

# A probabilistic weak formulation of mean field games and applications

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## Price impact model

- ▶  $n$  brokers trade in the same asset and maximize wealth.
- ▶ Brokers face identical limit order books.
- ▶ Broker  $i$  controls his rate of trade  $\alpha_t^i$ .
- ▶ The asset price is a martingale plus a drift given by price impact.  
(Almgren-Chriss '01, Carlin et al '09)

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$$dX_t^i = \alpha_t^i dt + \sigma dW_t^i$$

- ▶ Asset price:

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- ▶ Broker  $i$ 's wealth is  $V_t^i = V_0^i + X_t^i S_t + K_t^i$ , or

$$dV_t^i = \left( \frac{\gamma}{n} \sum_{j=1}^n c'(\alpha_t^j) X_t^i - c(\alpha_t^i) \right) dt + \sigma S_t dW_t^i + \sigma_0 X_t^i dB_t$$

## Price impact model - optimization

Broker  $i$  maximizes expected wealth  $\mathbb{E}[V_T^i]$ :

$$\sup_{\alpha^i} \mathbb{E} \int_0^T \left( \frac{\gamma}{n} \sum_{j=1}^n c'(\alpha_t^j) X_t^i - c(\alpha_t^i) \right) dt,$$
$$\text{s.t. } dX_t^i = \alpha_t^i dt + \sigma dW_t^i$$

Are there Nash equilibria?

## Price impact model - general objectives

Additional objective functions  $G$  and  $F$  allow for time- $T$  liquidation demands, tracking requirements, etc.

$$\sup_{\alpha^i} \mathbb{E} \left[ G(X_T^i) + \int_0^T \left( \frac{\gamma}{n} \sum_{j=1}^n c'(\alpha_t^j) X_t^i - c(\alpha_t^i) - F(t, X_t^i) \right) dt \right],$$

s.t.  $dX_t^i = \alpha_t^i dt + \sigma dW_t^i$

The optimization problems are coupled through the **empirical distribution of the controls**. Limit  $n \rightarrow \infty$ ?

# Mean field stochastic differential games (MFG)

Player  $i$ 's state process and objective:

$$\sup_{\alpha^i} \mathbb{E} \left[ \int_0^T f(t, X_t^i, \bar{\mu}_t^n, \bar{\nu}_t^n, \alpha_t^i) dt + g(X_T^i, \bar{\mu}_T^n) \right],$$

$$\text{s.t. } dX_t^i = b(t, X_t^i, \bar{\mu}_t^n, \alpha_t^i) dt + \sigma(t, X_t^i) dW_t^i,$$

$$\bar{\mu}_t^n := \frac{1}{n} \sum_{j=1}^n \delta_{X_t^j}, \quad \bar{\nu}_t^n := \frac{1}{n} \sum_{j=1}^n \delta_{\alpha_t^j}$$

We study the **mean field limit**, as proposed (with no  $\bar{\nu}_t^n$  dependence) by Lasry & Lions and independently by Caines, Huang, & Malhamé in '06.



Limit  $n \rightarrow \infty$ 

1. Fix measure flows  $t \mapsto (\mu_t, \nu_t)$
2. Solve a standard optimal control problem

$$\sup_{\alpha} \mathbb{E} \left[ \int_0^T f(t, X_t, \mu_t, \nu_t, \alpha_t) dt + g(X_T, \mu_T) \right], \text{ s.t.}$$

$$dX_t = b(t, X_t, \mu_t, \alpha_t) dt + \sigma(t, X_t) dW_t,$$

3. Let  $\mu'_t$  denote the law of the optimally controlled state process at time  $t$  and  $\nu'_t$  the law of the optimal control at time  $t$ .
4. Find a fixed point  $(\mu'_t, \nu'_t) = (\mu_t, \nu_t)$ .

## Existence and uniqueness theory

Any approach to stochastic optimal control may be applied in step 2.

1. PDEs - Lasry & Lions, Caines et al.
2. Stochastic maximum principle - Carmona & Delarue, Bensoussan et al.
3. **Weak formulation**

# Assumptions

- ▶ compact convex control space  $A$
- ▶ admissible controls  $\mathbb{A} =$  progressively measurable  $A$ -valued processes
- ▶  $b, f, g$  jointly measurable and continuous in  $(\mu, \nu, a)$ , at points where  $\mu \sim$  Lebesgue
- ▶  $\sigma$  is measurable and bounded away from zero, and there exists a unique strong solution to  $dX_t = \sigma(t, X_t)dW_t$
- ▶  $b, \sigma$  bounded
- ▶ some growth assumptions for  $f$  and  $g$

## Setup

On  $(\Omega, \mathcal{F}, P)$ , solve  $dX_t = \sigma(t, X_t)dW_t$ . For  $(\mu, \alpha)$  fixed,

$$\frac{dP^{\mu, \alpha}}{dP} := \exp \left[ \int_0^T \sigma^{-1} b(t, X_t, \mu_t, \alpha_t) dW_t - \frac{1}{2} \int_0^T |\sigma^{-1} b(t, X_t, \mu_t, \alpha_t)|^2 dt \right]$$

Under  $P^{\mu, \alpha}$ ,  $X$  is a weak solution of the state equation,

$$dX_t = b(t, X_t, \mu_t, \alpha_t) dt + \sigma(t, X_t) dW_t^{\mu, \alpha}$$

Let  $\Phi(\mu, \alpha)_t = (P^{\mu, \alpha} \circ X_t^{-1}, P^{\mu, \alpha} \circ \alpha_t^{-1})$ .

