

On backward propagation of chaos

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Motivation

Forward interacting particles

- Let ξ^1, \dots, ξ^n be i.i.d., \mathcal{F}_0 -measurable random variables and consider

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$$X^{i,n} \rightarrow X^i$$

with

$$X_t^i = \xi^i + \int_0^t b_u(X_u^i, \text{law}(X_u^i)) du + \int_0^t \sigma_u(X_u^i, \text{law}(X_u^i)) dW_u^i,$$

↪ McKean-Vlasov SDE.

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Forward interacting particles

- Propagation of chaos: $\text{law}(X^{1,n}, \dots, X^{k,n}) \rightarrow \text{law}(X^i)^{\otimes k}$
- Concentration: $P\left(\mathcal{W}_2\left(\frac{1}{n} \sum_{i=1}^n \delta_{X_t^{i,n}}, \text{law}(X_t^i)\right) \geq x\right) \leq 2e^{-Cnx^2}$
- Approximation and trend to equilibrium for nonlocal Fokker-Plank equations, e.g. $\partial_t \rho = \nabla \cdot (\nabla \rho + \rho \nabla W * \rho)$

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How about backward particles?

Digression:
Functional inequalities for BSDEs
joint with D. Bartl

Transportation inequality

On a Polish space (E, d) , we consider the **Wasserstein distance**

$$\mathcal{W}_p^p(\mu, \nu) := \inf \left\{ \iint_{E \times E} d^p(x, y) d\pi, \pi_1 = \mu; \pi_2 = \nu \right\}$$

and the **Kullback-Leibler information divergence**

$$H(\nu|\mu) := \begin{cases} E_\nu \left[\log \frac{d\nu}{d\mu} \right] & \text{if } \nu \ll \mu \\ +\infty & \text{else.} \end{cases}$$

Transportation inequality

- The probability measure μ satisfies $T_\rho(C)$ if

$$W_\rho(\mu, \nu) \leq \sqrt{CH(\nu|\mu)} \quad \text{for all } \nu \in \mathcal{P}(E)$$

~ Talagrand (1996)

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\leadsto Talagrand (1996)

Interesting application for us:

- Concentration of measure phenomenon \leadsto Marton, Talagrand, Ledoux

$$\mu^{\otimes n} \left(F - \int F d\mu^{\otimes n} \geq x \right) \leq e^{-cx^2} \iff T_2(C)$$

cf. Gozlan (2009)

Transportation inequality

These inequalities are known to hold for various diffusion, including

- Brownian motion \leadsto Feyel & Üstünel
- Stochastic differential equations \leadsto Djellout-Guillin-Wu; Üstünel; Pal; Saussereau.

Transportation inequality

T_2 for laws of BSDEs

Theorem

Let $F : [0, T] \times \mathcal{C} \times \mathbb{R}^m \times \mathbb{R}^{m \times d} \rightarrow \mathbb{R}^m$ and $G : \mathcal{C} \rightarrow \mathbb{R}^m$ be Lipschitz continuous and let (Y, Z) satisfy

$$dY_t = -F_t(Y_t, Z_t) dt + Z_t dW_t, \quad Y_T = G.$$

Then, *the law of Y satisfies $T_2(C)$ with $C = 2(L_G + TL_F)^2 e^{2TL_F}$.*

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- The constant C is optimal
- If $m = 1$; and $F_t(\omega, y, z) = F_t(z)$ is convex, it is enough to take $\text{id} < F \leq \text{quadratic}$, and $C = 2L_G$.

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- Method:

show that $Y = \varphi(W)$, with φ Lipschitz

\leadsto Ekren-Touzi-Zhang

Backward propagation of chaos

join with M. Laurière

Backward particles

For $\mathbf{x} = (x_1, \dots, x_n) \in (\mathbb{R}^m)^n$, put

$$L^n(\mathbf{x}) := \frac{1}{n} \sum_{i=1}^n \delta_{x_i}.$$

- Let (G^1, \dots, G^n) be i.i.d. \mathcal{F}_T -measurable and consider the system

$$Y_t^{i,n} = G^i + \int_t^T F_u(Y_u^{i,n}, Z_u^{i,n}, L^n(\mathbf{Y}_u)) du - \sum_{k=1}^n \int_t^T Z_u^{i,k,n} dW_u^k$$

where (W^1, \dots, W^n) are n independent Brownian motions.

