

Mean Field Games with incomplete information

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1 General Model

2 Product Differentiation

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- 2 Product Differentiation

A Mean Field Game is a continuous-time dynamic game with a continuum of non-atomic players, where :

- at each time instant t , each player in position $X_t \in \mathbb{R}^n$ chooses its velocity $\alpha_t \in \mathbb{R}^n$ and its new position is $X_{t+dt} = X_t + \alpha_t dt + \sigma dB_t$,
- at each time instant t , each player suffers a stage cost that depends on position X_t , velocity α_t and **distribution of players** at time t .
- The total cost is the expected average cost between time 0 and $T > 0$ (alternatives : discounted, long-run average payoff...).

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- Bank run (Carmona, Delarue and Lacker 2017)
- Traders market (Cardaliaguet and Lehalle 2017)
- Macro-economic growth model (Achdou et al. 2014)

We introduce a MFG model with incomplete information :

- Cost function depends on a fixed parameter called *state*, that is unknown to players.

- Players get a private stream of signals about the state during the game.

- An (unknown) state $S \in \mathbb{R}^m$, such that $S \sim \mathcal{N}(0, Id)$.
- A signalling process (Z_t) for the representative player, that is a process on \mathbb{R}^m that follows the equation ($\sigma > 0$)

$$dZ_t = Sdt + \sigma dB_t.$$

- The representative player has a position $X_t \in \mathbb{R}^n$, that he observes, and that evolves according to

$$dX_t = \alpha(t, X_t, Z_t)dt + \sqrt{2} dB'_t,$$

where (B_t) and (B'_t) are independent Brownians, and α is measurable and chosen by the player.

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- Given the state s , the density of players that are at time t in position x is denoted by $\rho_s(t, x)$.
- Fix $T > 0$. In the problem starting from t , the player minimizes

$$\mathbb{E} \left(\int_t^T \frac{1}{2} |\alpha(t, X_t, Z_t)|^2 + F(S, X_t, \rho_s(t, X_t)) dt \right),$$

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For all t , the law of $S|(Z_u)_{0 \leq u \leq t}$ is $\mathcal{N}(r_t(Z_t), \sigma_t^2)$, where

$$r_t(z) = \frac{z}{t + \sigma^2} \quad \text{and} \quad \sigma_t^2 = \frac{\sigma^2}{\sigma^2 + t}.$$

- Given $(Z_u)_{0 \leq u \leq t}$, the law of S depends only on t and Z_t .
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Definition

Let m^0 be an initial player density over \mathbb{R}^n . For each $s \in S$, let $m_s : [0, T] \times \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}$ "regular enough", such that for each t , $m_s(t, \cdot)$ is a probability density over $\mathbb{R}^n \times \mathbb{R}^m$.

The distribution $m = (m_s)_{s \in S}$ is an equilibrium if there exists a control α s.t. :

- α is a minimizer in the control problem $\Gamma(t, x, z)$ with cost

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- For any s and t , the distribution of (X_t, Z_t) under control α , given that $X_0 \sim m^0$ and $S = s$, is $m_s(t, \cdot)$.

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Assume for this slide that the state is known (standard case).

Let m be a distribution. Let $u(t, x)$ be the optimal cost in $\Gamma(t, x)$. Then m is an equilibrium if and only if (m, u) satisfies the MFG system

$$-\partial_t u + \frac{1}{2} |D_x u|^2 - \Delta_x u = F(x, m) \quad (\text{Hamilton-Jacobi equation})$$

$$\partial_t m - \operatorname{div}_x(D_x u m) - \Delta_x m = 0 \quad (\text{Fokker-Planck equation})$$

- The Hamilton-Jacobi equation corresponds to a dynamic programming principle for $\Gamma(t, x)$.
- The Fokker-Planck equation describes the evolution of the density distribution of the process (X_t) , given that $dX_t = \alpha(t, X_t)dt + \sqrt{2}dB_t$, and $\alpha(t, X_t) = -D_x u(t, X_t)$.

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For $(t, x, z) \in \mathbb{R}_+ \times \mathbb{R}^n \times \mathbb{R}^m$, and $m = (m_s : [0, T] \times \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R})_s$, let

$$\tilde{F}(t, x, z, m) = \mathbb{E}_{t,z}(F(S, x, \rho_S(t, x))),$$

where the expectation $\mathbb{E}_{t,z}$ is w.r.t. $S \sim \mathcal{N}(r_t(z), \sigma_t^2)$, and $\rho_S(t, \cdot)$ is the marginal of $m_s(t, \cdot)$ on \mathbb{R}^n .