## Setup

For  $(\mu, \nu)$  fixed, define the value function

$$V_t^{\mu, \nu} := \operatorname{ess\,sup}_{\alpha \in \mathbb{A}} \mathbb{E}^{P^{\mu, \alpha}} \left[ \int_t^T f(t, X_t, \mu_t, \nu_t, \alpha_t) dt + g(X_T, \mu_T) \middle| \mathcal{F}_t \right]$$

Hamiltonian and maximized Hamiltonian:

$$h(t, x, \mu, \nu, z, a) := f(t, x, \mu, \nu, a) + z \cdot \sigma^{-1} b(t, x, \mu, a),$$

$$H(t, x, \mu, \nu, z) := \sup_{a \in A} h(t, x, \mu, \nu, z, a)$$

Can show  $V_t^{\mu, \nu}$  solves the BSDE

$$\begin{cases} dV_t^{\mu, \nu} &= -H(t, X_t, \mu_t, \nu_t, Z_t^{\mu, \nu}) dt + Z_t^{\mu, \nu} dW_t, \\ V_T^{\mu, \nu} &= g(X_T, \mu_T) \end{cases}$$

By comparison principle, the set of optimal controls is exactly

$$\mathbb{A}(\mu, \nu) := \{ \alpha \in \mathbb{A} : \alpha_t \in A(t, X_t, \mu_t, \nu_t, Z_t^{\mu, \nu}) \, dt \times dP - a.e. \}$$

$$A(t, x, \mu, \nu, z) := \arg \max_{a \in A} h(t, x, \mu, \nu, z, a)$$

## Existence and uniqueness

A MFG solution is a fixed point

$$(\mu, \nu) \in \Phi(\mu, \mathbb{A}(\mu, \nu)) = \{\Phi(\mu, \alpha) : \alpha \in \mathbb{A}(\mu, \nu)\}$$

### Theorem

- ▶ Assume the Hamiltonian  $h$  is concave in  $a$  and  $f = f_1(t, x, \mu, a) + f_2(t, x, \mu, \nu)$ . Then there exists a fixed point.
- ▶ Assume the Hamiltonian  $h$  is strictly concave in  $a$ ,  $f = f_1(t, \mu, \nu) + f_2(t, x, a)$ ,  $g = g(x)$ , and  $b = b(t, x, a)$ . Then the fixed point is unique.

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### Proof.

1. Kakutani's fixed point theorem.
2. Translate Lasry & Lions' proof into probabilistic language. □

# Approximate equilibria for the finite-player game

## Theorem

*If  $\alpha = \alpha(t, X_.)$  is an optimal feedback control for the MFG problem, then the strategy profiles  $\alpha(t, X^i)$  form an approximate Nash equilibrium for the finite-player game - for some  $\epsilon_n \downarrow 0$ , for each  $n$ , no player in the  $n$ -player game can increase his expected reward by more than  $\epsilon_n$  by unilaterally changing strategy.*



## Price impact model, revisited

Price impact model corresponds to:

- ▶  $b(t, x, \mu, \alpha) = \alpha$
- ▶  $\sigma$  constant
- ▶  $g(x, \mu) = G(x)$
- ▶  $f(t, x, \mu, \nu, \alpha) = \gamma x \int c' d\nu - c(\alpha) - F(t, x).$

### Theorem

*For a bounded order book, with  $c'$  continuous, the mean field price impact model has a solution. Moreover, the errors  $\epsilon_n$  are  $O(1/\sqrt{n})$ .*

## Other types of interactions

- ▶ Rank
  - ▶  $f(t, x, \mu, \nu, a)$  contains a term  $F(\mu(-\infty, x])$
  - ▶ Guéant, Lasry, Lions - oil production model

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  - ▶ In dimension 1,  $f(t, x, \mu, \nu, a)$  involves quantile function  $F_{\mu}^{-1}(\cdot) = \inf\{y \in \mathbb{R}; \mu(-\infty, y] \geq \cdot\}$
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  - ▶ Analogous “nearest-neighbor” functions in multiple dimensions are useful for flocking models
- ▶ Sub-populations or Types
  - ▶  $f(t, x, \mu, \nu, a) = \hat{f}(t, x, F_1(\mu), \dots, F_m(\mu), \nu, a)$ , where  $F_i(\mu)(B) := 1_{\{\mu(B_i) > 0\}} \mu(B \cap B_i) / \mu(B_i)$  and  $B_i \subset \mathbb{R}^d$  have positive Lebesgue measure
  - ▶ e.g. income brackets