But the PDE is nonlinear and nonlocal.

Apply Itô's on $\mathcal{C}(Y_t)$. take expectation:

$$\frac{d}{dt} \bar{E}^C \mathcal{P}(Y_t) = \bar{E}^C b(Y_t, M_t) \cdot \nabla \mathcal{P}(Y_t) + \frac{1}{2} \operatorname{Tr} c$$

$$c \sigma^T(Y_t, M_t) \sigma^2 \mathcal{P}(Y_t)$$

Suppose M_t has density $M(t, x)$.

$$\Rightarrow \int_{\mathbb{R}^n} b(x) \partial_t M(t, x) dx \stackrel{\text{integrate by part}}{=} \int_{\mathbb{R}^n} \underbrace{c - \nabla_x \cdot b(x, M(t)) M(t, x)}_{M(t, x) \circ b(x)} + \frac{1}{2} \operatorname{Tr}(\tilde{\sigma}^2 \sigma^T \alpha x \cdot M(t))$$

So we have the PDE w.r.t M :

$$\partial_t M(t, x) = -\nabla_x \cdot b(x, M(t)) M(t, x) + \dots \quad (\star)$$

Rank: i) Nonlinear because of existence

of $b(x, \cdot)$ and $\sigma^T(x, \cdot)$.

ii) called

F-P equation iii) Nonlocal because $b(x, \cdot)$ depends on $M(t, x)_{x \in \mathbb{R}^n}$.

iv) Even if M_t isn't smooth enough. We can consider M as weak solution of the PDE. (Actually it has unique one)

② Alternative for weak solution case:

i) Consider $\varphi \in C_c^\infty$. (L^n defined as before)

$$d\varphi(X_t^{n,i}) \stackrel{\text{def}}{=} \int_{\Omega} m_i^n \varphi(X_t^{n,i}) dt + \nabla \varphi(X_t^{n,i}) \cdot \sigma(X_t^{n,i}, m_i^n) dw_t^i$$

Average

$$\Rightarrow d\langle m_t - \varphi \rangle = \langle m_t \cdot L^{\hat{m}_t} \varphi \rangle dt + dm_t.$$

$$\text{where } \hat{m}_t^n = \frac{1}{n} \sum_{i=1}^n \nabla \varphi(X_t^{n,i}) \cdot \sigma(X_t^{n,i}, m_i^n) dw_t^i$$

$$N_t \langle m^n \rangle_t = \frac{1}{n} \sum_{i=1}^n \int_1^t |D\varphi(X_s^{n,i})|^2 ds \sim O(\frac{t}{n})$$

$$\text{So: } \overline{E} \langle (\hat{m}_t^n)^2 \rangle \stackrel{\text{BDG}}{\leq} \frac{Ct}{n} \rightarrow 0 \quad (n \rightarrow \infty)$$

2) Show $(\hat{m}_t^n)_n$ is tight in $(\mathcal{C}^k, \mathcal{P}^2)$.

$$\Rightarrow \exists m^{\infty} \rightarrow \mathcal{M}. \quad \text{So: } d\langle \hat{m}_t^n, \varphi \rangle = \langle \hat{m}_t^n \cdot L_{\hat{m}_t^n} \varphi \rangle.$$

3) Show every subseq limit are identical.

13) Loss of Markov Property:

It's essentially from the corresponding Fokker-Planck equation (*) is nonlin.

Def: Solutions of SDE is called a

Markov family if:

P_x is law of solution starting
at x . \Rightarrow For the solution with
random initial position $\sim m$. Then

$$\text{law } P_m(A) = \int P_x(A) \lambda_{m(x)}.$$

Rmk: If we solve the SDE with
every deterministic initial state
then we can get the law
of randomized initial one by
integrating it.

But the solutions of MV-equation

isn't Markov family though it has
Markov property (By Lip. conditions)

(4) Common Noise:

Consider the extension of main model:

$$\begin{aligned} dX_t^{n,i} &= b(X_t, \tilde{M}_t) dt + \gamma(X_t, \tilde{M}_t) dW_t \\ &\quad + \sigma_0(X_t, \tilde{M}_t) dB_t. \quad X_0 = \xi^i. \end{aligned}$$

RMK: Note each $X_t^{n,i}$ is also driven by common BM: (B_t) .

SLLN \leftarrow In this case, the asymptotic i.i.d. isn't applicable
Wont hold as before.

