

# Backward propagation of chaos and large population games asymptotics

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Bachelier Finance Society One World seminar series  
Sept. 2020

## Motivation

### Forward interacting particles

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

$$X_0^{i,n} = \xi^i$$

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$$X_t^{i,n} = \xi^i + \int_0^t b_u \left( X_u^{i,n}, \frac{1}{n} \sum_{j=1}^n \delta_{X_u^{j,n}} \right) du + \int_0^t \sigma_u \left( X_u^{i,n}, \frac{1}{n} \sum_{j=1}^n \delta_{X_u^{j,n}} \right) dW_u^i$$

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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.

## Motivation

### 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.  $\partial_t \rho = \nabla \cdot (\nabla \rho + \rho \nabla V * \rho)$

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How about backward particles?

# Backward propagation of chaos



## 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.  $\mathcal{F}_T^n$ -measurable and consider the system

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

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- Further consider the McKean-Vlasov BSDE

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

## 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  $(y, z, \mu)$ ) 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  $n \rightarrow \infty$ .

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

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

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

$r_{n,m,k}$  and  $\tilde{r}_{n,x,k}$  are 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

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

- By [Bartl & T. \(2020\)](#),  $\mathcal{L}(Y)$  satisfies dimension-free Talagrand inequality.
- Use results by [Fournier & Guillin \(2015\)](#) and [Horowitz & Karandikar \(1994\)](#) to conclude.
- See also [Sznitman](#).

## Backward particles

### Examples of applications

- Crowd motion with tagged pedestrians  $\rightsquigarrow$  Aurell & Djehiche (2018, 2019)

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- Relaxed controls

$$\alpha \rightsquigarrow q \in \mathcal{P}([0, T] \times \mathbb{A})$$

$\rightsquigarrow$  Lacker (2016, 2018); Fischer (2017); Djete (2020)

- PDE characterizations (PDE on the Wasserstein space)

$$\alpha_t \equiv \alpha(t, X_t)$$

$\rightsquigarrow$  Delarue, Lacker, Raman (2018); Cardaliaguet, Delarue, Lasry, Lions (2019).

# Large population games asymptotics: The Markovian case



## Motivating example

### Trading with impact

- Consider a large number of investors trading a stock  $S$ . Their inventories is

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- Assuming trading has a **permanent impact** on the stock price  $S$ , we have (c.f. **Almgren-Chriss 00'**)

$$dS_t = \frac{1}{n} \sum_{j=1}^n \alpha_t^j dt + \sigma^o dW_t^o$$

- Assuming (quadratic) **transaction costs**  $\frac{1}{2}|\alpha_t^i|^2$ , the wealth of trader  $i$  is

$$dV_t^i = \left( -\frac{1}{2}|\alpha_t^i|^2 + X_t^i \frac{1}{n} \sum_{j=1}^n \alpha_t^j \right) dt + X_t^i \sigma dW_t^o + S_t \sigma dW_t^i.$$

## Motivating example

### Trading with impact

- If the traders face the terminal liquidation constraint  $g(X_T^i)$  and are risk neutral, then they have to solve

$$J(\underline{\alpha}) = E \left[ V_T^i - g(X_T^i) \right] \rightarrow \max$$

That is,

$$\begin{cases} \inf_{\alpha^i} E \left[ \int_0^T \frac{1}{2} |\alpha_t^i|^2 - X_t^i \frac{1}{n} \sum_{j=1}^n \alpha_t^j dt + g(X_T^i) \right] \\ dX_t^i = \alpha_t^i dt + \sigma dW_t^i \end{cases}$$

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- As usual, one is interested in computing a Nash equilibrium.

## Trading with impact

### The mean field game

- The mean field game analogue of this game takes the following form: Fix a flow of measure  $(\nu_t)_{t \in [0, T]}$  and find  $\hat{\alpha}^\nu$  solving

$$\begin{cases} \inf_{\alpha} E \left[ \int_0^T \frac{1}{2} |\alpha_t|^2 - X_t \int x \nu_t(dx) dt + g(X_T^\alpha) \right] \\ dX_t^\alpha = \alpha_t dt + \sigma dW_t^i. \end{cases}$$

- The mean field equilibrium is  $\hat{\alpha} = \hat{\alpha}^{\hat{\nu}}$  such that  $\text{law}(\hat{\alpha}_t^{\hat{\nu}}) = \hat{\nu}_t$ .
- It is the optimal strategy for an infinitesimal representative player

↪ Lasry, Lions 2006; Huang, Malhamé, and Caines 2006.  
Gomes, Patrizi, and Voskanyan 2014

## Theorem

If  $\hat{\alpha}^n = (\hat{\alpha}^{i,n}, \dots, \hat{\alpha}^{i,n})$  is a Nash equilibrium, then it satisfies  $\alpha_t^{i,n} = -\frac{1}{1+\frac{1}{n}X_t^{i,n}} Y_t^{ii,n}$  with  $(X^{i,n}, Y^{ij,n}, Z^{ijk,n})$  solving the FBSDE

$$\begin{cases} dX_t^{i,n} = -\frac{1}{1+\frac{1}{n}X_t^{i,n}} Y_t^{ii,n} dt + \sigma dW_t^i \\ dY_t^{ij,n} = -\delta_{ij} \frac{1}{n} \sum_{k=1}^n \hat{\alpha}_t^{k,n} dt - \sum_{k=1}^n Z_t^{ijk,n} dW_t^k \\ Y_T^{ij,n} = \delta_{ij} g'(X_T^{i,n}) \end{cases}$$