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where (W^1, \dots, W^n) are n independent Brownian motions.

- Further consider the McKean-Vlasov BSDE

$$Y_t^i = G^i + \int_t^T F_u(Y_u^i, Z_u^i, \mathcal{L}(\mathbf{Y}_u)) du - \int_t^T Z_u^i dW_u^i.$$

Theorem

If $F : [0, T] \times \mathcal{C} \times \mathbb{R}^m \times \mathbb{R}^{m \times d} \times \mathcal{P}_2(\mathbb{R}^m) \rightarrow \mathbb{R}^m$ is Lipschitz continuous (in (y, z, μ)) and there is $k > 2$ such that $E[|G|^k] < \infty$, then

$$\sup_t E \left[\mathcal{W}_2^2(L^n(\mathbf{Y}_t), \mathcal{L}(Y_t)) \right] \leq Cr_{n,m,k}$$

for some rate $r_{n,m,k} \downarrow 0$ as $n \rightarrow \infty$.

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If $k > m + 5$ and $\sup_t E[|Z_t|^{2k}] < \infty$, then

$$E \left[\sup_t \mathcal{W}_2^2(L^n(\mathbf{Y}_t), \mathcal{L}(Y_t)) \right] \leq Cn^{-\frac{2}{m+8}}$$

$r_{n,m,k}$ is explicitly given, and depends on m and k .

Lemma

Lipschitz continuity yields

$$\mathcal{W}_2(L^n(\mathbf{Y}_t), \mathcal{L}(Y_t)) \leq e^{L_F T} \mathcal{W}_2(L^n(\tilde{\mathbf{Y}}_t), \mathcal{L}(Y_t))$$

where $\tilde{\mathbf{Y}} := (\tilde{Y}^1, \dots, \tilde{Y}^n)$ and $(\tilde{Y}^1, \tilde{Z}^1), \dots, (\tilde{Y}^n, \tilde{Z}^n)$ are iid copies of (Y, Z) solving

$$\tilde{Y}_t^i = G^i + \int_t^T F_u(\tilde{Y}_u^i, \tilde{Z}_u^i, \mathcal{L}(Y_u)) du - \int_t^T \tilde{Z}_u^i dW_u^i.$$

- Use results by **Fournier & Guillin (2015)** and **Horowitz & Karandikar (1994)** to conclude.
- See also **Sznitman**.

Theorem

If $E[|G|^k] < \infty$ for some $k > 4$, then

$$P\left(\mathcal{W}_2(L^n(\mathbf{Y}_t), \mathcal{L}(Y_t)) \geq x\right) \leq Cr_{n,x,k}$$

with $r_{n,x,k} \downarrow 0$ (as $n \rightarrow \infty$) exponentially fast.

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If F is Lipschitz in ω , and n large enough

$$P\left(\mathcal{W}_{2,c}(L^n(\mathbf{Y}), \mathcal{L}(Y)) \geq x\right) \leq e^{-Cnx^2}$$

Proposition

If $E[|G|^k] < \infty$ for some $k > 2$, then

$$E \left[\sup_t |Y_t^{i,n} - Y_t^i|^2 + \int_0^T |Z_t^{i,n} - Z_t^i|^2 dt \right] \leq Cr_{n,m,k}.$$

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If the interaction is "*linear*", then $r_{n,m,k} = n^{-1}$, *optimal, dimension-free rate!* (see *Buckdahn, Djehiche, Li & Peng (2009)*)

~>

- Propagation of chaos: $\mathcal{W}_{2,c}(\theta^{k,n}, \mathcal{L}(Y)^{\otimes k}) \leq kCr_{n,m,k}$ with $\theta^{n,k} := \text{law}(Y^{1,n}, \dots, Y^{k,n})$.
- If the interaction is linear, then $r_{n,m,k} = n^{-1}$.