B Limit Thms:

consider conditional MV equation:

$$dY_t = b(Y_t, \tilde{M}_t) dt + \sigma(Y_t, \tilde{M}_t) dW_t + \sigma_0(\cdot) dB_t,$$

where $\tilde{M}_t := \mathbb{E}(Y_t | B_t)$ in sense

$$\text{of } \int e^{\lambda \tilde{M}_t} = \mathbb{E}(e^{\lambda Y_t} | B_t), \quad \forall \lambda \in \mathbb{C}^1.$$

and $Y \in \mathcal{F}_0$. B_t is $\mathcal{F}_t - \text{BM}$.

Rmk: i) $\mathcal{L}(Y_t | \mathcal{B}) = \mathcal{L}(Y_t | \mathcal{F}_{\leq t}^{\mathbb{P}})$

ii) For b, σ, σ_0 are uniformly Lip.

$\Rightarrow E$ & κ still hold

Thm. If Rmk ii) holds. Then :

i) There exists the unique strong solution for the cond. MV eqn.

ii) $\exists M \in \mathbb{P}^2$. s.t. $\lim_{n \rightarrow \infty} \mathbb{E}(W_{t, z}^{(n)} | M \cdot m')$
 $= 0$. and $\mathcal{L}(X^{(n)}, X^{(n)}) \xrightarrow{n \rightarrow \infty} M^{\otimes k}$.

Pf: Similar as the usual case

i) Estimate $\|X^{(n)} - Y^{(n)}\|_{\mathbb{E}}$ where

$(Y^{(n)})$ is copy of solution :

$$dY_t^{(n)} = b(Y_t^{(n)}, \tilde{M}_t) dt + \sigma(\dots)$$

ii) easy to see :

$$W_2^2(\mathcal{L}(Z' | \mathcal{B}), \mathcal{L}(Z | \mathcal{B})) \leq$$

$$\mathbb{E}(\|Z' - Z\|^2 | \mathcal{B})$$

PDEs:

$$\text{SDE } \dot{L}_m \varphi(x) = b(x, m) \cdot \nabla \varphi(x) + \frac{1}{2} \text{Tr}(c \sigma \sigma^T \\ + \sigma_0 \sigma_0^T)(x, m) \nabla^2 \varphi(x)$$

Apply Ito's on $\varphi(X_t^{n,i})$ and average.

$$\Rightarrow \langle M_t^n, \varphi \rangle = \langle M_t^n, \dot{L}_{M_t^n} \varphi \rangle dt + dM_t^n \\ + \langle M_t^n, \nabla \varphi(\cdot)^T \sigma(\cdot, M_t^n) \rangle dB_t$$

where $M_t^n = \frac{1}{n} \int_0^t \sum_i \nabla \varphi(X_s^{n,i}) \cdot \sigma(\cdot, X_s^{n,i}) dW_s$.

prove as before: (weak form)

$$\langle M_t^n, \varphi \rangle = -M_t^n \cdot \dot{L}_{M_t^n} \varphi + \langle M_t^n, \nabla \varphi(\cdot)^T \\ \cdot \sigma(\cdot, M_t^n) \rangle dB_t.$$

If $M_t(x) = M(t, x) dx$. then:

$$\partial_t M(t, x) = -\text{Div}_x (b(x, M(t, x)) M(t, x)) + \frac{1}{2} \text{Tr}($$

$$\nabla^2 c \sigma \sigma^T + \sigma_0 \sigma_0^T)(x, M(t, x)) M(t, x))$$

$$- \sum_{j=1}^n \text{Div}_x (M(t, x), \sigma_0^j(t, x)) dB^j_t$$

(5) Long Behavior:

Next, we consider let $t \rightarrow \infty$.

e.g., (Metastability)

$$\lambda X_t^{n,i} = (\alpha(\bar{X}_t^n - x_i) + (x_t - \bar{X}_t^n))\lambda t$$

$$+ \sigma dW_t^i. \quad \bar{X}_t^n = \frac{1}{n} \sum_{i=1}^n X_t^{n,i}.$$

$$\Rightarrow \mathbb{E}_m := \frac{1}{\sum_m} \mathbb{E} \left[\frac{\frac{1}{\sigma^2} (\alpha m x - \frac{\alpha}{2} x^2 + \frac{c}{2} x^2 - \frac{c}{4} x^4)}{4} \right] = V(x)$$

is stationary dist. ($\mathbb{E}_m(L_f) = 0$)

if $m = \int g \lambda \mathbb{E}_m(g)$ holds.

Actually, there $\exists \sigma_c > 0$. if $\sigma \geq \sigma_c$.

the only solution is $m=0$; if $\sigma < \sigma_c$. there also $\exists m_+$. St. $\pm m_+$, we

solutions. (tunneling phenomenon)

Now when σ is large. the noise

term dominates. When σ is small. the drift term dominates and most of particles will be near minimizers ± 1 of $V(x)$.