## Theorem

Assume that  $E[|X_0^i|^k] < \infty$  for  $k > 2$  and  $g$  Lipschitz continuous. If  $\hat{\alpha}^n = (\hat{\alpha}^{1,n}, \dots, \hat{\alpha}^{n,n})$  is a Nash equilibrium, then there is  $\hat{\alpha}^j$  such that for all  $i = 1, \dots, n$  it holds

$$E[|\hat{\alpha}_t^{i,n} - \hat{\alpha}_t^j|^2] \leq C r_{n,m,d,k}$$

and  $\hat{\alpha}^j$  is a mean field equilibrium. In addition, if  $X_0^i$  is constant then for  $n$  large enough,

$$P(\mathcal{W}_2(L^n(\hat{\alpha}_t^n), \mathcal{L}(\hat{\alpha}_t^j)) \geq a) \leq \frac{C}{a^2 n^2} + C e^{-a^2 n^2}.$$

# Large population games asymptotics: The non-Markovian case



## Motivating example

### Weak formulation

- If the traders face the terminal liquidation constraint  $g(X^i, L^n(\mathbf{X}))$  and are risk neutral, then they have to solve

$$\begin{cases} \inf_{\alpha^i} E \left[ \int_0^T \frac{1}{2} |\alpha_t^i|^2 - X_t^i \frac{1}{N} \sum_{j=1}^N \alpha_t^j dt + g(X^i, L^n(\mathbf{X})) \right] \\ dX_t^i = \alpha_t^i dt + \sigma dW_t^i \end{cases}$$

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- We rather consider the weak formulation of the problem:

$$\begin{cases} \inf_{\alpha^i} E^{P^\alpha} \left[ \int_0^T \frac{1}{2} |\alpha_t^i|^2 - X_t^i \frac{1}{N} \sum_{j=1}^N \alpha_t^j dt + g(X^i, L^n(\mathbf{X})) \right] \\ dX_t^i = \alpha_t^i dt + \sigma dW_t^i \\ dP^\alpha = \mathcal{E} \left( \sum_{i=1}^n \int \alpha_t^i dW_t^i \right) dP \end{cases}$$

## Motivating example

### Mean field game in weak formulation

- The mean field game analogue of this game takes the following form: Fix a flow of measure  $\xi_t = (\mu_t, \nu_t)$  and find  $\hat{\alpha}^\xi$  solving

$$\begin{cases} \inf_{\alpha} E^{P^\alpha} \left[ \int_0^T \frac{1}{2} |\alpha_t|^2 - X_t \int x \nu_t(dx) + g(X, \mu) \right] \\ dX_t^i = \alpha_t dt + \sigma dW_t^i \\ dP^\alpha = \mathcal{E}(\int_0^T \alpha dW_t) dP. \end{cases}$$

- The mean field equilibrium is  $\hat{\alpha} = \hat{\alpha}^\xi$  such that  $P^{\hat{\alpha}} \circ (X_t, \hat{\alpha}_t^\xi)^{-1} = \xi_t$ .

↪ Carmona & Lacker 2015

## Trading with impact

### A general characterization

#### Theorem

If  $\hat{\alpha}^n$  is a Nash equilibrium, then it satisfies  $\hat{\alpha}_t^{i,n} = \frac{1}{1+\frac{1}{n}X_t^i} Z_t^{ii,n}$  where  $(Y^{i,n}, Z^{ij,n})$  solves

$$\begin{cases} dY_t^{i,n} = -\left\{ f(X_t^i/n, Z^{ii,n}) + X_t \frac{1}{n} \sum_{j=1}^n Z_t^{jj,n} \right\} dt - \sum_{j=1}^n Z^{ij,n} dW_t^j \\ Y_T^{i,n} = g(X^i, L^n(\mathbf{X})) \end{cases}$$

## Main convergence result

### Theorem

Assume that  $g$  is bounded and Lipschitz continuous. If there is a Nash equilibrium  $\hat{\alpha}^N$ , then there is  $\hat{\alpha}^i$  such that for all  $i$  it holds

$$E^{P^{\hat{\alpha}^i}} \left[ \int_0^T |\hat{\alpha}_t^{i,n} - \hat{\alpha}_t^i|^2 dt \right] \leq CE^{P^{\hat{\alpha}^i}} [\mathcal{W}_{2,L^2}(L^n(\mathbf{X}), \mathcal{L}_{\hat{\alpha}}(X))] + Cr_n.$$

and  $\hat{\alpha}^i$  is a mean-field equilibrium. In particular, up to a subsequence,  $(\hat{\alpha}_t^{i,n})$  converges to  $\hat{\alpha}_t^i P^{\hat{\alpha}^i} \otimes dt$ -a.s.

# Convergence to mean field game limit

## Concluding remarks

- Some advantages of our approach
  - Convergence rates and concentration inequalities
  - Non uniqueness assumed
  - The Markovian case has interesting PDE interpretations.
  - "Explicit" representations of equilibria

# Convergence to mean field game limit

## Concluding remarks

- Some advantages of our approach
  - Convergence rates and concentration inequalities
  - Non uniqueness assumed
  - The Markovian case has interesting PDE interpretations.
  - "Explicit" representations of equilibria
- Some issues
  - Explicit representations needed
  - Closed-loop controls
  - Symmetry essential

## Summary

- Backward propagation of chaos
  - explicit convergence rates
  - concentration inequalities
- Forward backward "particles"
  - convergence to extended MFG
- Interaction through control process
  - convergence to non-Markovian extended MFG



Thank You!