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$$-\partial_t u + \frac{1}{2} |D_x u|^2 - r_t(z) \cdot D_z u - \Delta_{x,z} u = \tilde{F}(t, x, z, m)$$

$$\forall s \in S \quad \partial_t m_s - \operatorname{div}_x(D_x u m_s) + \operatorname{div}_z(s m_s) - \Delta_{x,z} m_s = 0.$$

The derivation is similar to the standard MFG model. Indeed,

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Proposition (Shmaya and Z.)

For each initial probability distribution, there exists an equilibrium. Moreover, $\alpha^*(t, x, z) = -D_x u(t, x, z)$ is an optimal control.

Proof : same as in the complete information case :

- Given a distribution m , consider the optimal cost u associated to m
- Let m' be the distribution generated by the optimal control α^* .
- Apply a fixed-point argument to the mapping $m \rightarrow m'$.

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Monotonicity assumption

For all $s \in \mathbb{R}^m$, for all $\mu^1 \neq \mu^2 \in \mathcal{P}(\mathbb{R}^n)$,

$$\int_{\mathbb{R}^n} (\mu^1(x) - \mu^2(x)) (F(x, \mu^1, s) - F(x, \mu^2, s)) dx > 0.$$

Remark : this holds when $F(x, \mu, s) = \mu(x)$.

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- Let m^1 and m^2 be two equilibria such that for all s , $m_s^1(0, \cdot) = m_s^2(0, \cdot) := m^0$. Let $\rho_s^i(t, \cdot)$ be the marginal of $m_s^i(t, \cdot)$ on \mathbb{R}^n .
- Denote by $\gamma(i, j)$ the cost of the player when he plays α_i and at each time t , the other players are distributed according to ρ^j , and the initial position of the player is randomized according to m^0 .

$$\gamma(i, j) := \mathbb{E} \left[\int_0^T \int_{\mathbb{R}^n} [\alpha_i(t, x, z)^2 + F(x, \rho_S^j, S)] m_S^i(t, x, z) dt \right].$$

$$\begin{aligned} & [\gamma(1, 2) - \gamma(2, 2)] + [\gamma(2, 1) - \gamma(1, 1)] = \\ & -\mathbb{E} \left[\int_{\mathbb{R}^n} [\rho_S^1(t, x) - \rho_S^2(t, x)] [F(x, \rho_S^1, S) - F(x, \rho_S^2, S)] dx \right] \end{aligned}$$

Thus, if $\rho^1 \neq \rho^2$, the player has either a profitable deviation in the equilibrium m^1 or in m^2 , which is absurd.

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Thus, if $\rho^1 \neq \rho^2$, the player has either a profitable deviation in the equilibrium m^1 or in m^2 , which is absurd.

- 1 General Model
- 2 Product Differentiation

- A continuum of firms sell the same object,
- The preferences of the consumers are unknown, and firms get signals about them,
- Firms aim at selling an object that reflect the preferences of consumers, but without facing too much competition.

- Position $x \in \mathbb{R}^n$ of a firm=characteristics of the object it sells
- Preferences of the consumer are represented by a state $S \in \mathbb{R}^n$.
- Cost function :

$$F(x, \mu, s) := g(|x - s|) + \int_{\mathbb{R}^n} h(-|x - y|)\mu(y)dy,$$

where g and h are continuous, increasing and integrable.

- Does more precise information leads to a better equilibrium ?
- Recall that players receive a stream of signals (Z_t) such that $dZ_t = Sdt + \sigma dB_t$.
- Is the equilibrium cost increasing in σ ?
- Simpler question : is the equilibrium better for $\sigma = 0$ rather than for $\sigma = +\infty$?

Proposition (Shmaya and Z.)

Assume that

$$\mathbb{E}_{S \sim \mathcal{N}(0, Id)}(g(|S|)) - g(0) > \|h\|_\infty.$$

Then for large duration of time T , the equilibrium for $\sigma = 0$ is better than the equilibrium for $\sigma = +\infty$.

Thank you for your attention !