FBSDE: infinite dimensional PDEs and large population games

joint with M. Laurière

Forward-backward particles

theoretical results

Now consider the system of FBSDEs

$$\begin{cases} dX_t^{i,n} = B_t(X_t^{i,n}, Y_t^{i,n}, L^n(\mathbf{X}_t, \mathbf{Y}_t)) dt + \sigma dW_t^i \\ dY_t^{i,n} = -F_t(X_t^{i,n}, Y_t^{i,n}, Z_t^{i,n}, L^n(\mathbf{X}_t, \mathbf{Y}_t)) dt + \sum_{k=1}^n Z_t^{i,k,n} dW_t^k \\ X_0^{i,n} = \xi^i \quad Y_T^{i,n} = G(X_T^{i,n}, L^n(X_T^{i,n})) \end{cases}$$

with B , F , and G Lipschitz continuous functions of linear growth,

$$|F_t(x, y, z, \mu)| \leq C(1 + |y| + |z| + M_2(\mu))$$

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and the McKean-Vlasov FBSDE

$$\begin{cases} dX_t = B_t(X_t, Y_t, \mathcal{L}(X_t, Y_t)) dt + \sigma dW_t \\ Y_t = -F_t(X_t, Y_t, Z_t, \mathcal{L}(X_t, Y_t)) dt + Z_t dW_t \\ X_t = \xi, \quad Y_T = G(X_T, \mathcal{L}(X_T)) \end{cases}$$

cf. **Carmona & Delarue**

Forward-backward particles

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If there is $k > 2$ such that $E[|\xi|^k] < \infty$, then

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with $r_{n,m,l,k} \downarrow 0$ as $n \rightarrow \infty$.

Forward-backward particles

Approximation of the master equation

Take $m = 1$. Let us consider the PDE

$$\begin{cases} \partial_t V(t, x, \mu) + B(x, V(t, x, \mu), \nu) \partial_x V(t, x, \mu) + \frac{1}{2} \text{tr}(\partial_{xx} V(t, x, \mu) \sigma \sigma') \\ \quad + F(x, V(t, x, \mu), \partial_x V(t, x, \mu) \sigma, \nu) \\ \quad + \int_{\mathbb{R}^d} \partial_\mu V(t, x, \mu)(y) \cdot B(y, V(t, x, \mu), \nu) d\mu(y) \\ \quad + \int_{\mathbb{R}^d} \frac{1}{2} \text{tr}(\partial_y \partial_\mu V(t, x, \mu)(y) \sigma \sigma') d\mu(y) = 0 \\ V(T, x, \mu) = G(x, \mu) \end{cases}$$

with $(t, x, \mu) \in [0, T] \times \mathbb{R}^d \times \mathcal{P}_2(\mathbb{R}^d)$. The derivative

$$\partial_\mu V(t, x, \mu)(y)$$

denotes the so-called L-derivative and ν is the law of $(\xi, U(t, \xi, \mu))$ when $\mathcal{L}(\xi) = \mu$.

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\leadsto Mou & Zhang (2019), Wu & Zhang (2019), Cardaliaguet et. al (2019), Chassagneux et. al (2015), Gangbo & Świech (2015)

Forward-backward particles

Approximation of the master equation

$$\left\{ \begin{array}{l} \partial_t v^{i,n}(t, \mathbf{x}) + B(x_i, v^{i,n}(t, \mathbf{x}), \frac{1}{n} \sum_{j=1}^n \delta_{(x_j, v^{j,n}(t, \mathbf{x}))}) \partial_{x_i} v^{i,n}(t, \mathbf{x}) \\ \quad + \frac{1}{2} \text{tr} (\partial_{x_i x_i} v^{i,n}(t, \mathbf{x}) \sigma \sigma') \\ \quad + F \left(x_i, v^{i,n}(t, \mathbf{x}), \partial_{x_i} v^{i,n}(t, \mathbf{x}) \sigma(x_i), \frac{1}{n} \sum_{j=1}^n \delta_{(x_j, v^{j,n}(t, \mathbf{x}))} \right) = 0 \\ v^{i,n}(T, \mathbf{x}) = G(x_i, L^n(\mathbf{x})), \quad \mathbf{x} = (x_1, \dots, x_n) \in (\mathbb{R}^d)^n \\ i = 1, \dots, n. \end{array} \right.$$

with $(t, \mathbf{x}) \in [0, T] \times (\mathbb{R}^d)^n$

Forward-backward particles

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with $(t, \mathbf{x}) \in [0, T] \times (\mathbb{R}^d)^n$

Theorem

If the master equation has a unique solution V such that $V(t, \xi, \mathcal{L}(\xi)) = Y^{t, \xi}$ for all \mathcal{F}_t -measurable $\xi \in L^2(P)$, then

$$E [|v^{1,n}(t, \xi_1, \dots, \xi_n) - V(t, \xi_1, \mu)|^2] \leq C r_{n,l,k,2}$$

for all t and all i.i.d. ξ^i with law μ .

- Chassagneux; Crisan & Delarue (2015) recently gave conditions under which

$$V(t, \xi, \mathcal{L}(\xi)) = Y_t^{t, \xi}$$

- by classical FBSDE theory (e.g. Ma; Protter & Young)

$$v^{1, n}(t, \xi_1, \dots, \xi_n) = Y_t^{t, \xi^1, \dots, \xi^n}$$

- thus, suffices to show $Y_t^{t, \xi^1, \dots, \xi^n} \rightarrow Y_t^{t, \xi}$.

PDE-based approach to a similar result \leadsto Cardaliaguet, Delarue, Lasry & Lions (2019)

Forward-backward particles

Extended mean-field games

Consider $dX_t^i = \alpha_t^i + \frac{1}{n} \sum_{j=1}^n \alpha_t^j dt + \sigma dW_t^i$ and

$$J^i(\alpha) := E \left[|X_T^i|^2 + \int_0^T \frac{1}{2} (\alpha_t^i)^2 + \left(\frac{1}{n} \sum_{j=1}^n \alpha_t^j \right)^2 dt \right] \rightarrow \min$$

Pontryagin for N-player game: $\hat{\alpha}^{i,n}$ n-Nash $\rightsquigarrow \hat{\alpha}^{i,n} = -Y^{ii} + R_n(Y^{ij})_{ij}$
with

$$\begin{cases} dX_t^i = \hat{\alpha}_t^{i,n} + \frac{1}{n} \sum_{j=1}^n \hat{\alpha}_t^{j,n} dt + \sigma dW_t^i \\ dY_t^{ij} = \sum_{k=1}^n Z_t^{ijk} dW_t^k, \quad Y_T^{ij} = 2\delta_{ij} X_T^i \end{cases}$$

and $R_n(Y^{ij})_{ij=1,\dots,n} \rightarrow 0$

Forward-backward particles

Extended mean-field games

Let $dX_t = (\alpha_t + E[\alpha_t]) dt + \sigma dW_t$ and a flow of measures μ . Find $\hat{\alpha}$ s.t.

$$\inf_{\alpha} E \left[|X_T^\alpha|^2 + \int_0^T \frac{1}{2} \alpha_t^2 + \left(\int x d\mu_t(x) \right)^2 dt \right] \quad \text{and } \mu_t = \text{law}(\hat{\alpha}_t^\mu).$$

Pontryagin for extended MFG: $\hat{\alpha} = -Y$ with

$$\begin{cases} dX_t = -(Y_t + E[Y_t]) dt + \sigma dW_t \\ dY_t = Z_t dW_t, \quad Y_T = 2X_T \end{cases}$$

Forward-backward particles

Extended mean-field games

Let $dX_t = (\alpha_t + E[\alpha_t]) dt + \sigma dW_t$ and a flow of measures μ . Find $\hat{\alpha}$ s.t.

$$\inf_{\alpha} E \left[|X_T^\alpha|^2 + \int_0^T \frac{1}{2} \alpha_t^2 + \left(\int x d\mu_t(x) \right)^2 dt \right] \quad \text{and} \quad \mu_t = \text{law}(\hat{\alpha}_t^\mu).$$

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By the above results:

$$\hat{\alpha}^{j,n} \rightarrow \hat{\alpha}$$

See also works by [Cardaliaguet, Delarue, Lasry & Lions \(2009\)](#), [Lacker \(2016, 2017, 2018\)](#), [Fischer \(2017\)](#), [Nutz, San Martin & Tan \(2018\)](#)

Summary

- Functional inequalities for BSDE
- Backward propagation of chaos
 - explicit convergence rates
 - concentration inequalities
- Forward backward "particles"
 - approximation of master equation
 - convergence to extended MFG

Thank You!